Response to reviewers: "Watershed classification for the Canadian Prairies"

Please note that we have changed the manuscript title to: "A WATERSHED CLASSIFICATION APPROACH THAT LOOKS BEYOND HYDROLOGY: APPLICATION TO A SEMI-ARID, AGRICULTURAL REGION IN CANADA".

Page and line numbers are shown for each change in reference to the marked-up version.

Response to Referee #1

Response to SPECIFIC COMMENTS

1. L430 (comment 5): the authors provide more detail about the accuracy of HydroSHEDs and state that "[…] the dataset provided an objective delineation over the region of interest and was sufficient for purpose of the current study." This argument would gain in strength if the authors can add how they came to this conclusion.

> *We thank the reviewer for drawing our attention to this comment. Our conclusion was based on data availability that both covered the geographic scale and resolution (i.e., 100 km^2) necessary for the purposes of our study. However, in light of the reviewer's comments, we added clarity to this sentence:*
>
> *"However, the dataset watershed delineations over the geographic region of interest and at a fine enough scale (i.e., 100 km^2), and thus, it was sufficient based on data availability for purpose of the current study." (7, 174)*

2. L474 (comment 8): the authors provide additional context for choosing the Thornthwaite PET method (which I think is justified) and also state a disadvantage of the method. It might be helpful to also include the practical impact of this disadvantage, because I don't quite understand.

> *We thank the reviewer for raising this concern, which was also raised by Referee #2. We have added the following to increase clarity on the impact of this method and assumption:*
>
> *"A disadvantage of the Thornthwaite approach is that it calculates PET solely as a function of air temperature and latitudinal position, and it assumes a fixed correlation between temperature and radiative forcing. As such, it integrates effects of other factors directly or indirectly influencing radiation or latent heat, like advection, vegetation, and humidity. The calculation adjusts for any lag in this relationship using corrections for latitude and month; however, it likely does not represent the full annual and seasonal variability in PET across a landscape, given regional heterogeneity of the aforementioned factors. Despite the limitations, the simplicity of this method is ideal for application across the wide geographic area of interest with limited data required as input, allowing for approximation of mean annual PET for the study area." (8, 209)*

3. L484 (comment 9): the authors provide reasons for not using any metrics related to snow in their study but acknowledge that this might be important. Is this mentioned anywhere in the manuscript? For example as a study limitation or an opportunity for further work.

> *We greatly appreciate the reviewer's comments in regard to the consideration of snow variables. We now reference in the Discussion the limitation of the current study in this regard, and that if*

data is and becomes available, it should be included, or considered, in future classification approaches.

"Where data is available, future work should consider variables related to snow formation and melt, as well the proportion of annual snow to rainfall as these variables are likely influential when describing hydrological behaviour of the watersheds and classes (Knoben et al., 2018; Shook and Pomeroy, 2012). (28, 810)

4. L532 (comment 14): the authors provide a statement about the accuracy of the findings of Spence and Saso (2005). Is this accuracy dependent on the number of observations used? (Spence and Saso (2005) seem to use n = 34, compared to n = 11 in this paper). Addition: I see the authors have clarified this on L1267.

We appreciate the reviewer's comment here. We do expect an impact on the uncertainty based on the smaller sample size used in the current study. As the reviewer indicates, we clarify this expectation when reporting results. To help with this concern, we clarify this expectation in our methods: "We note that greater uncertainty than that reported by Spence and Saso (2005) may result when using the CCA approach with a smaller sample size." (12, 335)

5. L555 (comment 17): the changed text in this response refers to Table 3, but the text in the manuscript refers to Table S3 (L1179).

The reference should be to Table S3, which shows the compositional datasets. We also refer to Table S2 which includes the source datasets. We have made the revision to say "[…] from pre-processing (Table S2; Table S3)". (13, 371)

6. L656 (comment 27): the authors provide evaluation against another data source (section 4.3) and include a sensitivity analysis of their clustering approach (section 4.4). The also point out (as reviewer #1 has mentioned) the rough correspondence between their clusters and the current understanding of eco-regions (section 5.1.2). The authors also comment that further evaluation is difficult due to the lack of data sources (e.g. L678). The authors state that (L928) "… those areas that are climatically and physio-geographically similar, and thus might be expected to respond in a hydrologically coherent manner to climate and land management changes." This is the critical assumption that underlies this clustering exercise. As I understand the manuscript, section 4.3 does not as much evaluate the entire classification, but only a part of it (wetland density). Further demonstrating that the defined clusters indeed respond in a coherent manner will add much more weight to this paper. However, if there is no data available than that is clearly not an option. If this is the case, the authors might want to further highlight the novelty of their work compared to the current understanding of eco-hydrology on the Prairies (e.g. the need for fuzzy treatment of watershed similarity as evidenced by section 4.4; the increased granularity possible with an approach such as the authors use, …).

We appreciate the attention to comment given by the reviewer and editor. The lack of hydrological data available (and thus that available for adequate validation) at the granularity and spatial consistency was one of the motivating intentions of this study. Although only representative of the part of the hydrological response, we show the differences in wetlands size distributions of the classes in Figure 8. Given the relationship with wetlands and hydrological response (citations therein), we also recognize the comparison suggests only potential coherent difference in response. Future work intends to build on the foundation laid in this study and compare the coherent hydrological responses to environmental change; however, we believe *including this analysis in addition to the current study would make the manuscript quite unwieldy. We agree with the reviewer's suggestion to highlight the novelty in our approach, specifically the scale and "fuzzy" boundaries, and we emphasize this in our discussion, such as the following addition: "Our results are novel in that they characterize in greater detail, and at small watershed scales, the potential for different hydrological behaviour of watersheds within the region." (23, 655)*

7. L733: the authors provide more detail about how they scaled variables during the PCA. I'm bringing this point up again in relation to the text on L1326: "Climate and elevation gradients are likely responsible for the west to east watershed clustering pattern." I wonder to what extent this is forced by the data preparation, where these variables are log-transformed but not normalized. Is it possible that the log-transformed range of climate and elevation variables spans a wider range than that of the other variables? For example, if logtransformed mean precipitation has range [0,3] (assuming P = 1 to 800mm) it would span three times the range of a fractional variable with range [0,1]. This might skew the clustering procedure towards treating P and elevation as more distinctive attributes for each cluster. I don't believe this is necessarily a bad thing, for example if there are reasons to believe that P and elevation are relatively important. However, the authors also comment that "[…] if one is particularly interested in such variables, one should consider strategies to weight their importance." Is it possible that some form of weighting has already happened in the current manuscript as a result of only log-transforming the variables?

Investigation of the log-transformed ranges of each variable might indicate this. This would be a relatively low-effort check compared to re-doing the full clustering analysis with differently prepared data. If found relevant, this might be added to the discussion in L1595-1602.

We thank the reviewer for this suggestion. We performed the log-transformed range check suggested by the reviewer for each input variable. Upon observation, there does not seems to be a relationship between the log-transformed range and those variables that were influential on the classification procedure. Interestingly, Elevation and Precipitation had relatively low log-transform range (1.8 and 0.8, respectively) compared to other log-transformed variables. It should be noted that because we used annual precipitation, the range in our data would not be between 1 to 800mm. We do reiterate our previous response to this concern in that variables were scaled when the PCA was performed. Our point in the discussion (26, 765) is to indicate that perhaps approaches should consider that some variables that are particularly impactful on a local scale (like the location of the largest pond), and that considering weighting might be a strategy to have a hierarchy in what variables might be considered more important. However, we recognize that a drawback to this approach is to increase the amount of assumptions one makes about the data prior to data analysis. We have added the following to indicate that variable ranges were scaled during the PCA: (1) "Variable unit ranges were also scaled during the PCA to reduce the impact of certain variables exhibiting a large range of values on the subsequent cluster analysis." (13, 365); and (2) "Variable importance in the classification was not related the log-transformed range exhibited by that variable (data not shown), and impact was mitigated by scaling the ranges of input variables in the PCA." (17, 478)

Comments on revised manuscript

8. L877: "regime" > "regimes"?

We have made the edit (4, 83).

9. L1033: I think the term "wet climate cycles" might still be confusing. Would "wet climate periods" be a suitable alternative?

We agree that "wet climate periods" is suitable. We have made the revision (9, 242)

10. L1208: it might be helpful to the reader to briefly summarize why wetland area distributions are simulated, if observations are also available (in the GSW data set). Am I correct in saying the GSW only gives the maximum wetland area, and the GPD simulation gives estimates of the full distribution of wetland sizes?

The wetland distributions were simulated to compare how the use of the Generalized Pareto parameters reflect the observed data based on the GSW dataset. Our simulations were restricted at the lower part of the distribution to reflect the data resolution of the satellite-based data from GSW (we reference this for example in P29, L853). Therefore, these simulations do not predict the "smaller wetlands". However, our results suggest that parameters might be used in future studies to predict across the distribution of wetlands, and they are useful parameters to describe watershed or class wetland size distributions (P29, L857). We have adjusted the section of concern as well to reflect this (14, 398).

11. L1224: "4175 set" > "set of 4175 watersheds"

Thank you. We have made the edit (15, 416).

12. L1299: "TPC3" > "PC3"

We thank the reviewer for the edit, and we have made the change (17, 494).

13. L1425: "[…] less than 1." > "[…] less than 1 km-2"

We have made the edit (21, 611).

14. L1652: This paragraph might be better placed directly after (or as part of) the paragraph that starts on L1610.

We appreciate the suggestion and agree that the discussion on boundaries and analysis should flow accordingly. We have moved the paragraph ahead (27, 796).

15. L1683: Is the reference to Wagener et al, 2007 correct? I don't believe that paper talks about the relation between management practices and classification approaches. Perhaps this should be Wagner et al, 2007 (which I haven't read but its title suggests it as being more likely)?

We appreciate the comment. The reference should indeed be for Wagner et al. 2007. We have made the change (30, 894).

16. L2127, Figure 6b: the number of points make this plot difficult to read. x-axis could be changed to cover the width of the page. Possibly cut of the y-axis at 10 for additional clarity.

Thank you for the suggestions. We have adjusted the width of the figure and transparency of the points in Figure 6b. Transparency was adjusted to show overlap among the points. We also cut the figure at 10 as the trend remains relatively unaltered (two watersheds with densities higher than 10 km^{-2} are not visualized).

17. Figure S1, c: text in the centre overlaps and is unreadable.

The soil zones (Dark gray and gray) in this part of the plot are not represented well by Axis 1 and Axis 2, which is the reason for the overlap. To help with interpretation, we have added to the figure description: "Note that the variables at the centre of plot (c) are "dark gray" and "gray" soil zones and are not well represented by Axis 1 and Axis 2."

Response to Referee #2

Response to GENERAL COMMENTS

The authors have addressed many of my comments on the original manuscript, but did not present adequate response to a number of comments. I have annotated a PDF file (attached) of their response with further comments and suggestions. In addition, in Line 862 and 1510 in the marked manuscript, Hayashi and Rosenberry (2002) is cited as a reference regarding ecoregions, but this paper did not discuss ecoregions at all. Please remove the reference from these sentences.

> *We thank the reviewer for noting this concern. We have removed the references for the respective lines.*

Response to SPECIFIC COMMENTS

Line 117. How is the Canadian Prairie defined? Please present a brief definition, and the source of the ecozone boundary shown in Figure 1.

Comments: I do not see the source in Figure 1.

> *We have added the citation to the figure for the ecozone boundary.*

Line 136-138. As it is written, the sentence indicates that the watershed of the Saskatchewan River is excluded from the analysis, which is clearly not the case.

Comment: This has not been done. Please address the comment.

> *Upon revision, we deemed that this clarification was not needed for describing which watersheds were included in our study region (i.e., those that fall within the ecozone boundary). We have removed the portion of concern, and the text now reads: "Thus, we constrained our study to the Canadian Prairie ecozone (4.7 x 105 km^2) and watersheds occurring therein." (7, 166)*

Line 141. The authors describe watersheds by referring the reader to Figure 1. However, Figure 1 does not show watersheds. Please refer the reader to Figure 5 instead, or add watershed boundaries to Figure 1.

Comment: The reference to Figure 1 has not been removed. Please show the watershed boundaries in Figure 1.

> *Here, the reference to the figure means to address the extent of the Canadian Prairie ecozone, not the watershed boundaries. We have removed the reference here for clarity, as it was determined unneeded (7, 177).*

Line 161. Temperature-index methods such as Thornthwaite do not give reliable estimates of "potential evapotranspiration" … please explicitly acknowledge its limitation.

Comment: This sentence is unclear. Please acknowledge more specifically the well-known bias and error in PET estimates using the Thornthwaite and similar temperature-based methods.

We thank the reviewer for raising this concern, which was also raised by Referee #1. We have added the following to increase clarity on the impact of this method and assumption:

"A disadvantage of the Thornthwaite approach is that it calculates PET solely as a function of air temperature and latitudinal position, and it assumes a fixed correlation between temperature and radiative forcing. As such, it integrates effects of other factors directly or indirectly influencing radiation or latent heat, like advection, vegetation, and humidity. The calculation adjusts for any lag in this relationship using corrections for latitude and month; however, it likely does not represent the full annual and seasonal variability in PET across a landscape, given regional heterogeneity of the aforementioned factors. Despite the limitations, the simplicity of this method is ideal for application across the wide geographic area of interest with limited data required as input, allowing for approximation of mean annual PET for the study area." (8, 209)

Line 162. The balance between precipitation and evapotranspiration is reflected in ecoregions of the Prairie, as plants are good indicator of long-term water balance. … Please provide an explanation.

Comment: This response is missing the point. Ecoregions are defined by the optimal vegetation community reflecting the climatic condition, not the actual land use and agriculture. Please present a more meaningful response.

We recognize that ecoregions are defined by the vegetation community, which results from climatic conditions. We instead used data from the CANGRID product to approximate the water balance across the region. Using the ecoregions to estimate the long-term water balance, while an appropriate method to use, would confine these estimates to the respective boundaries of the ecoregions. We aimed to provide an analysis independent of the pre-defined ecoregions (but see our discussion in section 5.1.2). Using the CANGRID product allowed for gradients of both precipitation and PET to be approximated, which we deemed preferable for the purpose of classifying sub-regions of watersheds.

Line 191. Please briefly explain the meaning of mu and beta, and indicate the dimension or unit. These must have a unit of area to maintain the dimensional homogeneity.

Comment: Please indicate the dimension of beta.

We explain the meaning behind the scale (β) and shape (ξ) parameters as well as their units (10, 259-264). We have replaced "scale" and "shape" with the Greek letter notation, respectively, for clarity.

Line 205. Surficial geology is mapped by geologists in each province using different terminologies. I am not sure if the "comparison across provincial boundaries" is straight forward. Please add a brief explanation on how the difference in terminology and mapping methods was reconciled.

Comment: Please add the explanation for this procedure in the texts.

We appreciate the request of more detail in this regard. We added clarity to the text, and it now reads: "Due to the different geological classification schemes for each province, more detailed classes were grouped to broader categories related to depositional environment and surficial materials using those from the Geological Survey of Canada (2014), which provided for *comparison across provincial boundaries." (11, 291). We also added this citation to the references section.*

Line 266. What are "the original variables"? Please explain, using a table if appropriate.

Comment: I do not see the reference to the table. Please address the comment.

Here, we are briefly describing the CCA method, and the "original variables" refer to those physico-climatic and hydrological input in the analysis. As such, we rescind the reference to the Table 2. We have adjusted the sentence to: "Briefly, CCA correlates the streamflow record of gauged basins to physico-climatic characteristics of watersheds by representing these variables as a reduced set of canonical variables. The analysis results in two canonical variable sets: one for the physico-climatic variables (i.e., V1 and V2) and another for the hydrological variables (i.e., W1 and W2). These canonical variables are constructed from linear combinations of the variable sets such that the correlation of the canonical variables are maximized." (12, 343)

Line 301. Please define alpha.

Comment: I do not see the definition. Please address the comment.

The term α refers to the level in which statistical significance was determined. We removed this reference to alpha and replaced it with "$p < 0.05$". (14, 395)

Line 310. What does this mean? Based on Line 269, does it mean that the result was very useful for V1-W1, and barely useful for V2-W2? Please explain.

Comment: The adjusted sentence does not address this comment. Please make a more meaningful adjustment.

We thank the reviewer for their feedback. We have re-arranged the sentences with the aim of adding more clarity. Here, we are indicating that because λ values were high, which indicate that physico-climatic variables represented trends in hydrological variables, we might choose from either set of canonical correlations. Although $\lambda2$ is slightly lower, the second canonical correlations for the hydrological variables (W2) were higher, and since we were interested in predicting these variables, we opted to use the second canonical correlations for physico-climatic variables (V2) in the regression. The paragraph now reads: "The canonical coefficients from the CCA were acceptably high at $\lambda1$ 0.97 and $\lambda2$ 0.77, respectively, indicating that the physico-climatic variables exhibited influence on the hydrological variables (Cavadias et al., 2001; Spence and Saso, 2005). Mean canonical correlation values between the hydrological variables and W2 were greater than those with W1 (Table 2); thus, the physico-climatic variables strongly associated to second canonical correlation (i.e., V2) were used in the multiple regressions." (16, 452)

Line 311. What correlation value would indicate "strong"? Does it have a statistical level of significance, like in the standard correlation analysis? Does a negative value indicate negative correlation? Please explain.

Comment: The minor modification in the sentences does not specifically address the comment.

We evaluate the strength of canonical correlations based on Cavadias et al. (2001) as those above 0.75. The specific method did not evaluate a level of significance, as is the case for other correlation analyses. A negative correlation value between the physico-climatic variables and a canonical component (e.g., V2 and Area, Table 2) describes the relationship in the canonical space, and does not necessitate a negative relationship with Q2 or Q100. The influential of the physico-climatic variables on those hydrological variables was determined by the multiple regression.

Line 311-312. It is true that the correlation value is strong between Q100 (1:100 flow) and W2, but it is weak for Q2 (mean annual flow) and W2. On the other hand Q2 and W1 has a strong correlation. Also the lambda value is much greater for V1-W1 combination than for V2-W2 combination. Given that, why was W2 chosen? Is it because the classification is designed for 1:100 flood prediction? Please provide an explanation.

Comment: This is not true. For Q2, W1 has a stronger correlation than W2. Please provide an objective explanation in the texts.

Please refer to the change mentioned for (Line 310). We chose to use V2 and W2 because both of the hydrological variables exhibited adequate relationships with W2 and a selection of physico-climatic variables were related to V2. In opposition, V1 was not associated with many of the variables, with the highest being 0.64.

Line 347. What are the "PCs from compositional datasets"? Are these different from PC1-PC6 in the header of Table 3? Please explain.

Comment: I do not see a new figure, or reference to it. Please address the comment.

We added reference to the compositional datasets to the heading of 4.1.1 to have consistent language. We also reference Table 1 to refer to the compositional datasets and the number of components used in the clustering analysis (17, 474).

Line 358. "Weaker", not "less strong".

Comment: This has not been revised.

We thank the reviewer for identifying that this was not changed in text. We have made the change to "weaker". (17, 485)

Line 472. Are there 11 study watersheds, as indicated in Line 255? If so, is that a high enough number to examine all seven classes? Please explain.

Comment: This does not address the comment. Please discuss the limitation of using hydrological data from only 11 watersheds.

We recognize that there is a limitation in the current approach for the 11 watersheds to represent the watersheds in the cluster analysis, and that this is an approximation of runoff. We have referenced this limitations in the text: "Ideally, a more detailed estimate of runoff for each *watershed would be a valuable contribution. In the current study, we used the CCA and eleven reference stations to approximate runoff values for the clustering watersheds. Given the number of watersheds included in the analyses, the diversity of physical characteristics and potential hydrological behaviour is likely not completely represented in the small sample size of available hydrometric stations and represents a limitation of this approach." (28, 829). We also note the limitation and potential impact to uncertainty in our methods, as per request be Referee #1 (12, 335).*

Line 637. Geography may not be an appropriate term here, because geography encompasses many things, not just landforms. I would say topography or landform is more appropriate.

Comment: This has not been done. Instead, geography has been replaced by physio-geography, which is likely an incorrect spelling of physiography. Please note that physiography is a broad term including the effects of climate, topography, hydrology, and all other variables in physical geography.

We thank the reviewer for their insight on the use of this term. We have revised the use of these terms throughout the manuscript for consistency.

A WATERSHED CLASSIFICATION APPROACH THAT LOOKS BEYOND HYDROLOGY: APPLICATION TO A SEMI-ARID, AGRICULTURAL REGION IN CANADA

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ABSTRACT

Classification and clustering approaches provide a means to group watersheds according to similar attributes, functions, or behaviours, and can aid in managing natural resources. Although they are widely used, approaches based on hydrological response parameters restrict analyses to regions where well-developed hydrological records exist, and overlook factors contributing to other management concerns, including biogeochemistry and ecology. In the Canadian Prairie, hydrometric gauging is sparse and often seasonal. Moreover, large areas are endorheic and the landscape is highly modified by human activity, complicating classification based solely on hydrological parameters. We compiled climate, geological, topographical, and land cover data from the Prairie and conducted a classification of watersheds using a hierarchical clustering of principal components. Seven classes were identified based on the clustering of watersheds, including those distinguishing southern Manitoba, the pothole region, river valleys, and grasslands. Important defining variables were climate, elevation, surficial geology, wetland distribution, and land cover. In particular, three classes occur almost exclusively within regions that tend not to contribute to major river systems, and collectively encompass the majority of the study area. The gross difference in key characteristics across the classes suggests that future water management and climate change may carry with them heterogeneous sets of implications for water security across the Prairie. This emphasizes the importance of developing management strategies that target sub-regions expected to behave coherently as current human-induced changes to the landscape will affect how watersheds react to change. The study provides the first classification of watersheds within the Prairie based on climatic and biophysical attributes, with the framework used being applicable to other regions where hydrometric data are sparse. Our findings provide a foundation for addressing questions related to hydrological, biogeochemical, and ecological behaviours at a regional level, enhancing the capacity to address issues of water security.

A WATERSHED CLASSIFICATION APPROACH THAT LOOKS BEYOND HYDROLOGY: APPLICATION TO A SEMI-ARID, AGRICULTURAL REGION IN CANADA

1. INTRODUCTION

Watershed classification methods provide a means of grouping watersheds according to similar attributes, or behaviours, and can identify sub-regions that are expected to exhibit coherent responses. This strategy can identify how catchment characteristics are similar, or dissimilar, among groups of watersheds and thus might influence hydrologic behaviour (McDonnell and Woods, 2004). Classifying watersheds can be useful for developing predictions in ungauged basins (Peters et al., 2012), and moreover, classification can be used to inform how changes to key traits (e.g., climate and land management) may affect system function. Establishing these links between watershed function and biophysical structure, including hydroclimate, is an opportunity of watershed classification (Wagener et al., 2007). Accordingly, the regionalization of hydrological response through watershed classifications has been used to inform natural resource management (Detenbeck et al., 2000; Jones et al., 2014).

Many different approaches to watershed classification have been employed to date, including non-linear dimension reduction techniques (Kanishka and Eldho, 2017), decision trees (Bulley et al. 2008), and independent component analysis (Mwale et al., 2011), among others. Hydrological characteristics (e.g., statistical properties of streamflow regime) are widely used to inform classification owing to their potential linkages between watershed features and hydrologic responses (Brown et al., 2014; Sivakumar et al., 2013; Spence and Saso, 2005). Other classification exercises have included a wider number of characteristics, including biophysical attributes along with streamflow response, to differentiate watershed classes (e.g., Sawicz et al., 2014;~~,~~ Burn, 1990). Ecoregions, which incorporate historical aspects of climate, topography, and vegetation regimes, have also served as a method of differentiation for eco-hydrological studies (~~Masaki and Rosenberry, 2002;~~ Loveland and Merchant, 2004). In select cases, classification is performed independently of streamflow response factors (Knoben et al., 2018). In arid or poorly gauged regions of the world, these types of approaches to classification that are independent from or not strongly dependent on hydrological indices (streamflow response), are needed, although few such classifications have been performed. The need for new approaches to watershed classification can also be true of regions undergoing strong pressures from climate change and land-use, where historical streamflow records may not reflect current behaviour, particularly if a regime shift has occurred.

In Canada, watershed classification has been applied in many regions (e.g., Cavadias et al., 2001; Ouarda et al., 2002; Spence and Saso, 2005). To date, most have focused on larger basins, and none have covered in detail the semi-arid Canadian Prairie, which spans nearly 5 x 10^5 km^2 in western Canada, from the Rocky Mountain foothills in the west to Lake Winnipeg in the east (Fig. 1). This is despite its importance as a major food producing region of the world and one that faces numerous water security challenges (Gober and Wheater, 2014; Spence et al., 2018). Earlier work by Durrant and Blackwell (1959) grouped large Prairie watersheds based on flood regimes. A recent classification that included the Prairie region focused on stream hydrology (e.g., MacCulloch and Whitfield, 2012) but was broader and included watersheds from mountainous and forested regions to the west and north, respectively. In the Canadian Prairie, and similar regions elsewhere, extrapolating catchment-scale field and modelling studies presents challenges. It is inherently difficult to explain or predict different responses among basins, as poorly developed stream networks with intermittent or seasonal flow do not easily lend themselves to classification methods featuring streamflow response. MacCulloch and Whitfield (2012), who found a single streamflow class across the Canadian Prairie, raised the question as to whether a single grouping is appropriate, and suggested the need to expand classifications to include a greater diversity of biological, physical and chemical properties.

Like many of the world's agricultural regions, the Canadian Prairie has undergone vast environmental change co-incident with the green revolution. Predominant agricultural practices have changed over the decades, and each is known to influence water cycling and storage, including tillage practices, summer fallowing, and cropping type (Awada et al., 2014; Van der Kamp et al., 2003; Shook et al., 2015). Significant warming over the last 70 years, especially in winter (Coles et al., 2017; DeBeer et al., 2016) has resulted in more rain at the expense of snow (Vincent et al., 2015), and multiple-day rainfall events have been increasing in frequency relative to shorter events in some regions (Dumanski et al., 2015; Shook and Pomeroy, 2012). These observed changes in precipitation have reduced the predictability of runoff derived from snowmelt, and add uncertainty to water management and agricultural decision-making.

Field Code Changed

Field Code Changed

Disentangling the relative impacts of climate and land-use changes on water quantity and quality is complex, particularly as their effects are heterogeneous across spatial extent and scale. For the Prairie and elsewhere, new approaches to classification that can distinguish sub-regional and, importantly, sub-hydrometric station variability, are needed. Further, because land management decisions in agricultural regions are intrinsically linked to system function, there is a need for classifications that can inform decision-makers at a relevant scale. Indeed, stable isotope-based investigations of runoff from small lake catchments in the Boreal Plains (north of the Prairie) emphasize the need for local-scale characterization of watershed behaviour (Gibson et al., 2010, 2016), while streamflow dynamics for the Prairie and nearby Boreal Plain are linked to local surface geology and land cover (Devito et al., 2005; Mwale et al., 2011), suggesting an opportunity for a new approach to watershed classification in the region. Another potential advantage of a more comprehensive approach is that by de-emphasizing available hydrometric observations for larger and well-studied or monitored basins and including other environmental characteristics, the risk of overlooking other functions (e.g., ecology, biogeochemistry) that may be equally important to the management of a watershed's natural resources can be reduced. A system-based watershed classification for the Prairie that avoids the prejudice of classifying only those watersheds where a reasonably robust understanding of hydrology or streamflow exists can serve as a template for other regions of the world where streamflow-based classification is not viable.

The objective of the present work is to develop a watershed classification system based on hydrologically and ecologically significant traits for the Canadian Prairie. In this region, assessment of localized hydrological response to change is challenged by limited spatial resolution of observed streamflow data, and higher order streamflow being unrepresentative of local response due to a poorly-developed drainage network. In establishing such an approach, we seek to advance our understanding of watershed hydrology and broader watershed behaviour within the Prairie whilst also providing a framework for similar classification exercises in other regions where streamflow-based methods are not ideal. Our approach avoids the limitations of classifying according to known hydrologic response, and increases the spatial resolution of watershed classification relative to many existing approaches. We compile physio~~-geo~~graphic characteristics, including geology, wetland distribution, and land cover, of watersheds approximately 100 km^2 to achieve the classification. This framework will identify those areas that are climatically and ~~physio-~~geographically similar, and thus might be expected to respond in a hydrologically coherent manner to climate and land management changes. Additionally, it provides a foundation on which to base prediction of watershed hydrologic, biogeochemical and ecological responses to these stressors.

2. DATA COLLECTION & COMPILATION

2.1. Region domain and description

The Canadian Prairie (Prairies ecozone) spans the provinces of Alberta, Saskatchewan, and Manitoba, and is part of the Nelson Drainage Basin (Fig. 1). Climate is semi-arid, with mean annual precipitation ranging between 350 and 610 mm (1970–2000) increasing from west to east. Mean annual temperature was 1–6˚C over the same period with warmer conditions towards the southwest (Mekis and Vincent, 2011; Vincent et al., 2012; http://climate.weather.gc.ca/climate_normals/index_e.html). Much of the region deglaciated during the Late Pleistocene approximately 10,000 years before present, resulting in an often hummocky landscape with numerous depressions. Combined with the dry climate, the relatively short post-glaciation history has prevented maturing of a ubiquitous drainage network, and many headwaters remain disconnected from higher order streams (Shook et al., 2015). Depressions in the hummocky landscape, and the wetlands that form within them, are important features for Prairie hydrology (Van der Kamp et al., 2016) and often facilitate groundwater recharge (i.e., depression-focused recharge) (Van der Kamp and Hayashi, 2009). The location of wetlands and their size, relative to the watershed outlet controls hydrologic gate-keeping (e.g., Spence and Woo, 2003), and thus the potential to contribute streamflow to higher-order watersheds (Leibowitz et al., 2016; Shaw et al., 2012; Shook et al., 2013). The size distribution of wetlands within a watershed and their spatial arrangement also dictate biogeochemical function and provide habitat and foraging for biota (Evenson et al., 2018). Terrestrial vegetation is typically open grassland, with aspen parkland ecotone along the northern edges of the ecozone boundary (Ecological Stratification Working Group, 1995).

2.2. Watershed boundaries

The focus of this study was on those watersheds that drain a distinctively prairie landscape, with watersheds defined according to topographic delineation. Thus, we constrained our study to the Canadian Prairie ecozone (4.7 x 10^5 km^2) and watersheds occurring therein~~;~~ ~~watershed areas of larger exotic streams in the region originating in the Rocky Mountains to the west were not included~~. Delineations of candidate study watersheds were obtained from the HydroSHEDS global dataset (Lehner and Grill, 2013). Watershed boundaries within this dataset were based on Shuttle Radar Topographic Mission (SRTM) digital elevation model (DEM) calculated at a 15 arc-second resolution. The resolution is equivalent to for example approximately 285 m east-west and 464 m north-south at Saskatoon, SK. As with other SRTM products, the HydroSHEDs dataset may be prone to errors in regions with low relief due to elevation precision of 1 m. However, the dataset provided ~~an objective~~watershed delineations over the geographic region of interest and at a fine enough scale (i.e., 100 km^2), and thus, it was sufficient based on data availability for purpose of the current study.

Only those watersheds completely within the Canadian Prairie ecozone ~~(Fig. 1)~~ were extracted (n = 4729) from the HydroSHEDs dataset. Those watersheds that were very large (>4000 km^2) or small (<5 km^2) were removed from analysis (see Table S1). Because HydroSHEDs includes the basins of larger water bodies, including lakes, watersheds consisting of a majority of water were removed as the study only concerns the uplands of these systems. Finally, highly urbanized areas (i.e., watersheds with cover being >40% urban) were removed. After considering these criteria, 4175 watersheds remained for use in subsequent analyses, covering a total area of 4.2 x 10^5 km^2. Mean watershed area for this subset was 99.8 ± 58.7 km^2.

2.3. Physio~~-geo~~graphic data collection

The physio~~-geo~~graphic watershed variables were assembled from Canadian provincial and federal governments and non-governmental agency datasets (see Table S2 for a full list of variables and their sources). Variables were derived from climatic, hydrologic, geological, geographic, and land cover data, and details are described briefly below. Spatial processing and statistical analyses were conducted in ArcGIS version 10.5 and R version 3.4.3 (R Core Team, 2018), respectively.

2.3.1. Climate

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Mean annual precipitation and temperature data were derived from the Canadian Gridded Temperature and Precipitation Anomalies (CANGRD) dataset spanning (ECCC, 2017). CANGRD is the only gridded climate product available for the region that uses adjusted and homogenized station data, and was picked for this reason (Mekis and Vincent, 2011; Vincent et al., 2012). The 1970–2000 period was chosen because the number of stations with adjusted and homogenized data used to derive CANGRD significantly diminished after 2000 (Laudon et al., 2017). Mean annual values over the 30-year period were constructed from 50 km resolution gridded cells (n = 626) within and surrounding the Prairie ecozone, and interpolated to a higher spatial resolution raster by kriging using a spherical semivariogram. Values were clipped according to the watershed boundaries, and averaged over the watersheds to obtain mean annual precipitation and temperature for each watershed. Mean annual potential evapotranspiration (PET) was derived as a measure of dryness across the region. To maintain consistency among climate data, and use the same temperature data as described above, options were limited with which to calculate PET. The Thornthwaite equation (Thornthwaite, 1948) was applied using the R package *SPEI* (Vicente-Serrano et al., 2010). A disadvantage of the Thornthwaite approach is that it ~~assumes~~ calculates PET solely as a function of air temperature and latitudinal position, and it assumes a fixed ~~a c~~correlation between temperature and radiative forcing. As such, it integrates effects of other factors directly or indirectly influencing radiation or latent heat, like advection, vegetation, and humidity. ~~and~~ The calculation adjusts for any lag in this relationship using corrections for latitude and month; however, it likely does not represent the full annual and seasonal variability in PET across a landscape, given regional heterogeneity of the aforementioned factors. Despite the limitations, the simplicity of this method is ideal for application across the wide geographic area of interest with limited data required as input, allowing for approximation of mean annual PET for the study area.~~.~~

2.3.2. Wetland traits

Large regions within the Canadian Prairie have been designated as being “non-effective”, where they do not contribute flow to the stream network, at least one year in two (Godwin and Martin, 1975). The location of these regions are shown in Figure 1. This definition stems from work by Agriculture and Agri-Food Canada where prairie drainage areas were divided into *gross* and *effective* drainage areas, whereby the former describes the area within a topographic divide that is expected to contribute under highly wet conditions, and the latter is the area that contributes runoff during a mean annual runoff event (Mowchenko and Meid, 1983). Thus, at its simplest, the non-effective area is the difference between the gross and effective drainage area; however, the exact area contributing runoff is dynamic and the controls complex, which include antecedent storage capacity and climatic conditions (Shaw et al., 2012: Shook and Pomeroy, 2015). Briefly, the "non-effective" regions are caused by the intermittent connectivity of runoff among the landscape depressions, which trap runoff, and prevent it from contributing to downstream flow when the depressions are not connected. Trapped surface water can form wetlands (hereafter, inclusively referring to water area ponded in these depressions). These depressions can store water, and are indicative of water storage of the basin. Thus the non-effective portion of a basin is an index of its lack of contribution and is an important quality when considering the hydrological dynamics of this region (Shook et al., 2012).

The Global Surface Water dataset (Pekel et al., 2016) provides a geographically comprehensive layer of any ~30 m x 30 m pixel that was inundated at least once between 1984 and 2015, as identified from the Landsat constellation of satellites. It was assumed that the dataset was indicative of potential maximum wetland coverage, as this period spanned several wet climate periods. As such, "wetland" in this context can include some seasonal ponds (i.e., prairie potholes) as well as larger or more permanent shallow water bodies (but see Section 2.2 and Table S1). Using the R package *raster* (Hijmans, 2017), wetland variables were calculated for each study watershed, including fractional wetland area, and the number of wetlands within the watershed (i.e., wetland density). The ratio of the area of the largest wetland to total wetland area in the watershed was also used as a metric (i.e., W_L). Further, we used the ratio of the linear distance of the largest wetland's centroid to the watershed outlet (L_W), to the maximum watershed boundary distance to the outlet (L_O) to represent a centroid fraction (L_W/L_O; i.e., the relative location of the largest wetland to watershed outlet). The basin outlet was defined as the point of lowest elevation on the watershed boundary. Both W_L and L_W/L_O can be used to evaluate the relative importance of hydrological gate-keeping; for example, larger wetland depressions located closer to the outlet control the likelihood of the watershed contributing flow downstream and attenuating peakflow (Shook and Pomeroy, 2011; Ameli and Creed, 2019).

To estimate wetland size distribution, it was assumed that they followed a Generalized Pareto Distribution (GPD) defined according to (Shook et al., 2013):

$$F(z) = GPD(\mu, \beta, \xi) = 1 - \left[1 + \xi\left(\frac{z - \mu}{\beta}\right)^{-1/\xi}\right] \tag{1}$$

Where z is wetland area, and μ is the location parameter (i.e., the minimum size for which the distribution was fitted and has units of m^2), and the scale (β) and shape (ξ) parameters are determined for each watershed. The β~~scale~~ parameter is an index of the dispersion of the distribution, similar to the standard deviation, with the same units as the data being fitted (in this case m^2). The ξ~~shape~~ parameter is dimensionless and, as its name suggests, governs the shape of the fitted distribution. Hosking and Wallace (1987) plot the effect of variation in the shape parameter on the GPD. The scale and shape parameters were used to quantify the size distribution of wetlands and thus to describe the wetland frequency distributions for the cluster analyses (see 3.2). Note that because the sizes of the water bodies were taken from infrequent remote-sensing measurements (i.e., the Landsat data have a minimum revisit time of 8 or 16 days), they also are biased against short-lived water bodies.

2.2.3. Topographical parameters

Topographic variables including the mean elevation, mean and coefficient of variation of slope, and stream density were also calculated for each watershed. Because of the hummocky nature of many regions in the domain, it is possible for a basin to have some fraction of its area located at an elevation below that of the outlet. As such, the fraction of area below the basin outlet (A_{BO}) was calculated for each basin. The elevation and slope variables were based on a DEM generated from the SRTM dataset. Stream vectors were obtained from the Hydrographic features CanVec (1:50000) series available from Natural Resources Canada (https://open.canada.ca/data/en/dataset?q=canvec&sort=&collection=fgp). The total length of streams within a watershed was calculated, and divided by the watershed area to produce the stream density. Additionally, the dimension shape factor (DSF) was used to describe watershed shape, as it has been found important for hydrological responses in previous Canadian catchment classification exercises (Spence and Saso, 2005). The DSF (km^{-1}) was calculated as follows:

$$DSF = \frac{(0.28 \cdot P)}{A} \tag{2}$$

Where P (km) and A (km^2) are the watershed perimeter and area, respectively, and derived from the HydroSHEDS global dataset (Lehner and Grill, 2013).

Geographical parameters of surficial geology, local surface landforms, soil particle size classes (sand, silt, clay), and soil zone were included in the analysis. Surficial geology polygons were derived by compiling provincial government data sources for Alberta (Atkinson, 2017), Saskatchewan (Simpson, 2008), and Manitoba (Matile, 2006). Due to the different geological classification schemes for each province, more detailed classes were grouped to broader categories related to depositional environment and surficial materials using those from the Geological Survey of Canada (2014), which provided for comparison across provincial boundaries. Local surface form (i.e., areas categorized by slope, relief, and morphology) and soil zone data were obtained from the Soil Landscape dataset (AAFC, 2013). The soil zones in the Canadian Prairie, used in the analyses were black, dark brown, brown, gray, and dark gray. The zones incorporate characteristics of colour and organic content, which are influenced by regional climate and vegetation. Clay, silt, and sand content were collected from the Detailed Soil Survey of Canada (AAFC, 2015). Mean catchment values ~~of each~~ of surficial geology, local surface landform, soil zone, and particle size class were determined by areal weighting of soil polygons within the watershed boundaries.

2.3.4. Land cover and cropland practice

Fractional areas of land-use types were derived from the Agriculture and Agri-Food Canada’s 2016 Annual Crop Inventory (AAFC, 2016). These raster data define land-use and land cover. Variables used in our analysis were standardized to watershed area and included unmanaged grasslands, forests (i.e., the sum of coniferous, deciduous, and mixed forest areas), pasture, and cropland (sum of cropped land areas). Predominant cropland practice was defined according to the fractional area of tillage by agricultural region sub-division (e.g., normalized to the area prepared for seed within that division by year). Averaged areas over the years 2011 and 2016 for each practice, including zero-till, conservation till (leaving crop residue on soil surface), and conventional till (incorporating residues into soil) (Statistics Canada, 2016), were used to describe these activities, and normalized as a fraction of the watershed.

2.3.5. Hydrological variable calculation

The relatively sparse hydrometric stream gauging in the domain, and the resulting paucity of data, presents two notable challenges to hydrologic response-based watershed classification. The first is that the basin network is biased to stations on higher-order (and often exotic) streams traversing the region (i.e., larger river basins), and thus there are a limited number of hydrometric gauges on streams draining solely Prairie watersheds, particularly at the spatial resolution of our study watersheds (~100 km^2). Further, only a subset of these are considered reference stations (i.e., gauging unmanaged flows). Second, in the more arid and/or cold regions of the Prairie, some of these hydrometric stations are operated only seasonally, presenting additional challenges in using these records for classification exercises (e.g., MacCulloch and Whitfield, 2012).

As a result, mean annual runoff (Q2) and 1:100 year flood (Q100) magnitudes were estimated for the 4175 watersheds using relationships defined from canonical correlation analysis (CCA) to correlate gauged data to multivariate climatic and physio~~-geo~~graphic data according to procedures given by Spence and Saso (2005). According to Spence and Saso (2005), expected uncertainty using these methods approached 50% but exhibited biases of less than 15% (n = 34). Hydrological stations used were those identified in MacCulloch and Whitfield (2012) and within the Prairie region (n = 11), and data were obtained from archived databases of the Water Survey of Canada (https://wateroffice.ec.gc.ca/search/historical_e.html) between 1990–2014. We note that greater uncertainty than that reported by Spence and Saso (2005) may result when using the CCA approach with a smaller sample size. Multivariate geo~~physio-geo~~graphic data were collected as outlined in the above sections according to the watershed boundaries for the hydrological stations. Due to the fact that many watersheds within the HydroSHEDS dataset are likely to drain internally and do not consistently connect to a higher-order stream network, these streamflow data were interpreted as "runoff", meaning the amount of water accumulated within the watershed polygon that drains to its lowest point annually.

Briefly, CCA correlates the streamflow record of gauged basins to physico-climatic characteristics of watersheds by representing ~~the original~~these variables as a reduced set of canonical variables. The analysis results in two canonical variable sets: one for the physico-climatic variables (i.e., V1 and V2) and another for the hydrological variables (i.e., W1 and W2).

These canonical variables are constructed from linear combinations of the ~~original~~ variable sets~~s~~ such that the correlation of the canonical variables are maximized. Canonical variables plotting similarly on X-Y plots (W1-W2 and V1-V2), indicate good correlation (Spence and Saso, 2005). Where canonical correlations (λ_1, λ_2) were above 0.75 (Cavadias et al., 2001), that set of physico-climatic variables was deemed useful for estimating hydrological variables. Those physico-climatic variables passing this threshold were included as variables in a multiple regression to develop a predictive equation for Q2. Analyses were performed using the R package *vegan* (Oksanen et al., 2018).

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3. DATA ANALYSIS

3.1. Pre-processing compositional datasets

Principal components analysis (PCA) was used as a pre-processing step to reduce the dimensionality associated with compositional datasets (e.g., topographical and land cover parameters) (Fig. S1). Using this approach, the principal components (PC) that could cumulatively explain 80% of the variation in a subset of compositional data were included in the subsequent cluster analysis. This procedure identified the major data patterns and aided in reducing the number of zero-weighted variables. Where necessary, variables that were not transformed into PCs were log-transformed to reduce data skewness. Variable unit ranges were also scaled during the PCA to reduce the impact of certain variables exhibiting a large range of values on the subsequent cluster analysis.

3.2. Agglomerative hierarchical clustering of principal components and watershed classification

Clustering analysis was performed on the suite of physio~~-geo~~graphic variables, which included PC variables derived from pre-processing (Table S2; Table S~~S~~3). Agglomerative hierarchical clustering of principal components (HCPC) was used to define clusters of watersheds using the *HCPC* function in the R package *FactoMineR* (Lê et al., 2008;~~,~~ Husson et al., 2009) to apply a PCA on the standardized multivariate dataset of watershed attributes and was the basis for clustering. The majority of physio~~-geo~~graphic variables were included as active variables in the PCA and thus influenced the arrangements of the PCs. In contrast, watershed area, DSF, latitude, and longitude were used only as supplementary variables, and thus did not explicitly affect the clustering analysis. These variables did, however, aid in watershed class characterization and interpretation. The first set of PCs that together explained 50% of the variation in the dataset ($n = 6$) was retained for agglomerative clustering. Retaining these first PCs at a threshold of 50% allowed for clearer focus on main trends in the data and reduced the impact of noise on subsequent analyses, which might occur if subsequent, less influential, PCs were retained.

The agglomerative hierarchical clustering was performed using the Euclidean distances (from the PCA) and Ward's criterion for aggregating clusters. Ward's criterion decomposes the total inertia of clusters into between and within-group variance, and this method dictates merging for clusters (or watersheds) such that the growth in within-group inertia is minimal (Husson et al., 2010). The total inertia is partitioned into within- and between-group inertias. Within-group inertia represented the homogeneity, or similarity, of watershed within a cluster. Consequently, watersheds located close to each other in PC-space were deemed to be similar in their attributes. This approach decomposes the total variability, or inertia, into within- and between-group inertias. Watersheds are grouped according to pairs that minimize within-group inertia (Begou et al., 2015), and are differentiated based on between-group inertia gained by adding clusters. The variables contributing to cluster characteristics were determined by v-tests (Husson et al., 2009), which assessed whether the cluster mean for a given variable was significantly ($\alpha =$ $p <$ 0.05) greater or smaller than the overall mean.

3.3. Comparing class-specific observed and simulated wetland depression data

To compare how well the GPD parameters predicted the observed wetland area distributions from the Global Surface Water (GSW) dataset, wetland size distributions were simulated for each class. Wetland area for select watershed class–specific percentiles (i.e., 25^{th}, 50^{th}, and 75^{th} percentiles) derived from the simulated data were then compared to the wetland areas for corresponding watershed class–specific percentiles of the observed watershed data to assess the potential usefulness of using these parameters in representing wetland size distribution.

For this comparison, the fitted wetland area distributions were constrained in their minimum and maximum values by the Global Surface Water dataset spatial resolution (i.e., the 30 m pixel size) and the median area of the largest wetland observed for each watershed class,

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3.4. Resampling and re-classifying procedure

The robustness of the HCPC procedure on characterizing Prairie watersheds was tested using additional hierarchical clustering on ten subsets of the entire set of 4175. For each iteration, ten percent of watersheds were removed from the original dataset (n = 4175) without replacement, and the remaining watersheds (n = 3757) were then re-analyzed according to the HCPC outlined above (Fig. S1). The number of potential classes allowed was set at seven (k = 7), for consistency with the complete analysis. The resulting classifications were then compared to the classification performed on the complete dataset, with the watersheds being assessed on the percentage of iterations in which they were assigned to the same class as the complete classification. The proportion of membership agreement was calculated and visualized to assess the likelihood of classing watersheds consistently.

4. RESULTS

4.1. Geographical data processing

4.1.1 Dimension reduction: Compositional datasets and ~~Variable~~ principal components analysis

Variation in geology and soil was best explained by two or three principal components (Table 1; Fig. S2). Two PCs captured over 80% of the variation in surficial geology, with PC1 (proportion explained: 73%) positively relating to glacial till deposits and negatively with glaciolacustrine deposits, and PC2 (14%) positively related to riverine or erosive deposits, such as glaciofluvial, alluvial, and eolian deposits. Particle size class data were explained by the first two PCs, where PC1 (75%) was positively associated with sand and negatively associated with silt and clay, while PC2 (14%) was related negatively to silt. Positive PC1 (55%) scores defined the dominance of black soils, and PC2 (43%) described dominance of brown or dark brown soils on positive or negative scores, respectively. Three PCs described the local surface form dataset. PC1 (55%) captured the change from greater portion of hummocky forms to undulating forms, and PC2 (24%) was negatively associated with higher river-incised landscape fraction. The portion of level surface form was negatively related to PC3 (12%).

Three PCs were needed to explain over 80% of the variation in land cover (Table 1; Fig. S2). Land cover PC1 (37%) was positively associated with higher cropland and negatively with unmanaged grassland; whereas PC2 (25%) was negatively associated with higher pasture and forest cover. PC3 was associated with greater fallow and pasture areal proportion (21%). Cropland practice was described by PC1 (90%), with zero-till practices being negatively associated to this component. Although it only explained 9%, PC2 was also retained to describe the change between conventional and conservation till practices, with the practices exhibiting a positive and negative relationship, respectively.

4.1.2 Canonical correlation analysis

The canonical coefficients from the CCA were acceptably high at λ_1 0.97 and λ_2 0.77, respectively, indicating that the physico-climatic variables exhibited influence on the hydrological variables (Cavadias et al., 2001; Spence and Saso, 2005).~~.~~ ~~Mean c~~Canonical correlation values between the hydrological variables and W2 were greater than those with W1 (Table 2);~~, and because both values of λ were acceptably large (Cavadias et al., 2001)~~ thus, the physico-climatic variables strongly associated to second canonical correlation (i.e., V2) were used in the multiple regressions. These variables were watershed area, DSF, areal fraction of rock, and areal fraction of natural area. Plots of observed and predicted runoff Q2 (R^2 = 0.45) and Q100 (R^2 = 0.48) show moderate agreement at lower flow values (Fig. 2). There is a negative bias estimated between 26 and 29%, which is greater than that documented by Spence and Saso (2005) using comparable extrapolation methods, but this is not unexpected because of the smaller sample size in the current study. As Q2 and Q100 exhibited high collinearity, only Q2 was included in subsequent cluster analyses to:

$$\log(Q2) = 0.130*\log(A) - 0.077*\log(N) + 0.117*\log(R) - 0.141*\log(DSF) - 0.620 \quad (3)$$

Where *A* was the watershed area, *N* was the natural area fraction and the sum of grasslands and forest, *R* was the rock fraction area, and *DSF* was the dimensional shape factor of the watershed.

The equation was then used to calculate Q2 for each watershed included in the clustering analysis.

4.2. Watershed classification

4.2.1. Principal component analysis

In total, 29 watershed attributes, including the PCs from compositional datasets (see Table 1), were used in the clustering analysis as active variables, and four were included as supplementary (Table 3). In the pre-clustering PCA, the first six PCs explained 54.3% of data variation, and were retained for the HCPC analysis (Fig. 3). The influence of subsequent PCs declined dramatically, and eleven PCs were required to explain >80%. Variable importance in the classification was not related to the log-transformed range exhibited by that variable (data not shown), and impact was mitigated by scaling the ranges of input variables in the PCA.

Principal components 1 and 2 captured changes in physical, land cover, and wetland characteristics (Fig. 3). PC1 was strongly associated with physical and land cover characteristics, such as elevation, wetland density, and the land cover PCs. PC2 was strongly related to metrics characterising the hydrological landscape, including river and wetland density, non-effective area fraction, landscape surface form, and size of the largest wetland (W_L). Subsequent PCs explained less variation and were more specialized in the variables associated with them. Generally, these PCs were associated with differences in soil zone and texture class, surficial geology, and varying surface land form. A more detailed account of associations of the variables with the PCs is provided below.

PC1 was positively associated with elevation, mean slope, land cover PC2, and surface form PC3, and negatively with, total annual precipitation, soil zone PC1, wetland density, land practice PC1, land cover PC1, and longitude (Table 3; Fig. 3). PC2 was associated with non-effective area fraction, wetland density, β, and surface form PC2, and negatively related to land practice PC1, W_L, and river density. ~~T~~PC3 was positively related to wetland fraction, W_L, ξ, soil texture PC2, and DSF. Watershed area and runoff were negatively associated with PC3.

Variable correlations were ~~less strong~~weaker for the remaining three PCs (Table 3). PC4 was mainly associated with soil texture PC1, surficial geology PC1 and surface landform PC1, characteristic of sandier soil areas featuring glacial till deposits and higher hummocky surface forms, as well as higher mean slope. PC4 was negatively related to land cover PC2. PC5 was related positively to PET, fraction below outlet, and soil zone PC2, and negatively to land cover PC1, river density, and slope CV. Finally, PC6 was mainly associated with soil texture PC2 and land cover PC3, and negatively with surface landform PC2.

4.2.1. Agglomerative hierarchical cluster analysis

Seven clusters were identified from the hierarchical cluster analysis based on the between-group inertia gained by increasing cluster number (k). The HCPC analysis suggested three clusters resulted in the greatest reduction of within-group inertia while minimally increasing k (Fig. 4). Further increasing k refined the separation and differentiation of clusters up to seven ($k = 7$). Minimal added separation was observed up to $k = 9$, and increasing $k > 9$ resulted in little inertia gained between clusters. Thus, seven clusters, or classes, were manually selected based on these observations (Fig. 4).

4.2.3. Class characteristics and interpretation

Our methodology yields sub-regional watershed classes according to climatic, physiographic, wetland, and land cover variables. The seven classes (Fig. 5), are defined by multivariate sets of attributes (Table 4). Influential classifying variables in all classes were mean elevation, total annual precipitation, land practice, surface forms, and wetland density. Other variables influential to class differentiation included fraction of non-effective area, land cover, and soil variables. Climate and elevation gradients are likely responsible for the west to east watershed clustering pattern. Moreover, we observe strong spatial concordance among some classes (Fig. 5), which is likely due to the hierarchical nature of the analysis. For simplicity, we interpret classes based on the variables where large, significant differences in class mean versus the overall mean of the dataset were observed. The classes can be assigned as follows: Southern Manitoba (C1); a Prairie Pothole region (C2, C3); Major River Valleys (C4); and Grasslands (C5, C6, and C7).

Southern Manitoba (C1)

The majority of Class 1 (C1; n = 365) watersheds occurred in the eastern prairie south of Lake Winnipeg (Fig. 5) and thus “Southern Manitoba” is used as the class name. Distinguishing characteristics associated with this class included soil zone PC1 (predominantly black soils) and cropland practice PC1 (predominantly conventional till) (Table 4). Southern Manitoba had a high incidence of glaciolacustrine and alluvial deposits, as indicated by moderately negative and positive relationships with surficial geology PC1 and PC2, respectively, and the class also had low mean elevation. Topography tended to be level, with mild slopes and strong association with land surface form PC3 (Table 4). Notably, these watersheds exhibited both high annual precipitation and PET compared to other classes, and this class was the only one to have no mean moisture deficit (i.e., precipitation – PET > 0) (Fig. 6). Southern Manitoba watersheds also exhibited smaller fractions of non-effective areas and grasslands than other classes (Fig. 7).

Prairie Potholes (C2 and C3)

The Prairie Pothole group, consisting of Class 2 (C2; n = 879), or Pothole Till, and Class 3 (C3; n = 681), Pothole Glaciolacustrine, represents the largest class of watersheds spatially, spanning the northern part of the Alberta prairie to the southeastern part of Saskatchewan (Fig. 5). Mean annual precipitation was relatively high for the study area, contributing to a slightly negative moisture deficit (Fig. 6). These watersheds contained large fractions of non-effective area (~75%) (Fig. 7a), and they exhibited positive scores on land cover PC1 (Table 4) indicating high cropland cover (~70%), whereas unmanaged grassland cover was typically very low (<20%) (Fig. 7b-c). On average, Pothole watersheds had high wetland densities (wetlands km^{-2}), with C2 exhibiting the greatest density of all classes (Fig. 8a).

Surficial geology differentiated classes C2 and C3. Overall, glacial till and hummocky landforms dominated the pothole region; however, C2 was more associated with these characteristics, scoring greater mean values on PC1 of local surface form and surficial geology. In contrast, glaciolacustrine deposits were more common in C3, and soils had a higher incidence of clay and silt, where C2 watersheds were sandier (Table 4). Although both classes contain many wetlands, C2 watersheds had the smallest values of W_L, indicating lower areal water extent was contained in the largest wetland (Fig. 8b).

Major River Valleys (C4)

Class 4 (C4; n = 536) watersheds were associated with river valleys, and as such, extend across the prairie region (Fig. 5) and generally coincide with major rivers (e.g., North and South Saskatchewan, Qu'Appelle) and large lakes. These watersheds had the greatest value of the fraction of water area in the largest depression (W_L) (Fig. 8b), as well as high slope CV, wetland fraction, and fractions of black soil (i.e., higher soil zone PC1 scores) (Table 4). These watersheds were also associated with soil texture PC1 and surficial geology PC2, suggestive of higher incidence of sandy riverine deposits (e.g., alluvial and glaciofluvial deposits). The Major River Valleys class tended to have large "wetland" area, which is interpreted as the area of water of these rivers.

Taken together, these watersheds were related to parameters typical of fluvial environments, including glaciofluvial or alluvial deposits, and sandier soils. Large values of mean and CV of slope were also typical of river valley watersheds. About half the basin area tends to be non-effective in these watersheds, compared to the much greater fractions in the pothole regions (Fig. 7a) that surround many of the Major River Valleys watersheds. Being river valleys, C4 watersheds were generally narrow and small in area. Higher DSF (i.e., narrower watersheds) and smaller areas were generally associated with lower Q2 values (Table 2). Thus, although these watersheds have a high likelihood of contributing to streamflow of major rivers, the watershed Q2 contributions were predicted to be small (Table 4).

Grasslands (C5, C6, and C7)

The southwestern Canadian Prairie, which includes the majority of southern Alberta and western Saskatchewan between the South Saskatchewan River and the Cypress Hills, was occupied by classes C5, C6, and C7. These watersheds tended to have large factions of unmanaged grasslands (negative land cover PC1) and mean elevation (Table 4). Compared to the rest of the Prairie, this sub-region tended to be arid, with a strong moisture deficit (Fig. 6). As a result, these classes exhibited relatively low wetland density (Fig. 8a).

Classes 5 (C5; n = 635), Interior Grasslands, and 6 (C6; n = 702), High-Elevation Grasslands, were characteristic of the grasslands in southeastern Alberta. These watersheds had the greatest values of mean fractional grassland area, with cropland and grassland fractions being comparable (35–40%) (Fig. 7). Distinguishing features of Interior Grasslands were greater values of the fraction of area below the basin outlet, A_{BO}, and a notably large non-effective area fraction (Fig. 7a). High scores on land cover PC2 and PC3 indicate large fractions of fallow and pasture. These watersheds also scored higher on soil zone PC2, suggesting more common occurrences of brown soils. Small magnitudes of mean slope and stream densities were observed, suggesting that the wetlands within the Interior Grasslands are relatively disconnected from the drainage network. This characteristic might explain why these watersheds have relatively large wetlands (Fig. 8c). In contrast, High Elevation Grasslands were characterized by greater mean elevation and slope values, and smaller non-effective fractions (Table 4; Fig. 7). These watersheds also had greater stream densities and smaller wetland densities.

Class 7 (C7; n = 377), Sloped Incised, watersheds are characterized by dissected, river-incised landscapes, as indicated by positive associations with local surface form PC3 (Table 4). Like High Elevation Grasslands (C6), Sloped Incised watersheds followed the Bow, Red Deer, as well as the Milk River valleys, suggesting a similar function to those of the Major River Valleys class. Wetland density is smallest in Sloped Incised watersheds, owing to their steepness, resulting in surface water reaching stream networks rather than collecting on the landscape (Fig. 8).

4.3. Predicting wetland size distributions from class parameters

Simulated wetland area distributions by class were compared to observed size distributions from study watersheds to evaluate the concordance of the approximate class-specific distribution to that of the observed distributions of watersheds, collectively. The median wetland density was greatest in C2, followed by C3, C1, and C5 (Fig. 8a). The median wetland densities in C6 and C7 were less than 1 km^{-2}. C4 had the greatest areal fraction of water in the largest wetland (W_L), which was over 40% (Fig. 8b), while C2 had the smallest value at ~10%. For the rest of the classes, this value was between 28% and 34%. The simulated wetland area distributions slightly overestimated those of the observed values, especially at the 25th percentile. However, the patterns of wetland area in the quartiles was generally consistent among all classes (Fig. 8c). The area of the smallest 25% of the wetlands appears quite consistent across the classes, with more variation occurring at higher percentiles. The largest difference among classes in wetland size was in the 75th percentile, with the greatest range being in C5 and the smallest in C1.

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4.4. Resampling and re-classifying procedure

The HCPC and watershed classification was repeated with ten random subsets of 3757 watersheds. The majority of watershed were removed from at least one iteration, with only 50 watersheds being removed a total of 4-6 times (Fig. S3). This resulted in ten unique watershed subsets to test clustering and agreement to the seven classes, outlined above.

Percent membership agreement of a watershed varied by class, with the majority of classes exhibiting high agreement even after resampling. Classes exhibiting high membership agreement were Pothole Till (C2), Interior Grasslands (C5), High Elevation Grasslands (C6), and Sloped Incised (C7), with a large proportion having more than 90% agreement with the seven classes from the complete classification (Fig. 9; Table S4). Although a large mean agreement was observed overall, a few watershed classes exhibited low agreement and inconsistent classification. Southern Manitoba (C1) exhibited a bimodal distribution, where most were generally classed as C1 over 75% of the time and a second set only ~60% agreement (Fig. 9). This was due to a new class appearing (Fig. 10). Hereafter, this class is referred to as "Eastern Manitoba". Briefly, Eastern Manitoba was association with large fraction of conventional tillage practice (i.e., positive association with land practice PC1 and land practice PC2) and large fractional effective areas (data not shown). The Major River Valleys class was the only one that did not include a watershed that achieved 100% agreement across the ten iterations; this class exhibited a peak of total agreement at approximately 60% (Fig. 9). Where Major River Valleys watersheds were classified inconsistently, the most common alternative classification were Pothole Glaciolacustrine (C3) or secondarily High Elevation Grasslands (C6) (Fig. 10). The loss of Major River Valleys occurred for iterations when the Eastern Manitoba class (C8) became apparent.

5. DISCUSSION

5.1. Classifying Prairie watersheds

5.1.1. Hydrological approaches

Our classification procedure grouped watersheds of approximately 100 km^2 into seven classes. Few studies anywhere have classified watersheds at this granularity, and our investigation givesspecifically within the Canadian Prairie with particular attention to these characteristics that control influence hydrological and ecological behaviour. Many previous studies in the region spanned larger areas, and this often results in the Prairie being identified as a homogenous region due to relatively low streamflow and atypical geology and surface topography (MacCulloch and Whitfield, 2012; Mwale et al., 2011). Our results are novel in that they characterize in greater detail, and at small watershed scales, the potential for different hydrological behaviour of watersheds within the region. The only similar example that was found in the literature was by Durrant and Blackwell (1959), whose findings parallel those of this study, but at a larger watershed scale. Durrant and Blackwell (1959) described broad regions of Saskatchewan and Manitoba based on mean annual flood, distinguishing five sub-regions including southwestern Saskatchewan, north and central Saskatchewan, and southern Manitoba near the Red River and Assiniboine River confluence. In the current study, surficial geology and land surface form strongly influenced how grasslands were separated into three classes, which reinforces the role of local topography on hydrological response, as seen elsewhere (Mwale et al., 2011). Likewise, surficial geology was particularly important for distinguishing the Pothole (Till and Glaciolacustrine) classes. Similarities to the work of Durrant and Blackwell (1959) based on streamflow in larger basins suggest that our approach, with consideration of factors important to watershed behaviour, can yield classification with relevance to hydrologic function, despite the use of few hydrologic indices in our analysis (Fig. 5). This approach holds potential for use in other regions of the world that are dry, ungauged, or feature low effective areas, and thus cannot rely on streamflow characteristics as a primary means of classification according to functional behaviour.

~~The~~ Our classification grouped Prairie watersheds using geological, biophysical, and hydroclimatic attributes. In their review of classification approaches, Sivakumar et al. (2013) indicate that solely using ~~physiographic~~geographic data is advantageous when there are limited hydrological data; however, the relationship between physical attributes and hydrologic behaviour is not necessarily definitive in all regions. For these reasons, it was important to include traits indicative of structural hydrological connectivity, such as Q2 estimates and wetland parameters. It is important to note that while Q2 emerged as a defining feature for several of the classes, it was ~~always~~ consistently one of many variables important for characterization of that class (Table 4), suggesting that while it provides value added, it does not stand out as a major driving factor in the classification. In particular, the immature drainage network and relatively high depressional water storage capacity make prairie hydrology relatively distinct (Jones et al., 2014; Shook et al., 2013, 2015). Notably, three classes (i.e., Pothole Till, Pothole Glaciolacustrine, and Interior Grasslands) occur almost exclusively within regions that tend not to contribute to major river systems, and collectively encompass the majority of the study area (Table 4; Fig. 5). It is therefore expected that hydrological response will be very different between classes that exhibit higher hydrological connectivity (i.e., potentially lower wetland to stream densities and non-effective area fractions), such as the Major River Valleys or Sloped Incised watersheds, than those that do not, such as Pothole classes.

5.1.2. Ecoregions and human impacts

Ecoregions are commonly used to characterize landscapes according to geographical or ecological similarity (~~Masaki and Rosenberry, 2002; ~~Omernik and Griffith, 2014). Similar to our approach, ecoregion classifications are often hierarchical in nature, allowing for differing levels of detail, spatial extent, and thus defining characteristics depending on the scale of interest (Loveland and Merchant, 2004). Ecoregion classifications used in the United States (Omernik and Griffith, 2014) and Canada (Ecological Stratification Group, 1995) employ a “top-down” approach, where broad categories are partitioned into smaller, more specialized units. In contrast, our approach provides a bottom-up, agglomerative approach where similar watersheds are merged. Assumptions are inherent in either approach; however, the latter was applicable to the current study to allow for grouping of watersheds given similarities in physio~~co-geo~~graphic characteristics. This approach does not limit class membership~~ification~~ to the geographic extent of a higher level class, allowing for ~~class ~~membership to potentially span ~~a large ~~the geographic extent of the Canadian Prairie domain (Fig. 5).

Despite the differing methods for distinguishing similarities (or differences), arrangements of watershed classes in some cases exhibited similar ranges to ecoregion boundaries. The boundaries of Lake Manitoba Plain and Mixed Grassland ecoregions (Ecological Working Group, 1995) correspond roughly to those of the broader Southern Manitoba (C1) and Grasslands (C5, C6, and C7) classes, respectively (Fig. S4). Mwale et al. (2011) also found that annual hydrological regimes based on data from 200 stations and physical attributes in Alberta linked closely with provincial ecoregions. Our emphasis on inclusion of hydrologically relevant characteristics, such as wetland traits and effective areas that are likely important contributors to function, has proven useful for further distinguishing among the Grassland classes as well as the Pothole classes (C2 and C3) (Fig. 5; Fig. S4). Due to the fundamental differences in effective areas and in wetland versus river dominated systems (Table

4; Fig. 8~~; Fig. 8~~), we expect different hydrological behaviour between these classes. This is an advantage of the HCPC classification approach in that it allows for identifying the potential similarity at relatively fine spatial scales, and does not require similar watersheds to be physically adjacent to one another. This confers the opportunity to further investigate these systems, such as ~~(e.g.,~~ through hydrological modelling and contrasting resulting responses under ~~of~~climate and land-use scenarios~~)~~.

The highly managed prairie landscape reinforces the importance of considering anthropogenic alteration in hydrological understanding. Crop rotation and the ways in which fields are managed for winter affect the accumulation and redistribution of snow (Fang et al., 2010; Harder et al., 2018; Van der Kamp et al., 2003). Spring snowmelt and consequent runoff are imperative to summer surface water availability (Dumanski et al., 2015; Shook et al., 2015), and depression-focused recharge of snowmelt into groundwater facilitates storage and mitigates flood impacts (Hayashi et al., 2016). Thus, classifying procedures in the Prairie must consider the human influence on the water cycle.

An example of the complexities introduced by human land management activities can be shown by the C1 (Southern Manitoba) watersheds, where the land practice variable was a strong class descriptor. Agricultural activity is high everywhere in the Prairie; however, only C1 was associated with low zero-till practices, instead favouring conventional tillage (Table 4). Manitoba has seen less coherent adoption of zero-till practices since the early 1990s in compared to Alberta and Saskatchewan, with conventional or other conservation till practices remaining common in Manitoba (reviewed in Awada et al., 2014). Sustained use of conventional tillage practice within this region may increase the risk of soil erosion, which can negatively affect downstream water bodies (Cade-Menun et al., 2013). This practice, combined with landscape modifications, such as artificial drainage networks, serve to facilitate removal of water and may contribute to concurrent nutrient export from agricultural lands (Weber et al., 2017).

These management practices can be viewed as a trade-off, where high numbers of wetlands and level topography can pose flood risk during wet periods as wetlands fill and merge (Leibowitz et al., 2016), inundating tracts of adjacent land. Conversely, where landscape modification to enhance water export occurs, local, field-scale flood risk may be reduced, while heightening the risk of downstream flooding. Land-use and land management are important factors in understanding the connectivity and chemical transport in prairie landscapes (Leibowitz et al., 2018). In southern Manitoba, where artificial drainage has been used to increase the area of arable land, beneficial management practices in the form of agricultural reservoirs have been implemented as a means of reducing nutrient export and improving downstream water quality while also mitigating the risk of downstream flooding (Gooding and Baluch, 2017). These factors illustrate the complexities when classifying and understanding hydrological response of watershed embedded in highly managed landscapes, and underscore that necessity of considering the human influence on the water cycle in such approaches.

5.2. HCPC as a clustering and classification framework

5.2.1. Using the HCPC approach and limitations

The HCPC method provides a procedure for integrating multiple physio~~-geo~~graphic attributes and describes resulting clusters by sets of significant variables (Husson et al., 2009). As discussed above, an advantage of the method is that it groups individual watersheds based on similarities.~~, and~~ T~~t~~herefore, it lends itself well as a~~to setting a~~ foundation for investigating hydrological behaviour ~~to be applied to~~through modelling efforts. In the case of the current study, modelling efforts can be applied at a 100 km^2 scale to evaluate responses to environmental changes. An additional advantage is that that one may select variables or sets of variables of interest to inform the clustering of watersheds, such as those based only on topographic parameters or those dictating local hydrology. For example, climate variables may be excluded if the goal of the classification is ~~informing application~~parameterizing ~~of~~ a hydrological model, as these variables could instead be ~~part of~~described by ~~model parameterization~~local climate forcing. The relative ease with which different sets of variables can be added to or excluded from the analysis to consider different permutations of the classification is a real strength of the approach. Although this may result in differing cluster results, assessment of how these classes change with addition or removal of certain datasets can identify the variables that control class definition as well as elucidate spatial patterning of classes.

There are a few considerations when using this method. First, the linear restrictions of this method are challenging when working with environmental data, which often do not conform to assumptions of normality. Non-linear PCA methods and self-organizing maps have been applied successfully to classify watersheds in Ontario and to regionalize streamflow metrics (Razavi and Coulibaly, 2013, 2017). Although these methods might be logical next steps for the

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current study, we chose to focus on conventional PCA due to its smaller computational cost when classifying the large number of watersheds in our study.

Second, the current analysis weighs all variables equally. This can bias the analysis towards attributes that exhibit greater variability, as these can overshadow other more constrained variables. For example, the location of the largest pond relative to the watershed outlet (coded as L_W/L_O) is important to controlling local prairie hydrology and hydrological gate-keeping potential (i.e., the likelihood of releasing surface water to the next order watershed) (Shook et al., 2013, 2015) and water quality (Hansen et al., 2018). Despite its hydrological importance, this variable had little influence on the clustering procedure overall, and was only a minor descriptor in certain classes, such as C5 and C6 (Table 4).

~~The classes resulting from the HCPC are ultimately dependent on the types of data included. The availability of data and its geographic coverage determined the environmental parameters included in our analyses. Ideally, a more detailed estimate of runoff for each watershed, as well as a soil moisture dataset would have been include. A comprehensive wetland inventory or an index of wetland drainage activity that is comparable across the three Provinces does not currently exist. These would be valuable additions to future efforts to classify Prairie watersheds given the important role of land modification on watershed functions.~~

The original set of watersheds in the clustering analysis can affect the final classification; however, there was a high degree of agreement between classified subsets of the original dataset, and the classification generated using the complete set of watersheds (n = 4175) (Fig. 9). Overall, watersheds designated as part of the Pothole and Grassland classes were classified consistently, with most exhibiting over 90% agreement. Major River Valleys exhibited the weakest agreement (Fig. 9), due to the appearance of a unique (new) class consistent with the Lake Manitoba Plain ecoregion (Fig. S4) for some of the subsets. In these cases, those watersheds previously classified as Major River Valleys were re-distributed to mainly High Elevation Grasslands or Pothole classes depending on ~~the~~ dominate watershed features (Fig. 10). Although we do not include a detailed account of the new Eastern Manitoba class that emerged during this exercise, defining characteristics included a high fraction of effective area (i.e., the most eastern portion of the Prairie in Fig. 1), low relief, and lower use of zero-till agriculture (as reviewed in Awada et al., 2014). Since this new class would not be expected to translate to notable differences in management outcomes. Moreover,~~.~~ previous reviews on the usefulness of ecoregion classifications agree that strict geographic boundaries are unlikely, and are instead more likely "fuzzy" (Loveland and Merchant, 2004; Omernik and Griffiths, 2014).

~~c~~Class membership in our approach is also determinate. In reality, there can be large variability in attributes within a class (e.g., Fig. 7), and membership is determined by the collective similarity of watershed attributes. Previous studies have used fuzzy c-means and Bayesian approaches that can assign a likelihood of membership to classes (Jones et al., 2014; Rao and Srinivas, 2006; Sawicz et al., 2011). An advantage to this approach is that it allows for fuzzy boundaries between classes where a gradient of features likely exists (Loveland and Merchant, 2004). Our re-classifying analysis supports the proposition that boundaries among classified regions are fuzzy and some watershed might flicker among class memberships (Fig. 10). Such approaches~~, which~~ are also un-supervised and~~, are~~ probabilistic in nature and will eliminate the subjectivity due to the researcher pre-defining the number of classes. ~~Our f~~Future work thus should consider these fuzzy boundaries and potential for watersheds to exhibit partial membership to multiple classes.~~will include applying a fuzzy cluster Bayesian framework to assess the current classification framework.~~

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5.2.2. Data quality and availability

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The classes resulting from the HCPC are also ultimately dependent on the types of data included. The availability of data and its geographic coverage determined the environmental parameters included in our analyses. Ideally, a more detailed estimate of runoff for each watershed would be a valuable contribution. In the current study, we used the CCA and eleven reference stations to approximate runoff values for the clustering watersheds. Given the number of watersheds included in the analyses, the diversity of physical characteristics and potential hydrological behaviour is likely not completely represented in the small sample size of available hydrometric stations, and is a limitation of our approach. Soil moisture would be important to consider in future studies given its role in influencing vegetation community composition, PET, and over all water balance (Hayashi et al., 2003; Shook et al., 2015). Where data is available, future work should consider variables related to snow formation and melt, as well the proportion of annual precipitation as snowfall. These variables are likely influential when describing hydrological behaviour of the watersheds and classes in the current study, and other cold regions (Knoben et al., 2018; Shook and Pomeroy, 2012). Furthermore, a comprehensive wetland inventory or an index of wetland drainage activity that is comparable across the three Provinces does not currently exist. These would be valuable additions to future efforts to classify Prairie watersheds given the important role of land modification on watershed functions.

One consideration with the Global Surface Water dataset is that the pixel size (30 m) is quite coarse and will miss numerous smaller wetlands, underestimating the number of wetlands observed. Consequently, it is likely that the analysis omitted some ephemeral wetlands for which persistence is short and size is small. Despite their known important ecological functions (Calhoun et al., 2017; Van Meter and Basu, 2015), their size and transient nature is a challenge to their inclusion in comprehensive datasets spanning large geographic areas. This may inadvertently result in the role of smaller wetlands being under-represented in our analysis, or others that rely on this dataset.

Use of the ξ and β parameters as indices of the wetland area frequency distributions were shown to estimate classes area distributions reasonably well (Fig. 8c). Although for consistency, we restricted our simulated dataset to the spatial resolution of the surface water raster, one could use these parameters to estimate the frequencies of smaller wetlands in watersheds, which would otherwise be missed by satellite measurements, assuming conformity to a Generalized Pareto Distribution (Shook et al., 2013). Our analysis supports this application as simulated wetland areas generally approximated those seen across the observed data (Fig. 8c). Nonetheless, in regions where wetland drainage has been undertaken, it is expected that wetland area distribution has been altered via preferential loss of smaller water bodies (Evenson et al., 2018; Van Meter and Basu, 2015). This is exacerbated by the fact that remote sensed satellite data tends to omit smaller, ephemeral ponds. A more robust characterization of the size and permanence of wetlands in our study watersheds would be expected to improve the current dataset and enhance the clustering and classification analyses.

Finally, class membership is determinate. In reality, there can be large variability in attributes within a class (e.g., Fig. 7), and membership is determined by the collective similarity of watershed attributes. Previous studies have used fuzzy c-means and Bayesian approaches that can assign a likelihood of membership to classes (Jones et al., 2014; Rao and Srinivas, 2006; Sawicz et al., 2011). An advantage to this approach is that it allows for fuzzy boundaries between classes where a gradient of features likely exists (Loveland and Merchant, 2004). Such approaches, which are also un-supervised, are probabilistic in nature and will eliminate the

~~subjectivity due to the researcher pre-defining the number of classes. Our future work will include applying a fuzzy-cluster Bayesian framework to assess the current classification framework.~~

5.3. Management implications

Classification frameworks help to define sub-regions with potentially similar characteristics or behaviours. For example, climatic zones can be delineated, specifically the dry Grassland watersheds in the southwest and the wet Potholes in the northeast and in Manitoba (Fig. 5). In some cases, this may be related to local wetland densities, with large densities observed corresponding with low moisture deficits (Fig. 6b) (Liu and Schwartz, 2012). Climate variation may divide watersheds with seemingly similar ~~physio-~~geography into differing classes, as is the case with Major River Valleys and Sloped Incised watersheds. Both sets of watersheds tended to follow river valleys, but the former exhibit greater precipitation and smaller PET (Table 4). These divisions can be used to give context to regions we might expect to behave similarly, whether hydrologically, or ecologically, based solely on physical attributes, and echoes other methods, such as ecodistricts (Ecological Stratification Working Group, 1995) to classify landscapes. For example, areas that are geologically similar may differ in terrestrial or aquatic community assemblages, which should influence how each area might be managed (Jones et al., 2014; Wagner et al., 2007). If classifications are used to inform management, the resulting decisions for a given location will depend on the strength of the delineation, the scale at which management is applied, relationships among management practices and the attributes used to define that area, and the relationship of those attributes to the response variable of concern (Wagner et al., 2007)~~(Wagener et al., 2007).~~.

This set of analyses was unique among watershed classification exercises in Canada in that it considered a suite of wetland variables. The arrangement of wetlands or landscape depressions and their size distribution define the hydrological behavior of Prairie watersheds (Shook et al., 2015; Shook and Pomeroy, 2011). The storage capacity and subsequent spilling or merging controls wetland connectivity, and thus the quantity of water available to move from one watershed to another (Leibowitz et al., 2016; Shaw et al., 2012; Shook et al., 2015). In turn, a wetland or depression’s hydrological gate-keeping potential, or its likelihood to prevent connectivity to the downstream watershed, is a function of both its storage capacity and landscape position. Large wetlands near an outlet have a great gate-keeping potential, as they block much of the watershed from connecting, and it takes a great deal of water to fill them before permitting flow to the next order watershed (Shook and Pomeroy, 2011). Simulated frequency distributions of wetland areas indicate that the depressional storages of the classes are very different (Fig. 8). It may be that wetland management practices will have different influences between each pothole class, and possibly among all the classes. This has implications for managing salinizing soils (Goldhaber et al., 2014), biodiversity (Balas et al., 2012), and floodings potential (Evenson et al., 2018; Golden et al., 2017).

Wetland drainage and wetland consolidation change hydrological connectivity and therefore the transport of nutrients and their loading into receiving water bodies (Brown et al., 2017; Vanderhoof et al., 2017). More positive values of the moisture deficit (i.e., where P >= PET) were associated with greater wetland densities (Fig. 6b) (Liu and Schwartz, 2012), and these areas were generally associated with greater fractions of cropland, such as Pothole Till, Pothole Glaciolacustrine, and Southern Manitoba watersheds. In these regions wetland drainage is widely practiced, historically or at present, and conflict over available arable land and wetland conservation is high (Breen et al., 2018).

Extensive drainage in combination with agricultural activity is known to increase the risk of agricultural nutrient mobility (Kerr, 2017) from the landscape to receiving water bodies. Increased connectivity also reduces water residence time and thus tends to decrease wetland nutrient retention (Marton et al., 2015). Over time, zero-till practices can promote nutrient stratification in soils, where concentrations (especially phosphorus) accumulate at the surface, which can increase nutrient loading when surface runoff is generated (Cade-Menun et al., 2013). The cropland-wetland interface might also have important implications for pesticide mobility in Pothole Till and northern Pothole Glaciolacustrine watersheds. These areas coincide with extensive use of canola, which has been linked to high application rates of neonicotinoid pesticides which are known to have high persistence in small, temporary wetlands (Main et al., 2014). Watersheds in the Pothole Till class appear to have more hummocky landscapes than the Pothole Glaciolacustrine classification and smaller, more numerous wetlands (Fig. 8). Moreover, the water area fraction occupied by the largest wetland differs between the classes. The landscape biogeochemical functionality of pothole wetlands is known to vary considerably according to pothole character (Evenson et al., 2018; Van Meter and Basu, 2015). As such, our classification may highlight contrasting biogeochemical functioning, including nutrient retention, between these classes. Thus, although water quality risks are common within the region, the classes may respond very differently to environmental and land management stresses.

6. CONCLUSION

This study provides an overview of a classification framework that can be applied in regions with limited understanding of or data describing streamflow. The HCPC procedure offers a flexible analysis to elucidate the spatial arrangement of watershed classes given a large number of units to classify and a diverse set of attributes to inform the classification. In contrast to classifications based solely on hydrological function, using physio~~-geo~~graphic data allows for classifying small basins, which are unlikely to be gauged, and confers advantages over alternate procedures that rely heavily on observations of hydrological parameters, namely statistics describing streamflow.

Use of the classification approach for small Canadian Prairie watersheds identified regions of similar climatic and ~~physio-~~geographic features and, potentially, of hydrological response (Fig. 5). This yielded watershed classes that consider not only drainage patterns, but also land cover,~~ and~~ land-use, and the underlying geology. In the Prairie region, wetland variables incorporate the hydrologic gate-keeping potential of wetlands as well as parameters indicative of wetland size distributions. With the classification based on a large and diverse set of attributes, a diversity of behaviours is captured. This represents a major step forward for classification of Prairie watersheds that have to-date offered only a much more homogenized depiction of watershed function in the region. The watershed classification framework presented promises to be useful in other dry or semi-arid regions, and those that are poorly gauged. Given the inclusive nature of the classification approach, which incorporates landscape controls on hydrology as well as those influencing biogeochemistry and ecology, it also provides a foundation to evaluate the efficacy of land and watershed management practices in the context of a changing climate.

Author contributions

JDW, CJW, and CS conceived the study, and JDW collected data and performed analyses. KRS wrote code to analyze basin and wetland data. JDW wrote the manuscript with input from all co-authors.

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~~Tables~~ TABLES AND FIGURES~~and Figures~~

Table 1 – Pre-processing of compositional data PCA results. Shown are the respective subsets, the number of initial fractional area variables before dimensional reduction, the number of principal components retains to reach over 80% of subset variation (except for tillage practice), and the proportion of variation explained by each component.

Variable subset	Number of initial variables	Number of principal components	Total variation explained by component
Surficial geology	6	2	1: 72.8% 2: 14.4%
Particle size class	3	2	1: 74.8% 2: 15.6%
Soil zone	5	2	1: 54.6% 2: 42.7%
Local surface form	5	3	1: 54.5% 2: 24.2% 3: 11.9%
Land cover	5	3	1: 36.8% 2: 25.2% 3: 20.6%
Tillage practice	3	2	1: 90.9% 2: 8.5%

Table 2 – Canonical correlation coefficients for watershed attribute and hydrological variables of hydrological research stations from the canonical correlation analysis. Those variables used in multiple regression equations are denoted with a ‘*’.

	Correlation	
Watershed attributes	**V1**	**V2**
Area*	0.36	–0.83
DSF*	–0.26	0.90
Fraction rock*	–0.64	0.61
Fraction natural area*	–0.26	0.71
Stream density	–0.27	0.37
Mean annual precipitation	–0.14	–0.30
Fraction water area	0.53	–0.19
Hydrological variables	**W1**	**W2**
Q2	–0.82	–0.58
Q100	–0.22	–0.98
Canonical λ	0.97	0.77

Table 3 – Correlation of study watershed attributes to principal components (PC). The values for the six PCs used in the cluster analysis are shown.

Variable	Abbreviation	PC1	PC2	PC3	PC4	PC5	PC6
Mean elevation	elevation	0.81	0.34	–0.14	0.09	–0.16	–0.17
Mean slope	slope.mean	0.61	–0.23	0.06	0.37	–0.10	0.11
Slope CV	slope.CV	0.30	–0.38	0.22	0.14	–0.41	–0.09
Total precipitation	precip	–0.85	–0.12	0.13	0.16	–0.07	0.30
Potential evapotranspiration	PET	0.31	–0.33	–0.33	–0.06	0.47	0.13
Non-effective area	NE.area	0.02	0.70	0.31	0.10	0.01	–0.15
Areal fraction below outlet (A_{BO})	A_BO	0.14	0.25	0.27	–0.17	0.42	–0.01
Stream density	stream.density	0.08	–0.42	–0.39	0.03	–0.41	0.08
Wetland density	wetland.density	–0.63	0.46	0.11	–0.04	0.12	0.24
Wetland fraction	wetland.frac	–0.30	0.19	0.66	–0.36	0.02	0.11
Water area in largest wetland to total in watershed (W_L)	W_L	0.31	–0.44	0.51	–0.32	–0.06	–0.12
Location of largest wetland to outlet (L_W/L_O)	L_W/L_O	–0.01	–0.06	–0.22	0.09	–0.07	–0.07
Beta (β)	beta	0.17	0.49	–0.02	0.01	0.09	0.05
Xi (ξ)	xi	0.21	–0.23	0.57	–0.31	–0.10	–0.17
Runoff (Q2)	Q2	–0.13	0.35	–0.47	0.00	–0.33	0.10
Soil texture PC1	Text.PC1	–0.07	–0.04	0.28	0.55	0.19	–0.32
Soil texture PC2	Text.PC2	0.02	–0.32	0.43	0.03	–0.31	0.54
Soil zone PC1	Soil.PC1	–0.65	–0.29	–0.07	0.19	–0.10	–0.24
Soil zone PC2	Soil.PC2	0.27	–0.12	–0.06	–0.11	0.40	0.25
Land cover PC1	LC.PC1	–0.44	0.38	–0.21	–0.26	–0.43	0.12
Land cover PC2	LC.PC2	0.42	0.22	–0.17	–0.53	0.15	0.03
Land cover PC3	LC.PC3	0.21	0.30	0.15	0.25	0.11	0.46
Surficial geology PC1	SF.PC1	0.06	0.21	–0.19	0.50	0.17	–0.09
Surficial geology PC2	SF.PC2	0.06	–0.38	0.24	0.47	0.11	–0.03
Surface form PC1	LL.PC1	–0.16	0.20	0.17	0.47	0.26	0.26
Surface form PC2	LL.PC2	–0.20	0.44	0.12	–0.03	0.04	–0.55
Surface form PC3	LL.PC3	0.41	0.38	0.20	0.21	–0.27	0.27
Land practice PC1	LP.PC1	–0.54	–0.58	–0.13	–0.10	0.32	–0.09
Land practice PC2	LP.PC2	0.14	–0.16	–0.24	–0.22	0.29	0.30
Supplementary variables							
Latitude	Lat	–0.15	0.24	0.26	–0.01	–0.33	–0.41
Longitude	Long	–0.73	–0.24	0.06	0.10	0.16	0.39
Area	Area	–0.05	0.27	–0.44	0.09	–0.15	–0.03
DSF	DSF	–0.02	–0.25	0.42	–0.05	0.12	0.01

Table 4 – Classes and distinguishing variables of prairie watersheds. The v-test statistics, based on Ward's criterion, are shown. Variables with v-test values greater or less than 10 and –10, respectively, are bolded to emphasize defining features of each class. All variables are significant to $p < 0.001$. *Classes: Southern Manitoba (1), Pothole Till (2), Pothole Glaciolacustrine (3), Major River Valleys (4), Interior Grasslands (5), High Elevation Grasslands (6), Sloped Incised (7).*

Class 1 (n=365)		Class 2 (n=879)		Class 3 (n=681)		Class 4 (n=536)	
Variable	v-test	Variable	v-test	Variable	v-test	Variable	v-test
LP.PC1	**48.11**	**wetland.density**	**28.23**	**LC.PC1**	**22.60**	**SF.PC2**	**19.83**
precip	**30.33**	**LL.PC1**	**24.81**	**wetland.frac**	**12.74**	**slope.CV**	**19.35**
Soil.PC1	**23.60**	**precip**	**22.74**	**Q2**	**12.63**	**xi**	**16.05**
LP.PC2	**14.74**	**SF.PC1**	**21.74**	**NE.area**	**11.12**	**W_L**	**15.39**
PET	**13.10**	**LC.PC1**	**17.19**	LL.PC2	9.45	**Text.PC2**	**15.07**
wetland.density	7.39	**LL.PC2**	**16.42**	wetland.density	8.05	**Text.PC1**	**14.40**
DSF	6.81	**Q2**	**15.77**	LC.PC2	6.70	**Soil.PC1**	**14.01**
SF.PC2	6.53	**Soil.PC1**	**15.76**	LL.PC3	6.53	**DSF**	**11.76**
stream.density	4.61	**NE.area**	**15.72**	xi	5.89	**precip**	**10.97**
LC.PC1	–3.37	**area**	**13.15**	W_L	4.58	**wetland.frac**	**10.92**
A_BO	–4.22	**Text.PC1**	**12.00**	precip	3.47	slope.mean	7.29
area	–5.46	LC.PC3	6.76	A_BO	–3.79	LP.PC1	3.52
slope.CV	–6.49	beta	5.31	slope.CV	–4.97	A_BO	–3.83
Q2	–8.47	L_W/L_O	4.20	L_W/L_O	–5.17	wetland.density	–4.41
SF.PC1	–8.90	LL.PC3	3.93	LP.PC2	–7.11	SF.PC1	–4.56
LC.PC2	–9.21	SF.PC2	–3.97	LC.PC3	–9.71	LC.PC1	–5.13
LL.PC2	**–14.18**	LP.PC1	–4.87	**LP.PC1**	**–12.38**	soil.PC2	–6.93
slope.mean	**–16.17**	stream.density	–5.92	**Soil.PC2**	**–13.01**	beta	–7.60
beta	**–16.88**	elevation	–7.15	**Text.PC1**	**–14.58**	elevation	–8.03
LC.PC3	**–18.13**	A_BO	–7.86	**slope.mean**	**–15.92**	**area**	**–11.04**
NE.area	**–28.97**	Text.PC2	–9.15	**SF.PC2**	**–17.03**	**LP.PC2**	**–11.44**
LL.PC3	**–36.59**	DSF	–9.93	**LL.PC1**	**–17.83**	**Q2**	**–13.27**
elevation	**–47.42**	**LP.PC2**	**–10.88**	**SF.PC1**	**–18.83**	**PET**	**–13.98**
		Soil.PC2	**–12.00**	**PET**	**–23.29**	**LC.PC2**	**–20.86**
		PET	**–13.15**				
		slope.mean	**–13.50**				
		slope.CV	**–16.26**				
		LC.PC2	**–16.29**				
		xi	**–21.49**				
		W_L	**–32.96**				

Table 4 – *(cont'd)*

Class 5 (n=635)		Class 6 (n=702)		Class 7 (n=377)	
Variable	v-test	Variable	v-test	Variable	v-test
A_BO	**34.10**	**elevation**	**29.29**	**Text.PC2**	**27.65**
LC.PC2	**21.53**	**PET**	**20.16**	**LL.PC3**	**25.69**
Soil.PC2	**20.81**	**slope.CV**	**17.67**	**slope.mean**	**22.32**
LC.PC3	**17.44**	**slope.mean**	**16.12**	**LC.PC3**	**14.84**
NE.area	**16.22**	**stream.density**	**14.55**	**stream.density**	**13.82**
beta	**15.96**	**LC.PC2**	**14.09**	**Soil.PC2**	**13.09**
elevation	**13.31**	W_L	9.47	**elevation**	**12.42**
PET	**11.47**	L_W/L_O	6.80	**PET**	**11.47**
LL.PC2	8.11	LP.PC2	5.73	SF.PC2	6.80
LP.PC2	7.67	area	3.72	LP.PC2	6.39
LL.PC3	7.31	LL.PC2	3.62	slope.CV	5.87
wetland.frac	5.77	LP.PC1	–3.60	W_L	4.63
LL.PC1	5.50	Q2	–3.94	precip	–4.75
SF.PC2	–4.74	DSF	–4.91	A_BO	–5.65
area	–4.86	A_BO	–9.47	LC.PC1	–7.62
L_W/L_O	–7.11	**Soil.PC1**	**–10.17**	Text.PC1	–8.34
Q2	–9.34	**LL.PC3**	**–10.62**	**LP.PC1**	**–11.42**
LP.PC1	–9.96	**LC.PC3**	**–13.17**	**NE.area**	**–13.33**
Text.PC2	**–11.36**	**NE.area**	**–14.11**	**wetland.frac**	**–13.64**
LC.PC1	**–11.38**	**LL.PC1**	**–15.44**	**wetland.density**	**–16.27**
slope.CV	**–12.42**	**Text.PC2**	**–15.78**	**Soil.PC1**	**–16.43**
precip	**–20.86**	**LC.PC1**	**–17.15**	**LL.PC2**	**–39.41**
Soil.PC1	**–23.58**	**wetland.frac**	**–21.48**		
stream.density	**–26.34**	**wetland.density**	**–29.58**		
		precip	**–37.27**		

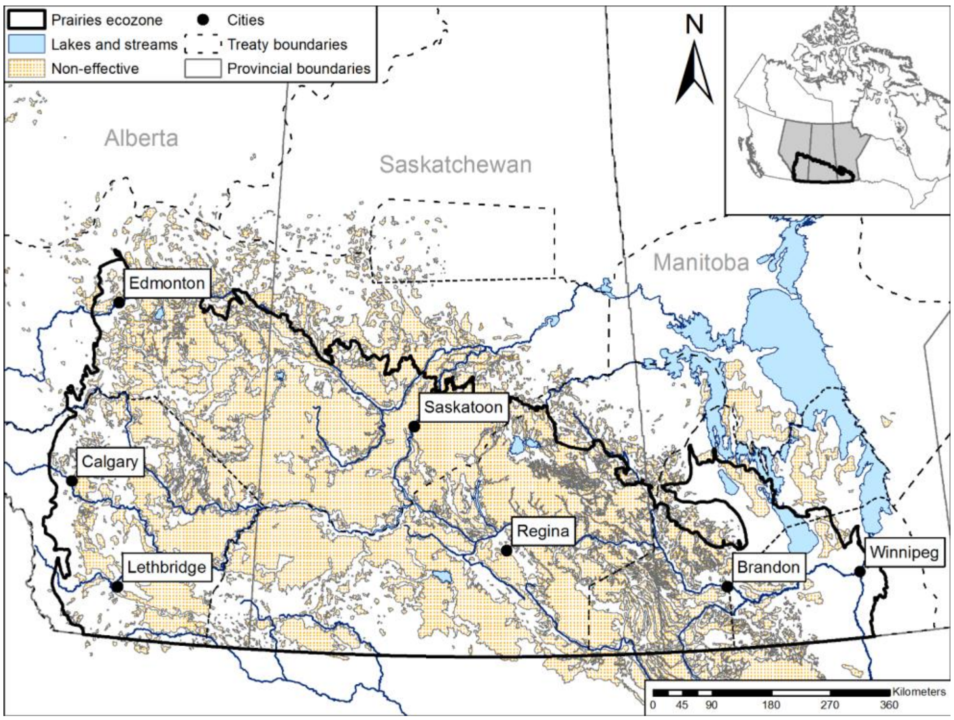


Figure 1 – Map of the study area spanning the Prairies ecozone in western Canada (inset). Large cities in each of the three provinces are shown for reference, while the region characterized as not contributing runoff (2-year) is also shown. Prairie ecozone based on the region classified by the Ecological Stratification Group (1995).

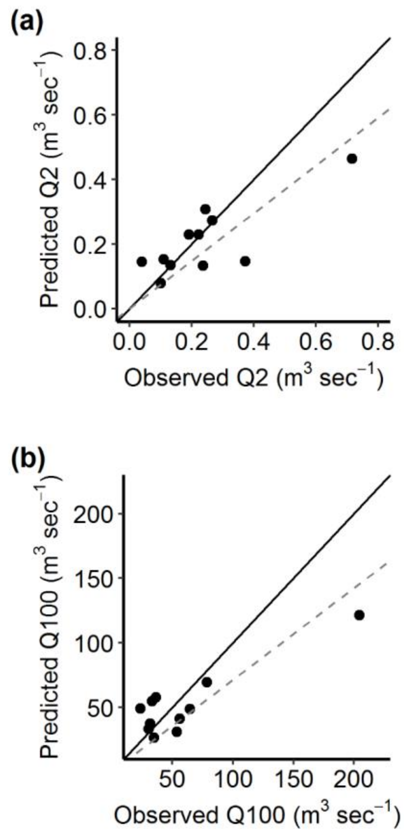


Figure 2 – Observed versus predicted estimates for (a) Q2, and (b) Q100. The dashed grey line depicts the linear regression between observed and predicted flow values, and the black, solid line shows a 1:1 relationship.

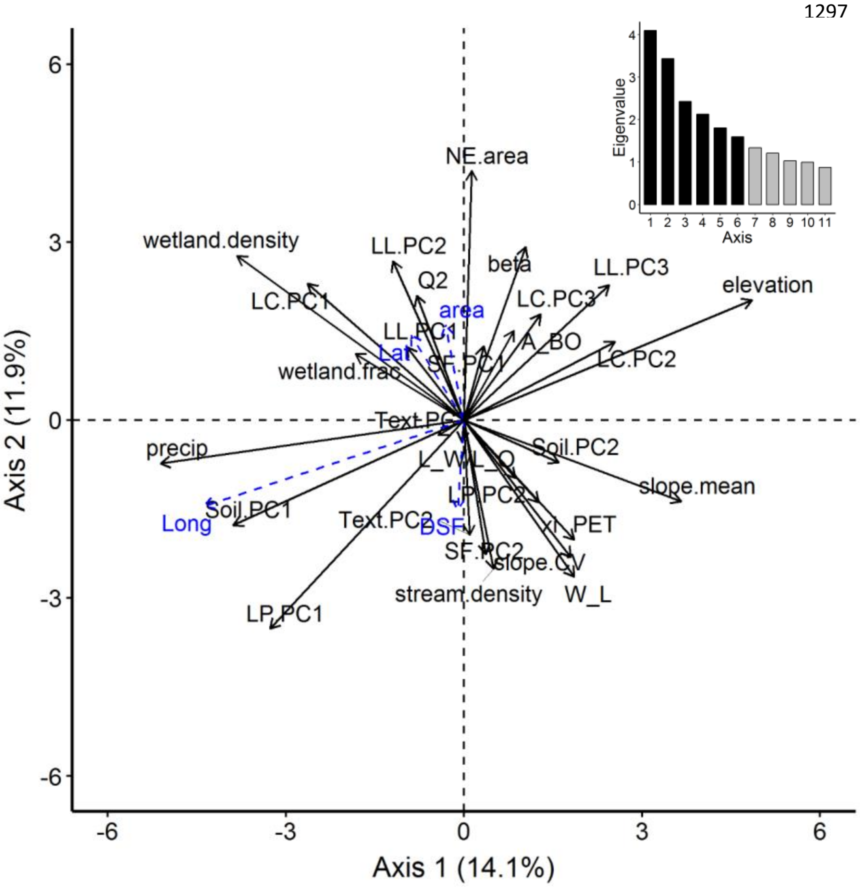


Figure 3 – Principal components analysis for candidate variables for classification. Active and supplementary variables are shown as solid black, and dashed blue arrows, respectively. Eigenvalues for PC axes are provided (inset), with black bars denoting the six PCs used in the hierarchical clustering analysis.

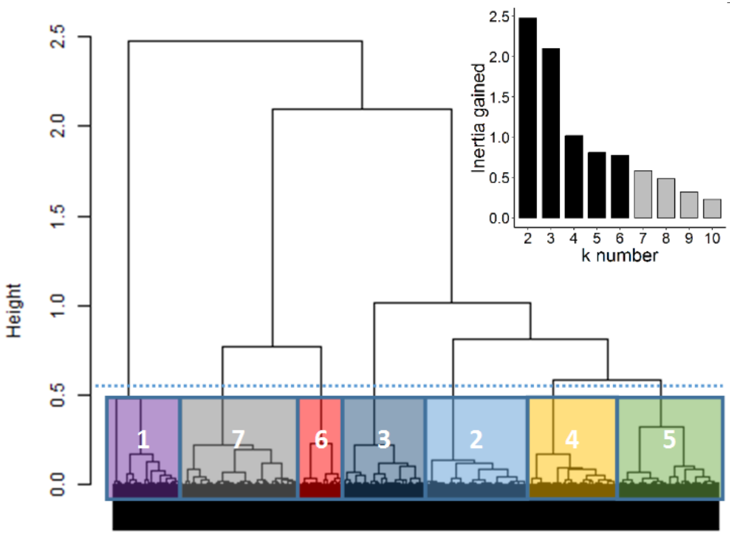


Figure 4 – Dendrogram resulting from the hierarchical cluster analysis of principal components. The blue, dashed line indicates the cut in the tree, resulting in seven clusters. The amount of inertia gained by increasing the number of clusters (*k*) is depicted in the inset panel.

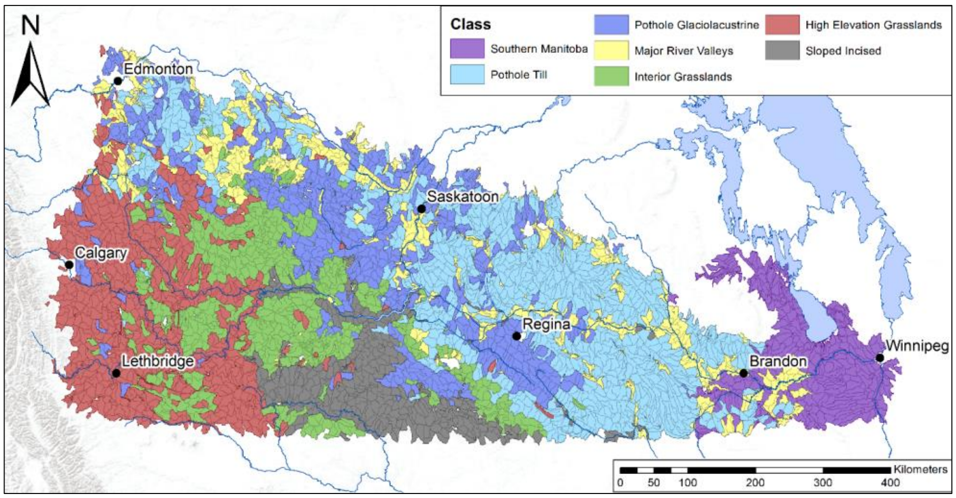
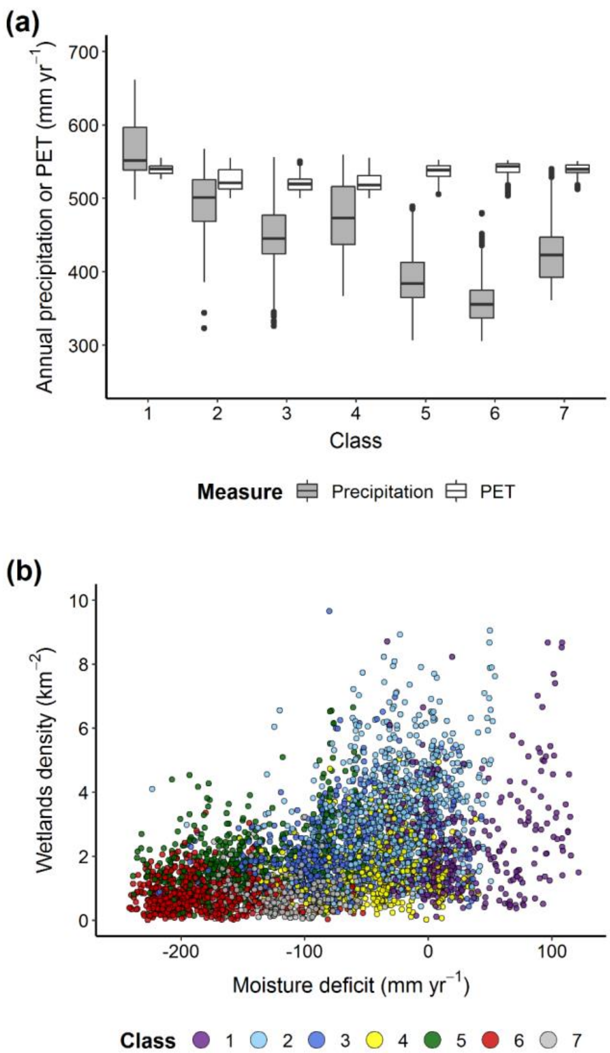


Figure 5 – Classification of Prairie ecozone watersheds. Watershed delineations are from Lehner and Grills (2013), available at www.hydrosheds.org. See text for detailed interpretation of the seven clusters.


(a)
Annual precipitation or PET (mm yr^{-1})
700
600
500
400
300
1
2
3
4
5
6
7
Class
Measure
Precipitation
PET
(b)
Wetlands density (km^{-2})
10
8
6
4
2
0
-200
-100
0
100
Moisture deficit (mm yr^{-1})
Class
1
2
3
4
5
6
7

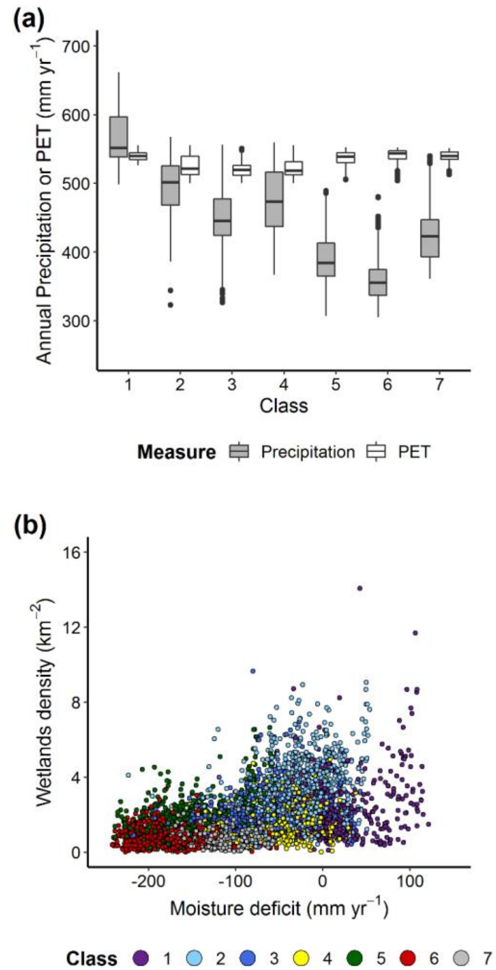


Figure 6 – Climatic variation among watershed classes. (a) Boxplots of total annual precipitation (grey) and potential evapotranspiration (PET) (white) for each watershed cluster. Lower, middle, and upper limits of boxes show the 25th, 50th, and 75th quantiles, respectively. (b) Wetland density to moisture deficit (Precipitation – PET). *Classes: Southern Manitoba (1), Pothole Till (2), Pothole Glaciolacustrine (3), Major River Valleys (4), Interior Grasslands (5), High Elevation Grasslands (6), Sloped Incised (7).*

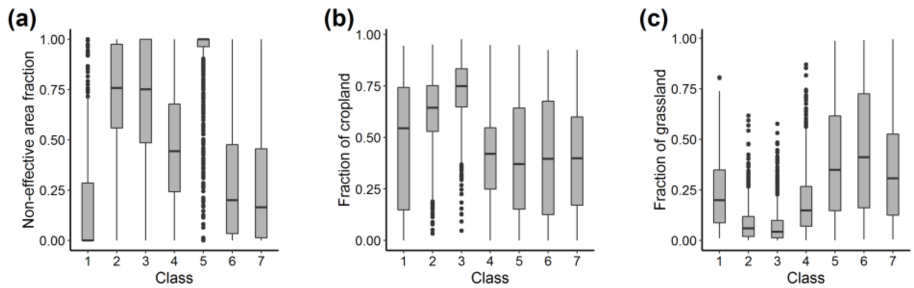
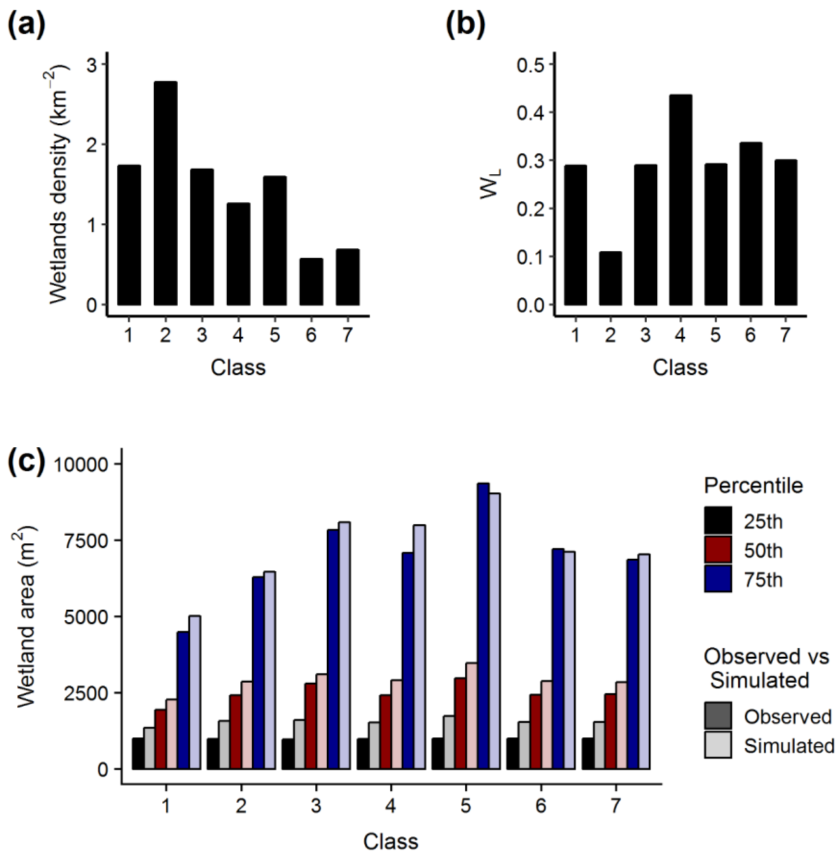


Figure 7 – Boxplots of select variables by watershed class: (a) fraction of non-effective area; (b) fraction of cropland; and (c) fraction of grassland. *Classes: (1) Southern Manitoba, (2) Pothole Till, (3) Pothole Glaciolacustrine, (4) Major River Valleys, (5) Interior Grassland, (6) High Elevation Grasslands, and (7) Sloped Incised.*


(a)
Wetlands density (km^{-2})
3
2
1
0
1 2 3 4 5 6 7
Class
(b)
W_L
0.5
0.4
0.3
0.2
0.1
0.0
1 2 3 4 5 6 7
Class
(c)
Wetland area (m^2)
10000
7500
5000
2500
0
1 2 3 4 5 6 7
Class
Percentile
25th
50th
75th
Observed vs Simulated
Observed
Simulated

1344

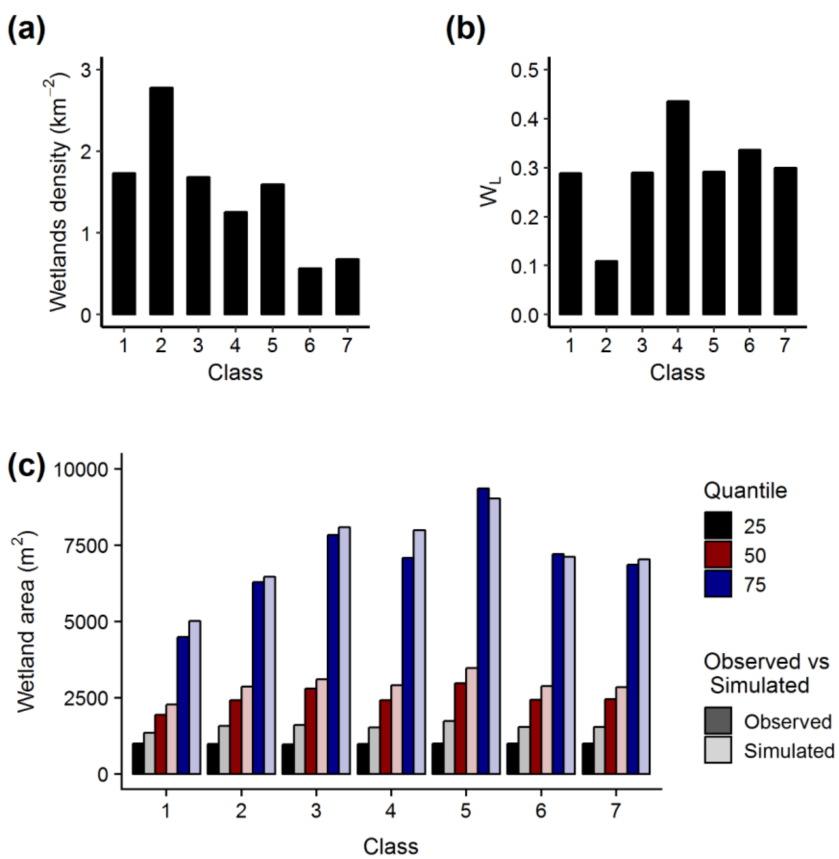


Figure 8 – Wetland variables and simulated size distributions. Median (a) density of wetlands and (b) fraction of total watershed water area in the largest wetland (W_L) are depicted by class. Panel (c) shows observed (dark) and simulated (light) percequantiles of wetland areas. Predicted values are based on a generalized Pareto distribution and using median parameters of β and ζ for each cluster. Simulated data were restricted to the raster pixel resolution of observed data from the Global Surface Water dataset. *Classes: Southern Manitoba (1), Pothole Till (2), Pothole Glaciolacustrine (3), Major River Valleys (4), Interior Grasslands (5), High Elevation Grasslands (6), Sloped Incised (7).*

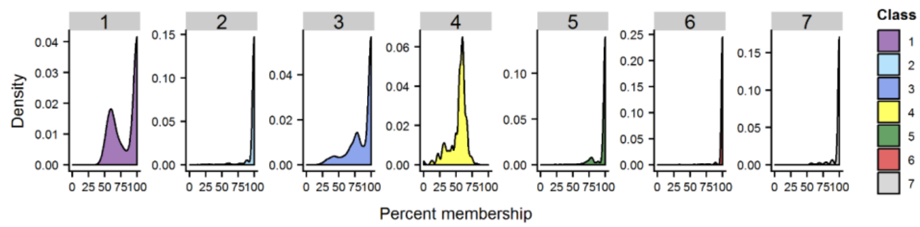


Figure 9 – Density distributions of percent agreement of watersheds to the classification in Fig. 5 by watershed class. *Classes: Southern Manitoba (1), Pothole Till (2), Pothole Glaciolacustrine (3), Major River Valleys (4), Interior Grasslands (5), High Elevation Grasslands (6), Sloped Incised (7).*

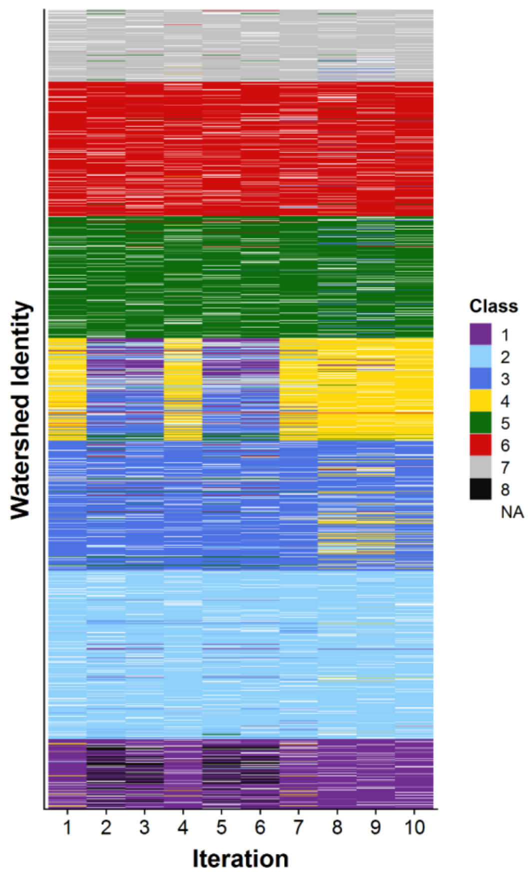


Figure 10 – Agreement of assigned watershed classification from the (original) complete analysis, with class assignments from the iterative approach using re-sampling. Classes are coloured according to that shown in Fig. 5, with those identified under a new class (C8) depicted in black. Watersheds that were removed from the subsets analyzed are in white. *Classes: Southern Manitoba (1), Pothole Till (2), Pothole Glaciolacustrine (3), Major River Valleys (4), Interior Grasslands (5), High Elevation Grasslands (6), Sloped Incised (7).*