

1 **Authors' responses to reviews for hess-2014-444**

2 **Groundwater surface mapping informs sources of**
3 **catchment baseflow**

4
5 **J. F. Costelloe¹, T. J. Peterson¹, K. Halbert², A. W. Western¹ and J. J.**
6 **McDonnell^{3, 4}**

7 [1]{Department of Infrastructure Engineering, University of Melbourne, Australia}

8 [2]{Ecole Centrale de Nantes, Nantes, France}

9 [3]{Global Institute For Water Security, University of Saskatchewan, Saskatoon, Canada}

10 [4] {School of Geosciences, University of Aberdeen, Aberdeen Scotland}

11 Correspondence to: J. F. Costelloe (jcost@unimelb.edu.au)

12
13 We thank Anonymous Referees #1 and #2 and Professor Ian Cartwright for their considered
14 and insightful reviews of our paper. Their helpful comments and suggestions have been
15 addressed below with responses to each individual comment. The page number and line
16 number of changes made in the revised manuscript are provided for each comment.

17
18 **Professor Ian Cartwright**

19 The main comment that I have is regarding the bores used for the water-table mapping. Being
20 familiar with the area, there are numerous groundwater bores constructed for the reasons
21 outlined in the paper. However there are two to three sets in the Gellibrand Valley. Many of
22 the bores are shallow and probably located in the near-surface alluvial aquifers that interact
23 directly with the rivers. However, there are numerous bores in the underlying confined
24 Eastern View Formation. The head levels in these two aquifers can be very different
25 (generally there are large upwards gradients between the Eastern View and the alluvials and
26 many of the deeper bores). Given the large number of bores in the area it is difficult to see
27 exactly which ones have been used for data analysis but presumably they are all in aquifers

1 that can be reasonably expected to be hydraulically connected such that a potentiometric
2 surface can be constructed. There needs to be more detail of which bores were used for this
3 analysis.

4 *Authors' response*

5 We have separated all bores from the original dataset of 88 bores in and around the catchment
6 into those with screened depths <40 m. The subset of shallow screened bores is considered to
7 represent the unconfined water table while the total dataset of bores represents more of a
8 potentiometric surface of the regional groundwater (typically in the Eastern View Formation).
9 The groundwater surfaces were mapped using both sets of bores and the results are presented
10 for both sets. The use of the smaller subset also addresses the question raised by Referee #1
11 about how sensitive the groundwater surface mapping is to a smaller set of bores being used.
12 We have also added additional supplementary material (Supplement B) detailing the bores
13 used in the study. The derivation of the two sets of groundwater surfaces are described in the
14 Methods (Section 2.2, see below) and incorporated into the analysis and results (e.g. Figs. 6,
15 7). The results and discussion have been described in our response to the first comment of
16 Referee #1

17 Section 2.2 (Methods), P7, L2-13.

18 “Eighty-eight groundwater monitoring bores in and around the boundary of the Gellibrand
19 catchment were identified and water level data were extracted from the Victorian
20 Groundwater Management System (http://www.vvg.org.au/cb_pages/gms.php). The area
21 contains a relatively large number of monitoring bores due to earlier investigations for a
22 potential damming of the Gellibrand River and also extraction of groundwater for urban water
23 supply (SKM, 2012). Groundwater surfaces were constructed from the total dataset and also
24 from a subset of 33 bores with screened depths of <40 m that only occur within the catchment
25 boundary (bore details in Supplement B). The total dataset contains bores that are screened at
26 greater depths in the Wangerrip Group (main aquifer) and these typically show higher heads
27 relative to nearby bores screened at shallower depths (typically in the Quaternary alluvium).
28 Groundwater surfaces from the total dataset represent more of a potentiometric surface while
29 the smaller dataset of shallow bores represents a water table surface.”

30

31 **Introduction**

1 1. I'm not sure exactly what you mean by "unconfined" – I presume that you mean an aquifer
2 that is intercepted by the stream rather than one which unconfined throughout the catchment.

3 *Authors' response*

4 We have refined our definition of unconfined groundwater to that suggested by Professor
5 Cartwright.

6 Section 1 (Introduction), P2, L21-25.

7 "The separation of baseflow contributions from regional groundwater (i.e. where aquifers are
8 unconfined in the vicinity of streams) from other shallower sources, like interflow, bank
9 storage return and perched aquifer discharge, is technically difficult to quantify. Nevertheless,
10 this is fundamentally important for quantifying how regional groundwater extraction may
11 affect baseflow in rivers (Wittenberg, 1999)."

12

13 2. Given that you use both tracers and physical parameters, it is worth mentioning that these
14 techniques often yield disparate results as they classify water differently. Specifically, as
15 transient stores of water (eg bank return flows) are likely to be chemically similar to the river,
16 then a chemical mass balance will record them as event water while a digital filter will record
17 them as part of the slow flow. The last paragraph on page 12407 views baseflow from the
18 physical perspective, from a geochemical perspective baseflow is all water that looks
19 chemically different from rainfall.

20 *Authors' response*

21 In the definition of baseflow we have noted that this is from a physical perspective and also
22 identified that tracer methods can deliver different results to physically based methods.

23 Section 1 (Introduction), P2, L29-30.

24 "From a physical perspective, the baseflow component of streamflow is the sum of the slow
25 flow pathways into the river (Ward and Robinson, 2000)."

26 Section 1 (Introduction), P4, L3-10.

27 "Tracer data are also commonly used to estimate groundwater discharge to streams (Cook et
28 al., 2003; McGlynn and McDonnell, 2003; Cartwright et al., 2011; Atkinson et al., 2015). The
29 tracer approach relies on the assumption that different contributors to streamflow have

1 distinctive and invariant chemical, isotopic or radiogenic end-member signatures that can be
2 apportioned in the streamflow mixture (McCallum et al., 2010). From a geochemical
3 perspective, mass balance estimates of baseflow using tracer data can differ from estimates
4 made by digital recursive filters as some slow flow components (e.g. bank storage) can be
5 geochemically similar to quick flow components (Cartwright et al., 2014).”

6

7 **Methodology**

8 This is generally clearly explained; however seen the comment above regarding the choice of
9 bores and the aquifers that they monitor. Also as discussed below, I think that your BFI value
10 needs more justification.

11 *Authors' response*

12 These two points have been addressed separately. The choice of bores is addressed in the
13 response to the general comment and the BFI value is addressed in the comments about
14 Results.

15

16 **Results**

17 1. The BFI used in the Eckhardt filter seems anomalously low. As explained in section 3.1,
18 values closer to 0.8 are expected for this type of catchment. Although you note this, do you
19 have an explanation? Adopting a BFI which minimises overestimates of baseflow wrt total
20 stream flow sounds logical, but are there other studies that you can point to which have done
21 this to lend some support for this methodology. I guess the related question is what the results
22 would be if a higher BFI were adopted?

23 *Authors' response*

24 The Eckhardt filter is generally applied using the constraint that estimated baseflow \leq total
25 streamflow for each time-step. This is an arbitrary constraint and was not applied in the
26 formulation of the filter (Eckhardt, 2005). We have taken the approach of investigating use of
27 the filter without this constraint (except in the comparison of baseflow to other estimates of
28 groundwater discharge in Fig. 9) and using BFI_{max} values derived from our own analysis and
29 from the literature (Eckhardt, 2005; Atkinson et al., 2015). We have followed the
30 recommendation of Professor Cartwright to investigate the effects of higher estimates of

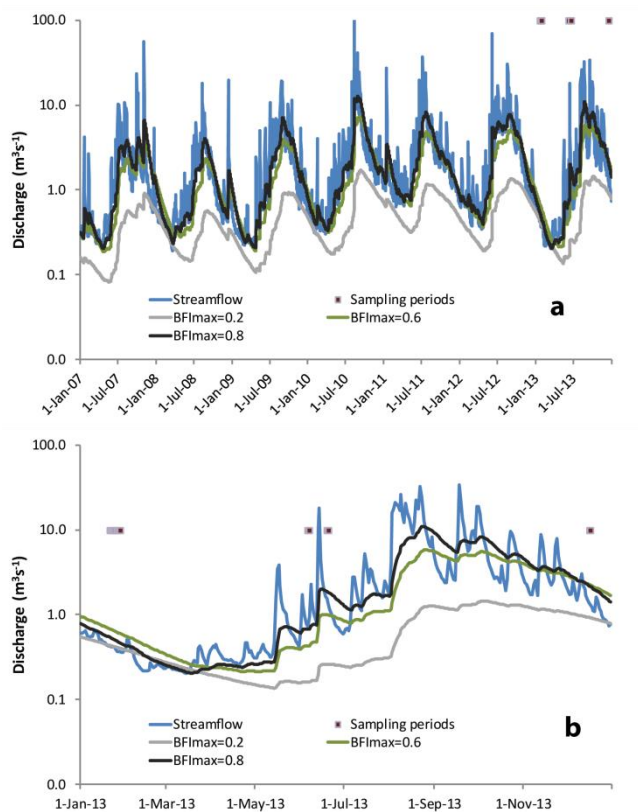
1 BFI_{max} in our analysis. Therefore, we have included in the analysis two additional baseflow
2 time-series using the Eckhardt (2005) recommendation of 0.8 for the BFI_{max} parameter and
3 also using the maximum BFI estimate of 0.6 identified using tracer analysis by Atkinson et al.
4 (2015) as the BFI_{max} value. The resulting differences in the estimated baseflow analysis using
5 the Eckhardt filter are shown in Figure 2 and also explained in the text in Section 2.3, 3.1, 3.4
6 (Fig. 6 – see response to comment 3 of Referee #1) and 4.1 (Fig. 9 – see response to comment
7 4 of Referee #1).

8 Section 2.3 (Methods), P9, L4-9.

9 “The BFI_{max} parameter (representing the maximum value of the baseflow index, i.e.
10 baseflow/total streamflow, that can be modelled by the filter algorithm) was chosen to
11 minimize periods of baseflow greater than observed streamflow. The filter is typically applied
12 with the condition that $b_k \leq Q_k$ (Eckhardt, 2005) but this is an arbitrary constraint and we
13 explore the resulting baseflow time-series without this condition, except where stated.”

14 Section 3.1 (Results), P10, L21-31, P11, L1-7.

15 “ The Eckhardt baseflow estimates produce patterns that follow the highly seasonal pattern
16 shown by the overall river discharge and indicated that baseflow significantly contributed to
17 overall streamflow (Fig. 2). The a parameter values declined moderately as the threshold flow
18 percentile value to define recession periods increased (30th – 0.990, 40th – 0.988, 50th –
19 0.985). The BFI_{max} parameter values that minimized periods of baseflow greater than
20 streamflow clustered around 0.2 but showed slight increases as a decreased (30th – 0.20, 40th –
21 0.20, 50th – 0.22). The resulting baseflow time-series using these parameter values were
22 similar and the time-series using $a=0.988$ and $BFI_{max}=0.20$ is shown in Fig. 2. This method
23 used for determining the BFI_{max} parameter produced values below the recommended range
24 (~0.8 for perennial rivers with porous aquifers, Eckhardt, 2005) and lie closest to the
25 recommended BFI_{max} value (0.25) for perennial rivers with hard rock aquifers. In Fig. 2 we
26 also show baseflow time-series using $a=0.988$ and the recommended BFI_{max} value for a river
27 such as the Gellibrand (0.80), and also using the maximum baseflow index value (0.60) found
28 for the Gellibrand River using tracer-based analysis by Atkinson et al. (2015). Using the
29 condition of $b_k \leq Q_k$, the filtered baseflow time-series produced mean monthly BFI estimates of
30 0.48-0.55 ($BFI_{max}=0.20-0.22$) and 0.63-0.58 ($BFI_{max}=0.60-0.80$) during the summer-autumn
31 period (December – May), and 0.21-0.24 ($BFI_{max}=0.20-0.22$) and 0.47-0.58 ($BFI_{max}=0.60-$
32 0.80) during the winter-spring period (June – November)”



1

2 Figure 2. (a) Hydrograph at Bunker Hill gauging station (235227) illustrating the seasonality
 3 of flow. Three baseflow separation hydrographs generated using different BFI_{max} parameter
 4 values (0.20, 0.60, 0.80 and $a=0.988$) for the Eckhardt filter are displayed, along with the
 5 periods of hydrochemical sampling of streamflow during 2013. (b) Hydrograph for 2013 to
 6 illustrate the detail of variations in baseflow using different BFI_{max} values without the
 7 constraint of $b_k \leq Q_k$.

8

9 **2.** I am not certain that the stable isotopes (section 3.3) add very much to this study. The
 10 values (Fig. 4) overlap and the differences between the sampling rounds are subtle. The
 11 interpretation on Page 12417 that the lower ^{18}O values in winter possibly reflect differences
 12 in source or imply a short residence time may be correct although some of the difference
 13 could be related to evaporation in the warmer months increasing ^{18}O (and this probably
 14 should be mentioned if the data are retained). Without the estimate of evaporation, it is
 15 difficult to use the stable isotopes for mass balance (especially given the large relative
 16 variability in the groundwater).

17 **Authors' response**

1 The isotope data do not support a large influx of groundwater discharge into streamflow and
2 this is consistent with the major ion analysis and groundwater surface volume changes.
3 However, the minor differences in isotopic end-member signatures between regional
4 groundwater and tributary streamflow and possibility of evaporitic enrichment mean that the
5 isotope data do not strongly contribute to the findings of the paper. Therefore, we have
6 accepted the suggestion of Professor Cartwright and removed Figure 4 and associated isotopic
7 analysis from the paper.

8

9 **3.** Section 3.4. I am not certain that that Fig. 6 shows the difference between March and
10 September (page 12420, line 10); looking at the caption to Fig. 6, it seems to be just the
11 September data (depth to water and the SD of the kriging)? This needs clarification.

12 *Authors' response*

13 In the original Figure 6 (Fig. 5 in the revised version) the March 2009 areas of artesian
14 groundwater were shown as polygons and this was shown in the legend of the figure.
15 However, these polygons have been removed from the revised figure as the focus of the figure
16 has changed to comparing groundwater surfaces generated from the potentiometric and water
17 table datasets. In the text, to cover the point being made in the original Fig. 6, we have stated
18 that the spatial locations of areas of shallow groundwater do not vary greatly between months.

19 Section 3.4 (Results), P14, L9-17, P13.

20 “For both groundwater datasets the results are generally not consistent with changes in the
21 saturated area being the dominant driver of peak variations in baseflow, as measured by the
22 Eckhardt filter. In particular, the potentiometric dataset shows a far more consistent range in
23 seasonal peaks compared to the digital filter estimated baseflow. While the water table dataset
24 does show a similar pattern in seasonal peaks, the water table rarely reaches the land surface,
25 The saturated areas largely coincided (e.g. see Fig. 5) and were restricted to the valley floor of
26 the catchment and with little variation in the location of these areas between dates. The
27 restriction of the saturated areas to the valley floors indicates little regional groundwater
28 discharge into minor tributaries and this is analysed further in Sect. 3.5.”

29

30 **Discussion**

1 Section 4.1. The chemical mass balance would be improved by the discussion of uncertainties
2 as noted by one of the other reviewers. Possibly propagating the variability in the
3 groundwater composition through the calculations would achieve this. Additionally, the
4 impact of the assigned BFI could be considered (especially as it appears to be lower than
5 expected).

6 ***Authors' response***

7 We have addressed the uncertainties in the chemical mass balance by varying the groundwater
8 end-member compositions by one standard deviation for each ion and the results are shown in
9 Table 1 and described in Section 3.3 (see response to comment 2 of Referee #1). The
10 relationship between the mass balance tracer estimates and different baseflow filter estimates
11 (particularly those generated with higher BFI_{max} values) are shown in Fig. 9 (see response to
12 comment 4 of Referee #1).

13

14 **Conclusions**

15 Some perspective regarding the impact that bore numbers and bore density has on the results
16 would be useful to researchers considering applying this to other catchments.

17 ***Authors' response***

18 We have added the following text to Conclusions to provide the requested perspective.

19 Section 5 (Conclusions), P21, L2-20.

20 “ Geostatistical mapping of unconfined groundwater surfaces provides a useful, independent
21 dataset for investigating sources of fluxes contributing to baseflow estimated by traditional
22 digital filter and tracer end-member approaches. In particular, the method can provide added
23 confidence in the lower bound of baseflow estimates that best correspond to regional
24 groundwater discharge in both low and high flow periods. Specifically, the groundwater
25 surface dataset can be used to identify whether variations in discharge area (i.e. groundwater
26 intersecting the land surface) or saturated volume can explain seasonal variations in baseflow,
27 as estimated using digital filters. This dataset is particularly useful in humid, hilly catchments
28 where interflow or perched aquifer discharge is likely to be a significant process and where
29 the different ‘slow flow’ fluxes have similar low salinity chemistry that hinders end-member
30 analysis. Sufficient monitoring bore data to construct water table maps are not available in all

1 catchments and the method is likely to be restricted to catchments where groundwater
2 investigations have resulted in the existence of an adequate bore network. The adequacy of
3 the network will depend on catchment size, the spatial distribution of bores (i.e. uniform
4 versus non-uniform distribution, location relative to the drainage network) and the spatial
5 correlation of the monitored water level. However, where adequate monitoring data are
6 available, this method adds significant value to water resource management by making better
7 use of an independent, but often under-utilised, dataset that can inform groundwater
8 contributions to streamflow.”

9

10 **References**

- 11 Abedini, M. J., Nasser, M., and Burn, D. H.: The use of a genetic algorithm-based search
12 strategy in geostatistics: application to a set of anisotropic piezometric head data,
13 *Comput. Geosci.*, 41, 136-146, 2012.
- 14 Atkinson, A. P., Cartwright, I., Gilfedder, B. S., Cendón, D. I., Unland, N. P., and Hofmann,
15 H.: Using 14C and 3H to understand groundwater flow and recharge in an aquifer
16 window, *Hydrol. Earth Syst. Sci.*, 18, 4951–4964, 2014.
- 17 Atkinson, A. P., Cartwright, I., Gilfedder, B. S., Hofmann, H., Unland, N. P., Cendón, D. I.,
18 and Chisari, R.: A multi-tracer approach to quantifying groundwater inflows to an upland
19 river; assessing the influence of variable groundwater chemistry, *Hydrol. Process.*, 29, 1-
20 12, 10.1002/hyp.10122, 2015.
- 21 Bredehoeft, J. D., Papadopoulos, S. S., and Cooper, H. H. Jr.: Groundwater: the water budget
22 myth in scientific basis of water resource management. National Research Council
23 Geophysics Study Committee, National Academy Press, Washington, DC, pp 51–57,
24 1982.
- 25 Brutsaert, W.: Long-term groundwater storage trends estimated from streamflow records:
26 Climatic perspective, *Water Resour. Res.*, 44, W02409, doi:10.1029/2007WR006518,
27 2008.
- 28 Cartwright, I., Gilfedder, B., and Hofmann, H.: Contrasts between chemical and physical
29 estimates of baseflow help discern multiple sources of water contributing to rivers,
30 *Hydrol. Earth Syst. Sci.*, 18, 15–30, 2014.

- 1 Eckhardt, K.: How to construct recursive digital filters for baseflow separation, *Hydrol.*
2 *Process.*, 19, 507-515, 2005.
- 3 Love, A. J., Herczeg, A. L., Armstrong, D., Stadter, F., and Mazor, E.: Groundwater flow
4 regime within the Gambier Embayment of the Otway Basin, Australia: evidence from
5 hydraulics and hydrochemistry, *J. Hydrol.*, 143, 297–338, 1993.
- 6 Nwankwor, G. I., Cherry, J. A., and Gillham, R. W.: A comparative study of specific yield
7 determinations for a shallow sand aquifer, *Ground Water*, 22, 764-772, 1984.
- 8 Ortiz, C. J. and Deutsch, C. V.: Calculation of uncertainty in the variogram, *Math. Geol.*, 34,
9 169-18, 2002. Perrin, J., Mascré, C., Pauwels, H., Ahmed, S.: Solute recycling: An
10 emerging threat to groundwater quality in southern India, *J. Hydrol.*, 398, 144-154, 2011.
- 11 SKM, (2010), Glenelg Hopkins CMA groundwater model – final model development report,
12 Report to the Victorian Department of Sustainability and Environment, available at:
13 https://ensym.dse.vic.gov.au/docs/GlenelgHopkins_TransientModelReport_FINAL.pdf,
14 last access: 5 February 2015, 2010.
- 15