

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

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

4
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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 **Anonymous Referee #2**

19 **General Comments**

20 Most of my comments echo those of the previous reviewer, for example I question how well
21 this method would work with fewer data points, and I also question why the authors didn't
22 use heterogeneous subsurface properties provided they were present.

23 ***Authors' response***

24 Please see responses and changes made to General Comments 1 and 2 of Referee #1

25 **Specific comments**

1 1. Following the usefulness of this method with reduced data points, were any of the kriging
2 parameters varied to determine the uncertainty relative to these parameters?

3 *Authors' response*

4 Further information is provided in the Methods section for how the kriging parameters used in
5 the groundwater mapping were calibrated. Essentially, the parameters were optimised to
6 reduce the variance in the mapped surfaces. The increase in the variance between the
7 potentiometric and water table sets of maps (see response to comment 1 of Referee #1) shows
8 that using fewer data points does result in increased uncertainty and changes in the Kriging
9 parameter values.

10 Section 2.2 (Methods), P7, L14-32, P8, L1-20.

11 “ In order to construct groundwater surface maps for specified dates, the periodic (generally
12 monthly) water level observations of the bore data were first modelled using the nonlinear
13 transfer-function-noise time-series modelling methodology of Peterson and Western (2014).
14 Water level estimates for the start of each month were then derived by adding the time-series
15 simulation, interpolated to the required data, to a univariate ordinary kriging estimate of the
16 timeseries model error at the required date, which ensured a zero error at dates with a water
17 level observation. groundwater surface maps were then produced for the first of each month
18 for the years 2007 to 2010 using the Kriging with external drift (KED) method (Peterson et
19 al., 2011). In applying the KED, the external drift term was the land surface elevation (Shuttle
20 Radar Terrain Model (SRTM) 30 m dataset). A model variogram was derived for the
21 component of the groundwater elevation not explained by the external drift. The KED
22 approach requires the estimation of three parameters for the residual model variogram and a
23 parameter for the maximum search radius during the mapping. Considerable effort was taken
24 to reliably calibrate the variogram parameters and set a search radius producing cross-
25 validation residuals that are approximately first-order stationary. The Kriging variance (see
26 example in Fig. 6) does provides an indicative estimate of the map reliability for the given
27 parameter set and the available water level observations. However, the density and location of
28 observations also influences the variogram parameters and the maximum search radius
29 parameter. Accounting for this parameter uncertainty in the groundwater mapping is not
30 trivial and future work is required to explore methods that account for variogram uncertainty
31 (Ortiz et al., 2002) and localised estimation of the search radius (Abedini, 2012). This
32 groundwater level component was first estimated using ordinary least squares regression and

1 then minimised by repeatedly fitting an isotropic exponential variogram, using multi-start
2 Levenberg-Marquardt optimization and re-derivation of the water level component, until a
3 stable model variogram was achieved. The depth to groundwater was calculated by difference
4 from the SRTM representation of the ground surface and used to measure changes in the
5 percentage of the catchment with very shallow groundwater surfaces (nominally “saturated“
6 within the uncertainty range of the groundwater surface position) over the period of mapping.
7 This was done for the parts of the catchment with an elevation of <100 m in order to analyse
8 changes in the saturated area around the valley floor and lower slopes of the catchment where
9 most monitoring bores were located and hence confidence in the groundwater surface
10 mapping was highest.”

11 **2.** Was any consideration given to whether the saturated areas fell in regions where surface
12 water was present (i.e. within the streambed) given that these areas would vary with stream
13 stage? For groundwater discharging to regions with little to no surface water present, was ET
14 taken into consideration?

15 ***Authors’ response***

16 The measurement of saturated areas within the catchment was used as a first-order
17 approximation of the interaction of the groundwater with the land surface. As such, we did
18 not consider how much of the saturated areas fell within the streambed and neither was ET
19 taken into consideration. Stream stage could result in local reversals in gradient between
20 groundwater and streamflow but could not be accurately determined at the scale of the 30 m
21 DEM used in the water table mapping. These aspects are addressed in Section 3.4 and 4.2.

22 Section 3.4 (Results), P14, L32-33, P15, L1-9.

23 “ We use a range of realistic but relatively high (Nwankwor et al., 1984) specific yield values
24 from 0.05-0.3 for the different geological units within the <100 m elevation mask for the
25 groundwater surfaces (see Fig. 1). The estimates of the ratio of monthly baseflow (from
26 Eckhardt filter) to monthly mapped volume change, shown in Table 2, are generated using the
27 same specific yield values across all geological units and also by varying the values consistent
28 with expected hydrogeological properties (i.e. specific yield of alluvium > Wangerrip Group
29 > Heytesbury Group). We consider that this range of estimates based on these specific yield
30 values provides an upper bound to the groundwater discharge, particularly since any phreatic
31 evapotranspiration flux, which would also account for some of the volume changes, is not
32 considered.”

1 Section 4.2 (Discussion), P19, L1-7.

2 “ Fluctuations in the water table remain a relatively coarse measure and provide only a first-
3 order estimate of possible groundwater discharge patterns. For instance, the mapping does not
4 have the resolution to identify the fine detail of channels and near-stream zones. Stage
5 variations in channels will have local effects on groundwater recharge and discharge that are
6 not captured by the groundwater mapping. Likewise, capillary fringing effects in near-stream
7 zones could lead to rapid increases in the water table with a small rise in water content in the
8 unsaturated zone (Gillham, 1984).”

9

10 3. Both references used on page 12407 lines 4 & 10 were found to contain significant content
11 that was improperly cited from other sources. It is recommended that the authors find the
12 original sources of the information in this work and cite those instead.

13 ***Authors’ response***

14 The original references have been deleted or replaced by the Bredehoeft et al. (1982)
15 reference.

16 Section 1 (Introduction), P2, L14-17.

17 “ It has been long recognised that over-extraction from aquifers may result in significant long-
18 term declines in groundwater levels resulting in decreases in baseflow in rivers (Bredehoeft et
19 al., 1982).”

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21 4. Page 12419 Lines 7-10 are a little redundant.

22 ***Authors’ response***

23 These lines have been removed.

24

25 **References**

26 Abedini, M. J., Nasser, M., and Burn, D. H.: The use of a genetic algorithm-based search
27 strategy in geostatistics: application to a set of anisotropic piezometric head data,
28 Comput. Geosci., 41, 136-146, 2012.

- 1 Atkinson, A. P., Cartwright, I., Gilfedder, B. S., Cendón, D. I., Unland, N. P., and Hofmann,
2 H.: Using ^{14}C and ^3H to understand groundwater flow and recharge in an aquifer
3 window, *Hydrol. Earth Syst. Sci.*, 18, 4951–4964, 2014.
- 4 Atkinson, A. P., Cartwright, I., Gilfedder, B. S., Hofmann, H., Unland, N. P., Cendón, D. I.,
5 and Chisari, R.: A multi-tracer approach to quantifying groundwater inflows to an upland
6 river; assessing the influence of variable groundwater chemistry, *Hydrol. Process.*, 29, 1-
7 12, 10.1002/hyp.10122, 2015.
- 8 Bredehoeft, J. D., Papadopoulos, S. S., and Cooper, H. H. Jr.: Groundwater: the water budget
9 myth in scientific basis of water resource management. National Research Council
10 Geophysics Study Committee, National Academy Press, Washington, DC, pp 51–57,
11 1982.
- 12 Brutsaert, W.: Long-term groundwater storage trends estimated from streamflow records:
13 Climatic perspective, *Water Resour. Res.*, 44, W02409, doi:10.1029/2007WR006518,
14 2008.
- 15 Cartwright, I., Gilfedder, B., and Hofmann, H.: Contrasts between chemical and physical
16 estimates of baseflow help discern multiple sources of water contributing to rivers,
17 *Hydrol. Earth Syst. Sci.*, 18, 15–30, 2014.
- 18 Eckhardt, K.: How to construct recursive digital filters for baseflow separation, *Hydrol.*
19 *Process.*, 19, 507-515, 2005.
- 20 Love, A. J., Herczeg, A. L., Armstrong, D., Stadter, F., and Mazor, E.: Groundwater flow
21 regime within the Gambier Embayment of the Otway Basin, Australia: evidence from
22 hydraulics and hydrochemistry, *J. Hydrol.*, 143, 297–338, 1993.
- 23 Nwankwor, G. I., Cherry, J. A., and Gillham, R. W.: A comparative study of specific yield
24 determinations for a shallow sand aquifer, *Ground Water*, 22, 764-772, 1984.
- 25 Ortiz, C. J. and Deutsch, C. V.: Calculation of uncertainty in the variogram, *Math. Geol.*, 34,
26 169-18, 2002. Perrin, J., Mascré, C., Pauwels, H., Ahmed, S.: Solute recycling: An
27 emerging threat to groundwater quality in southern India, *J. Hydrol.*, 398, 144-154, 2011.
- 28 SKM, (2010), Glenelg Hopkins CMA groundwater model – final model development report,
29 Report to the Victorian Department of Sustainability and Environment, available at:

- 1 https://ensym.dse.vic.gov.au/docs/GlenelgHopkins_TransientModelReport_FINAL.pdf,
- 2 last access: 5 February 2015, 2010.
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