

# Response

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Many thanks for reviewing the manuscript and for your comments and suggestions. Please see below the answers to your review comments.

## Reviewer 1

### The Review

This work presents visual diagnostic tools for different methods of projection pursuit optimization. They use line plots, PCA and the tour method to visually compare the behavior of different optimizers. Finally, they showcased the analysis of three optimizers using their tool.

The tool is likely to be of interest to the R community. Most of my comments are about the technical aspect of the content. The writing is overall good, but please see my comment under the “writing” regarding to the clarity of the text.

### Major comment

- “Sometimes, two optimisers can start with the same basis but finish with bases of opposite sign... However, this creates difficulties for comparing the optimisers since the end bases will be symmetric to the origin. A sign flipping step is conducted...” and the “Reconciling the orientation” section
  - For 1D projections, flipping the sign is sufficient. For p-D projections of  $\mathbb{R}^n$ , one may need a little more. To my knowledge most index functions are invariant to any rotations that happens within the p-subspace. This makes me wonder that in general, one may want to align two bases so that their difference does not contain any “within-subspace” components. In other words, one may want to compare representations in the Grassmannian manifold (space of k-subspaces) instead of in the Stiefel manifold (space of orthonormal k-frames).

With p-D bases, the correction of orientation is a bit more complicated. Take 2D as an example, two bases can be aligned in the first axis but not in the second. This would means there are  $2^p$  possibility to check.

Once you go to 2D you have to take into account rotation in addition to flips, so the thing to check would be if the two frames are spanning the same subspace, and aligning them will be a bit more work. Might be a good idea to add 1-2 sentences in the paper and rephrase the reply.

### Minor comment

- The software seems to be specific to projection pursuit. I wish the authors could comment more on its potential generality to other optimizations. For example, can one generalizes the tour method to visualize any parameter trace of an optimizer?

It is feasible to generalise the tour method for diagnosing other optimisers, providing the parameter space can be simulated. In our examples, the parameter to trace (projection base  $(p \times d)$ ) lives in a p-sphere, which can be simulated by the ‘geozoo’ package. For other optimisers, reserachers need first to create a

data collection object in a tibble that track the parameter during the optimisation, and then to use fern to produce the diagnostic plots as shown in the paper.

providing -> given that the ...?

Here could reiterate points about constrained optimisation and that we simulate from the allowed parameter space, in those cases using the tour method to visualize the trace is especially interesting. Is there any documentation of how the data structure needs to look like for fern input? It might be a good idea to point to that.

- Figure 4 "... All the bases in PD have been flipped for easier comparison of the final bases..."
  - This is related to the major comment. In the figure, PD flipped in sign but stopped at a basis representation similar to the destination of the CRS. Flipping the sign back, does it mean that PD and CRS diverged but stopped at the similar subspace with opposite basis representations? Also, given that PD and CRS have the same starting bases and one is flipped, why wouldn't the dashed line pass the origin (gray dot) in the figure?

Yes the reviewer is giving a decent explanation in describing the effect of flipping. The gray dot should land on the dashed line and this is a mistake due to the data not being updated after flipping in the code. This has now been corrected.

Thanks for sharing your interpretation on the effects of flipping, that's right!...

- "Recall that the columns in a 2D basis are orthogonal to each other, so the space of  $p \times 2$  bases is a  $p$ -D torus".
  - It is not clear to me why "so". For  $p=2$ , the space of  $2 \times 2$  bases is a pair of 1-D tori instead of just one (because you have two orientations for a pair of orthogonal vectors), right? For  $p=3$ , the space of  $3 \times 2$  bases is  $SO(3)$ , which is not a 3-D torus? Ref: [https://en.wikipedia.org/wiki/Stiefel\\_manifold](https://en.wikipedia.org/wiki/Stiefel_manifold)

We changed the wording of the sentence to "\*\*\*a torus in the  $p$ -D space\*\*"

I think Di mentioned a reference to cite when you first mention the torus in the text, it should be added to the paper and mentioned here in the reply.

## Writing

- "Along with the bases recorded during the optimisation and a zero basis, the parameters (centre and radius) of the 2D space can be estimated by PCA. The centre of the circle is the first two PC coordinates of the zero matrix, and the radius is estimated as the largest distance from the centre to the basis."
  - It took me multiple passes to get what this means. Perhaps instead starting with something like: "We use PCA to project and visualize the parameters/bases in 2D. The centre of the 2D view is..."?

The paragraph has now been updated to "We use PCA to estimate the 2D space that contains all the bases recorded and the zero basis. The centre of the 2D view is the first two PC coordinates of the zero matrix, and the radius is estimated as the largest distance from the centre to the basis."

Change -> The paragraph has now been updated to...

You could be a bit more careful with the phrasing still, the PCA finds the 2D space that is best suited to visualize the bases (while any 2D projection will contain them).

- "Figure 6 shows some frames from the tour plot of the same two optimisations in its original 5D space."
  - I think this is the first mention of "5D space"/5D data in the main text (apart from the caption in Figure 4), so it may be worth clarifying what the original 5D space is.

The explanation of that data set is introduced in the subsection \*Simulation setup\* under \*Diagnosing an optimiser\*. However, to avoid ambiguity, we change the "original 5D space" to "original space" and give more details about the original space in the figure caption: "The basis space in this example is a 5D unit sphere ...".

The data set is first introduced in subsection .... However, to avoid ambiguity -> However, to avoid confusion

- “While the previous few examples have looked at the space of 1D bases ... 2D bases”
  - I don’t think the terms “1D bases” and “2D bases” are standard. Perhaps (orthonormal) “1-frames” and “2-frames”?

Those terms have now been changed and the sentence now reads: "While the previous few examples have looked at the space of 1-frames in a unit sphere, this section visualises the space of 2-frames."

I changed to ->Those terms have now been changed and the sentence now reads:

## Reviewer 2

These experiments are interesting and so is the methodology. The conclusions could do with beefing up What can we learn about how to optimise the holes index for 1 and 2d? Does this apply to other smooth indices? What about non-smooth indices? What about indices in other dimensions?

A paragraph further summarising the findings has been added in the second paragram in the conclusion

The new paragraph should be improved a bit further. Can you connect how the method was useful to how other people might want to use it? I think Fig 10 is a nice example of how we understand the optimizers both in terms of how well they climb towards higher index values, how well they cover the search space and also how direct the path towards the optimum is. This lead to insights for the guided tour example, and would be useful generally when comparing optimizers on difficult problems.

1. Should refer to earlier work on scagnostic indices, page 3

New references have been added as follows:

Added reference of (Wilkinson et al., 2005) and (Wilkinson and Wills, 2008):

L. Wilkinson and G. Wills. Scagnostics distributions. *Journal of Computational and Graphical Statistics*, 17(2):473–491, 2008

L. Wilkinson, A. Anand, and R. Grossman. Graph-theoretic scagnostics. In *IEEE Symposium on Information Visualization*, 2005. INFOVIS 2005., pages 157–164, Oct 2005.

2. figure 2 caption “using the holes index” “and a boxplot”. What is the colour scale for the points? The text related to figure 2 in the bottom of page 5 should mention the dataset and the index, giving references if needed.

Fixed language. Added mention the color scale in the figure 2 caption: "The color scale is from the customised botanical palette in the fern package."

The caption has now been re-written and color scale explanation added

You should mention what color is indicating here, is it just a way of distinguishing iterations or does it carry any additional information?

3. “with higher index value” -> “with a higher index value”. The word “the” is missing in a few other places in the paper.

That has now been corrected.

4. In Figure 3 would be better if the y-axes scales were the same for the two graphs.

It is now fixed

5. What is the star in figures 4 and 5? This is not mentioned in the text or caption

The caption has now been ammended to: "The black star represents the theoretical best basis the optimisers aiming to find." in the caption of figure 4.

6. In the text you say things like “details=TRUE” and “animate=TRUE” but what functions are you referring to?

Mention of "details=TRUE" and "animate=TRUE" is removed from subsection diagnostic 3a and 3b to the implementation section: 'explore\_space\_pca()' produces the PCA plot of projection bases on the reduced space. Figure 4 includes the additional details of anchor and search bases, which can be turned on by the argument 'details = TRUE'. The animated version in Figure 5 is produced with argument 'animate = TRUE'.

7. In Figure 9, it seems CRS is doing better here than in Figure 8— maybe this could be mentioned.

Figure 8 and 9 focuses on different aspects: Figure 8 compares the effect of implementing an interruption while Figure 9 compares the effect of polishing. Cross comparing the two experiments is less meaningful.

It would still be interesting to mention what was different, i.e. were all the settings the same and the better performance is just random (by different random seed), or did you change some setting of the optimizer that might have helped improve the performance?

8. “not the case for all the indexes” missing a “.” page 11

It has now been added

9. For the Kolmogorov index, why are you using randomly generated data instead of normal quantiles?

The Kolmogorov Smirnov index here looks at the difference between a theoretical pdf and an empirical one based on the data. In our case we are not assuming normality on the theoretical pdf and that is the reason why we are not using normal quantiles.

I think the reviewer is confused here because in the text it is written Normal Kolmogorov. However, the definition is just the standard Kolmogorov Index, right? Shall we just changed to standard Kolmogorov Index instead?

10. Fix references with “et al”

Fixed!