

Optimal Unstructured Mesh Data Compression and Reconstruction Using Compressed Sensing

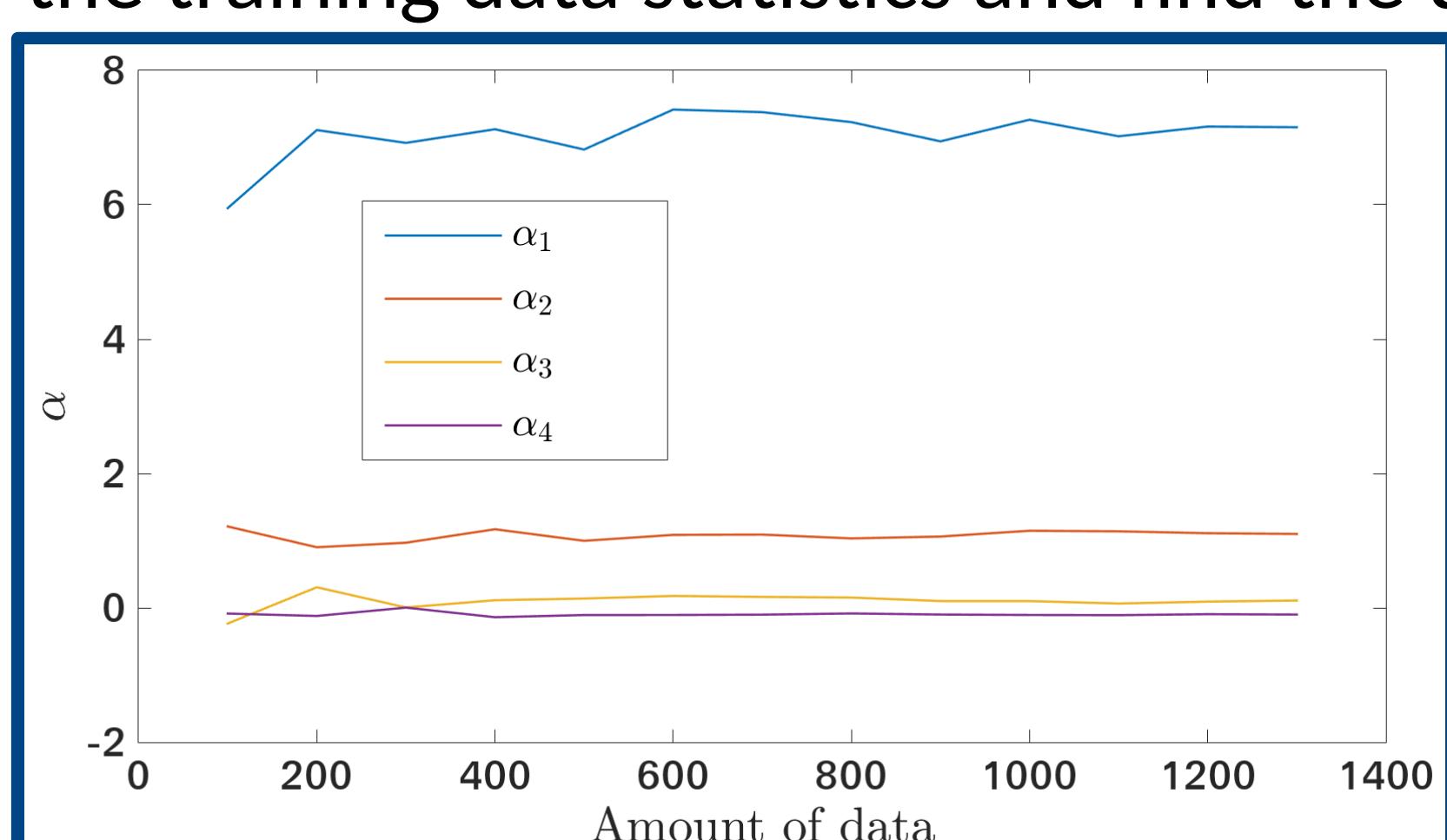
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Introduction

- **Challenge:** Compress large amounts of data efficiently by determining a relationship between the data features and how well they compress.
- **Solution:** **Compressive sensing** (CS) samples the data within an alternative function space using second generation wavelets and then reconstructs the data.
- CS is fast and, unlike regular wavelet compression, it *allows interactive reconstruction and visualization* according to the required accuracy and quality.
- We empirically determine a correlation that produces an *optimal compression ratio* (R) depending on the gradient of the input function being compressed.
- Applied to 2D irregular meshes and geometries

Procedure

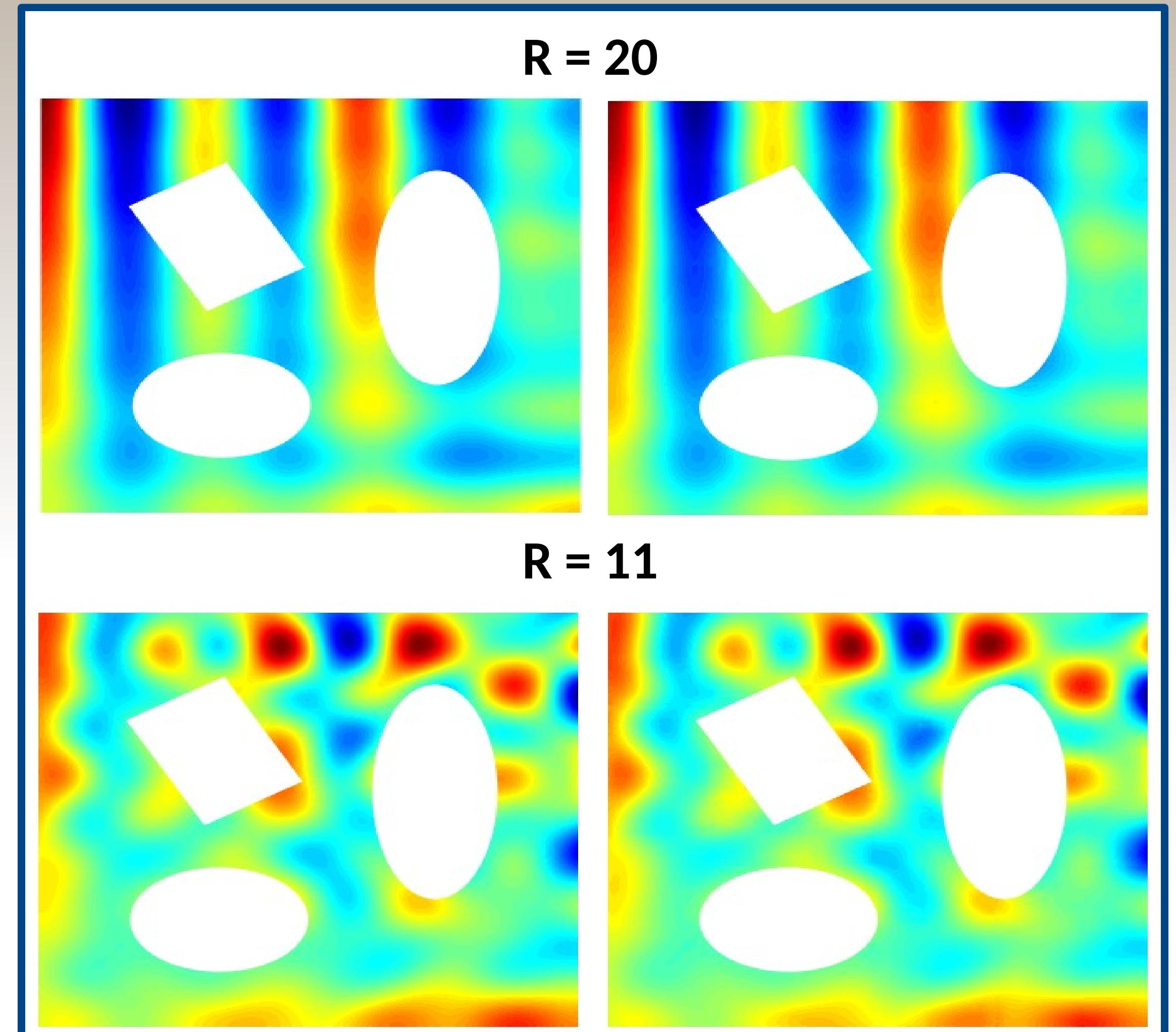
1. Generate hundreds of functions using a combination of sines and cosines
2. Compress the large vector f that represents a function using a sampling matrix Φ into the much smaller y and measure the R that grants a reconstruction accuracy at a given tolerance $y = \Phi f$
3. Compute several statistics (mean, median, etc.) for the gradients of f
4. Propose a correlation model for the compression ratio R as a function of the input statistics ($M = \max$). Fit the model below to the training data statistics and find the α_i



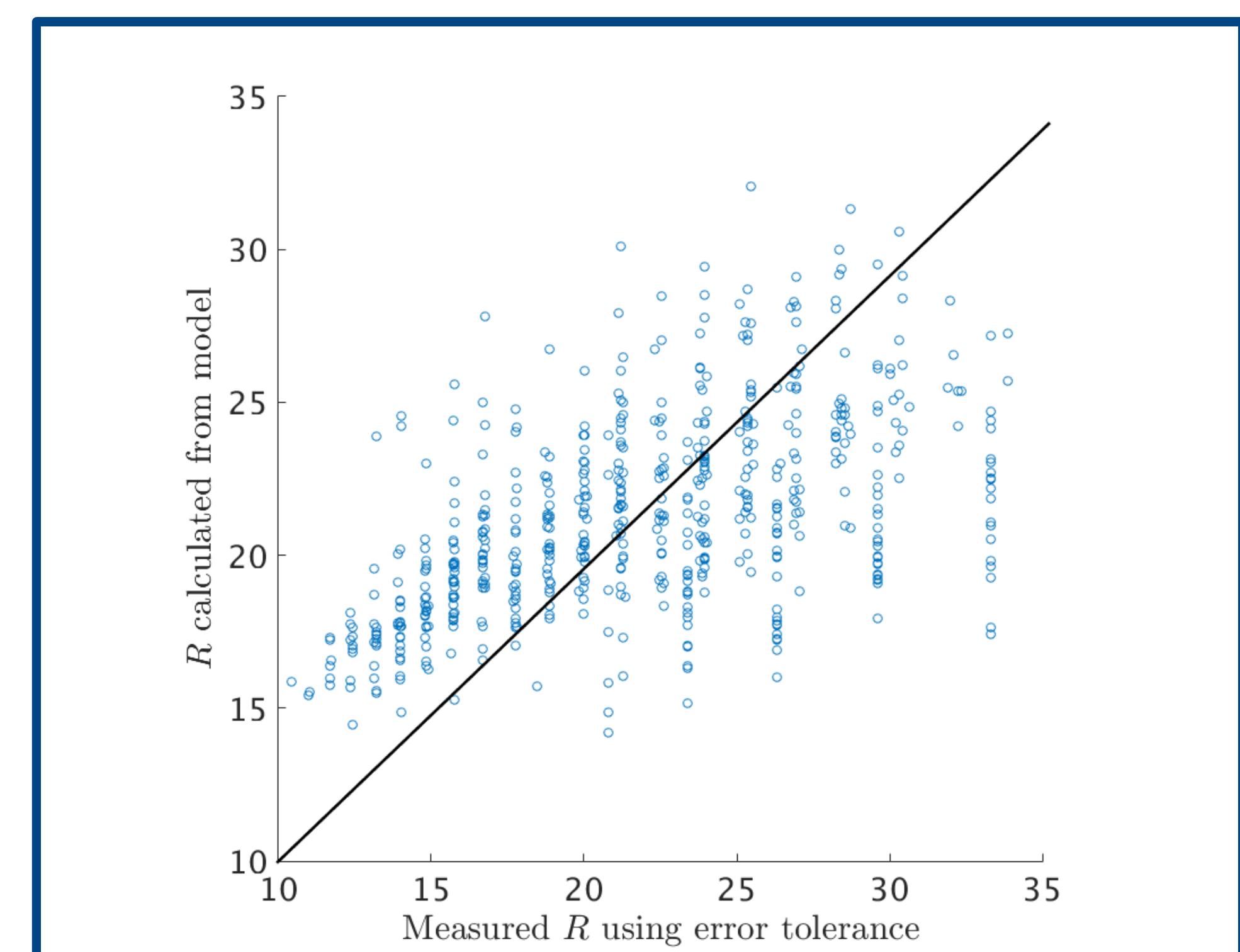
$$R = (e^{\alpha_1}) / (\mu^{\alpha_2} \sigma^{\alpha_3} M^{\alpha_4})$$

5. Test new functions by using the obtained correlation to compute a compression ratio, compress the function, and reconstruct the function. Ideally, the reconstructed image is difficult to distinguish from the original image.

Results and Conclusion



Two functions (~33,000 mesh points per function) compressed and reconstructed using a correlation model and CS. We find that as the function complexity decreases, the optimum R increases.



R calculated from the correlation model is plotted against the measured R. For 1000 functions, the optimal compression ratios are overestimated for small R and underestimated for large R.

- **Future work:** Will aim to fill in the “gaps” between the clusters of R and apply this knowledge to realistic 3D mesh data, as well as try different models for R.

References:

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2. Donoho, D. L., Tsaig, Y., Drori, I. and Starck, J.L. Sparse solution of underdetermined systems of linear equations by stagewise orthogonal matching pursuit. IEEE Transactions on Information Theory, 58(2):1094-1121, 2012.