

Design Automation & Optimisation for Multi-Material 3D Printing

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1 Algorithm comparisons

Current work

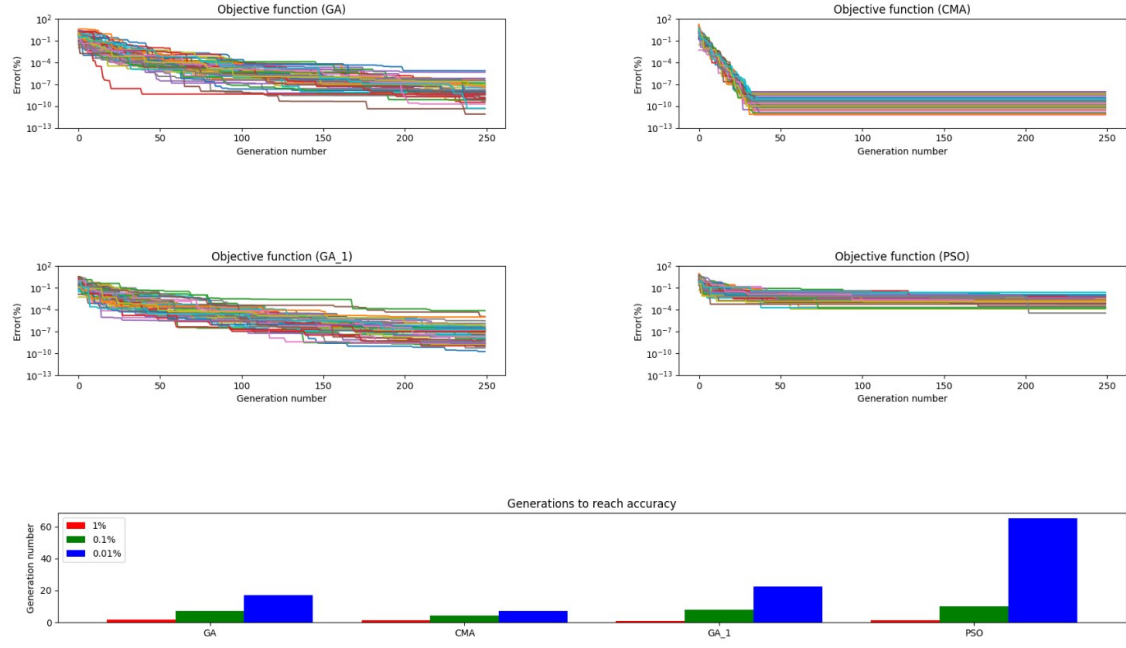


Figure 1: Comparison of algorithms for a 1D beam

- 1D 40 voxel beam using PLA and AI
- Objective function: Sum of absolute differences to target frequencies
- Trying to match a known beam - comparing GA, CMA-ES (binary encoding), PSO, GA (adaptive) in 250 generations
- All algorithms perform fairly well, reaching $< 1\%$ error in 250 generations.
- CMA-ES performed the best, but hard to draw conclusions.

Future work

- Test using 2D 40 voxel and flex/PLA/steelfill
- Test using custom GA crossover
- Test using time as stop criterion, not number of generations
- Test using sum of squares as objective function

2 Population size testing

Current work

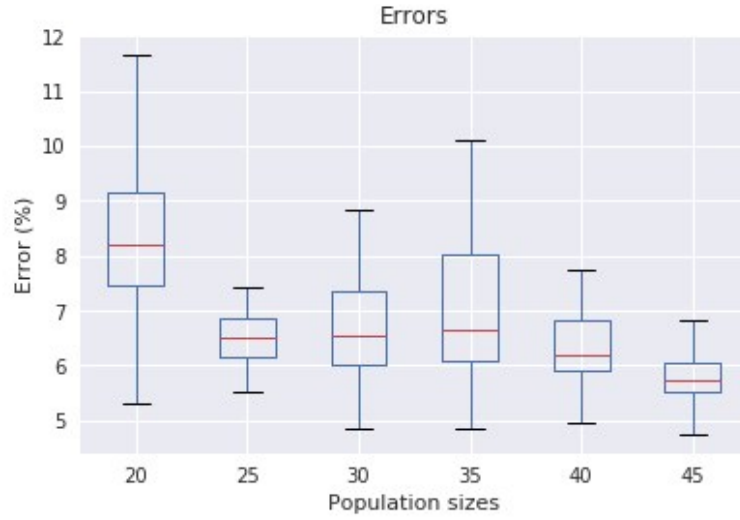


Figure 2: Population size vs final error

- 2D 40 voxel beam using PLA and AI
- Objective function: Sum of absolute differences to ($f1 = 200$, $df = 600$)
- Testing population size for GA
- Population size ≥ 25 seems to be the best, with minimal gains after (6.5% vs 8.5%)
- This ignores time taken - assumes number of cores ≥ 45

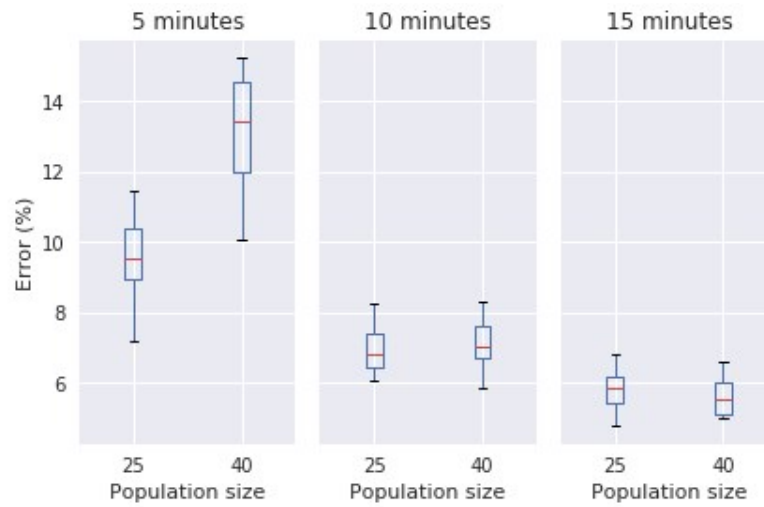


Figure 3: Population size vs time vs final error

- 2D 40 voxel beam using PLA and AI
- Objective function: Sum of absolute differences to ($f1 = 200$, $df = 600$)
- Testing population size for GA, using time as stop criterion
- Difficult to draw any conclusion here

Future work

- Test using flex/PLA/steelfill
- Test using custom GA crossover
- Test using machine with large number of cores.
- Repeat second experiment with a larger range of populations
- Test using sum of squares as objective function

3 Uniform objective function

Current work

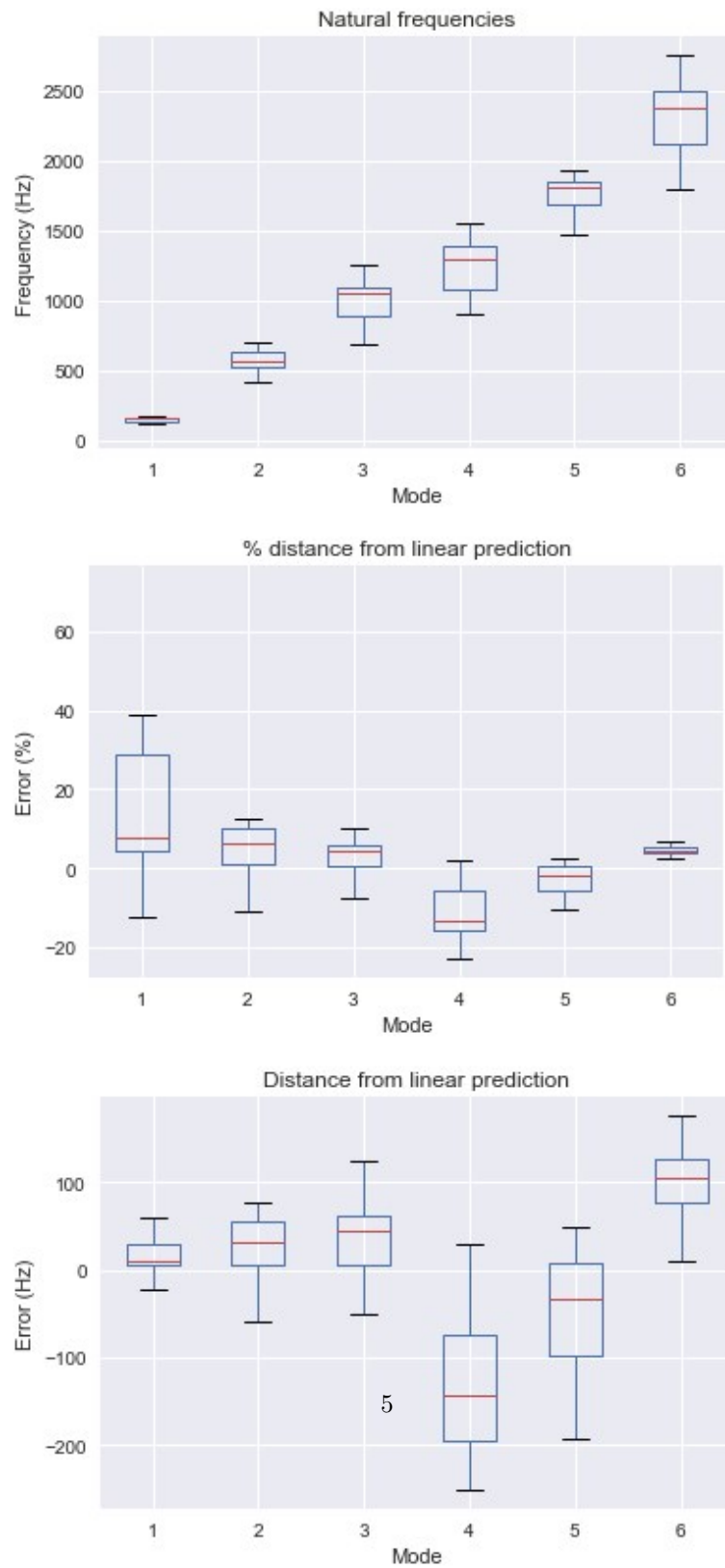


Figure 4: Linear fit as objective function

- 2D 40 voxel beam using PLA and Al
- Objective function: Fit to a line, minimizing r^2
- Using GA for 50 generations
- $r^2 = 0.98 \pm 0.008$, $f1 = 140 \pm 30$, $df = 420 \pm 40$ ¹

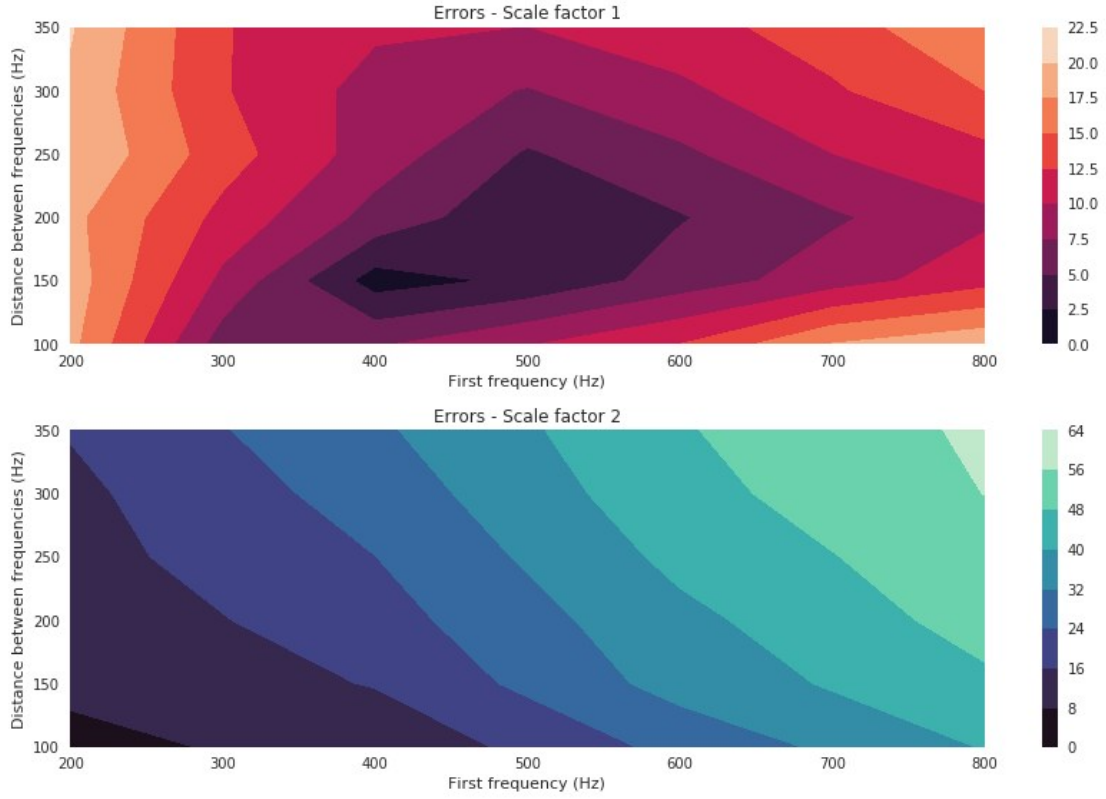


Figure 5: Uniform fit over a range of frequencies

- 2D 40 voxel beam using PLA and Al, doubled dimensions for second graph
- Objective function: Sum of absolute differences to (f1, df)
- Using GA for 1 hour
- Best spot around $f1 = 200$, $df = 600$ ²

¹<https://github.com/byi649/Design-Automation-Optimisation-for-Multi-Material-3D-Printing/blob/master/Results/2018-05-02/Statistical%20analysis%20-%20uniform.ipynb>

²<https://github.com/byi649/Design-Automation-Optimisation-for-Multi-Material-3D-Printing/blob/master/Results/2018-07-06/Statistical%20analysis%20uniform.ipynb>

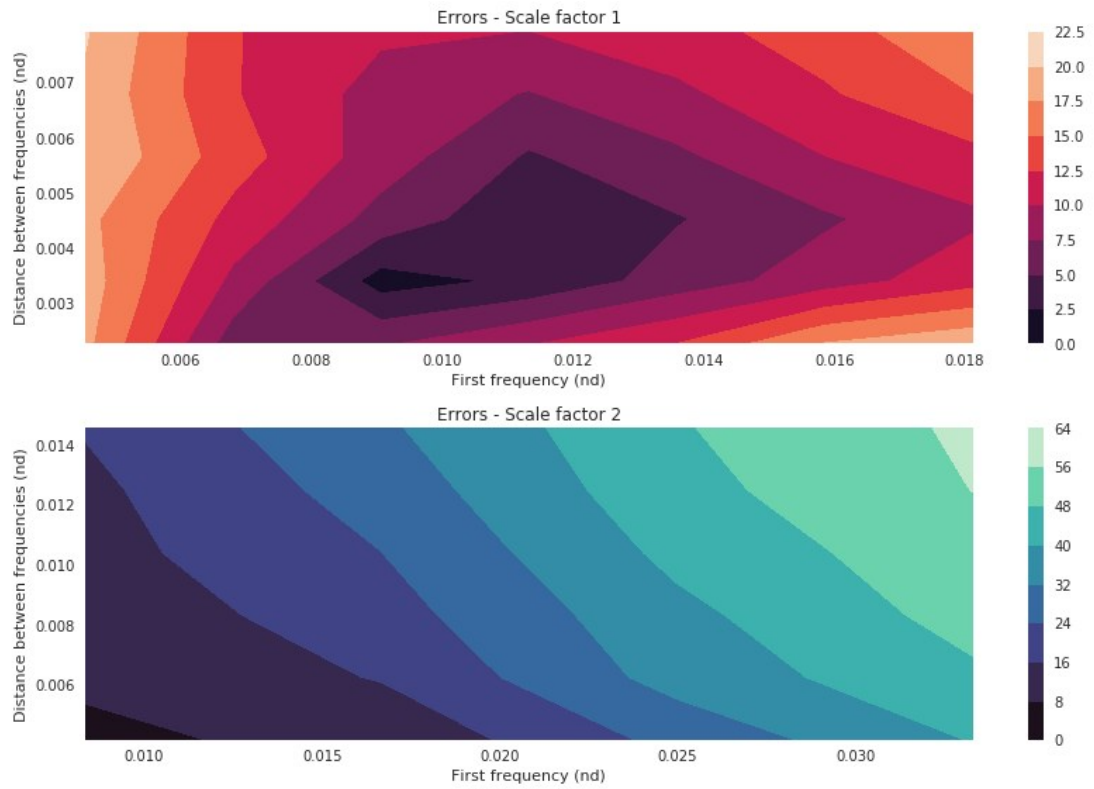


Figure 6: Uniform fit over a range of frequencies (non-dimensionalised)

- Same as above, but non-dimensionalised by average wave speed
- Best spot around $f_1 = 0.009$, $df = 0.0035$

Future work

- Test using flex/PLA/steelfill
- Test using custom GA crossover
- Test using time as stopping criterion
- Test more beam sizes
- Non-dimensionalise by individual wave speed
- Test using sum of squares as objective function

4 Voxel number

Current work

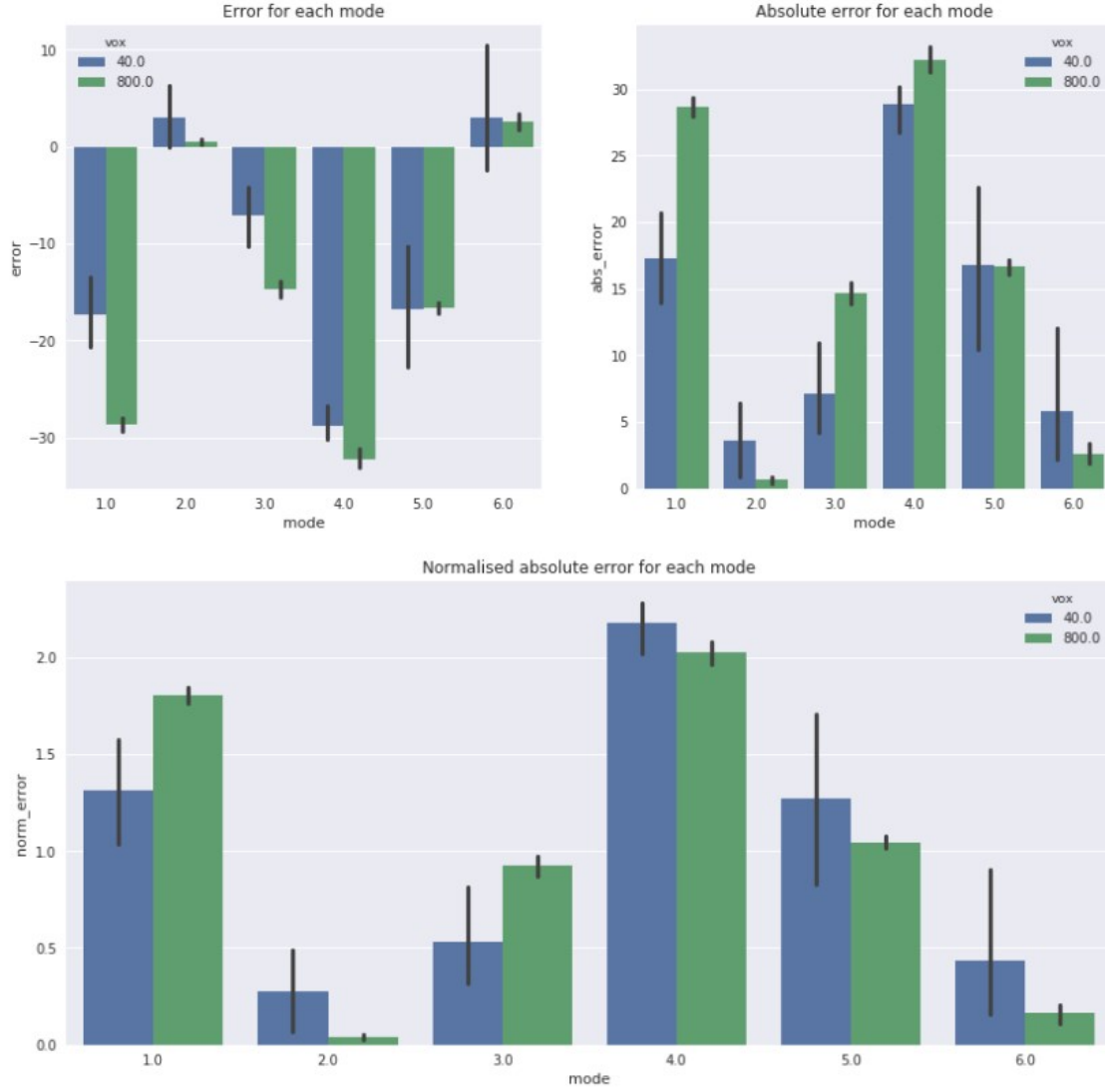


Figure 7: 40 voxel vs 800 voxel comparison

- 2D 40 voxel beam using PLA and AI, 3D 800 voxel beam using PLA and AI
- Objective function: Sum of absolute differences to ($f1 = 200$, $df = 600$)
- Using GA for 1 hour

- 40 voxel performs better than 800 voxel³, no other patterns found

Future work

- Test using flex/PLA/steelfill
- Test using custom GA crossover
- Test using sum of squares as objective function

³<https://github.com/byi649/Design-Automation-Optimisation-for-Multi-Material-3D-Printing/blob/master/Results/2018-06-08/Statistical%20analysis.ipynb>

5 Custom crossover

Current work

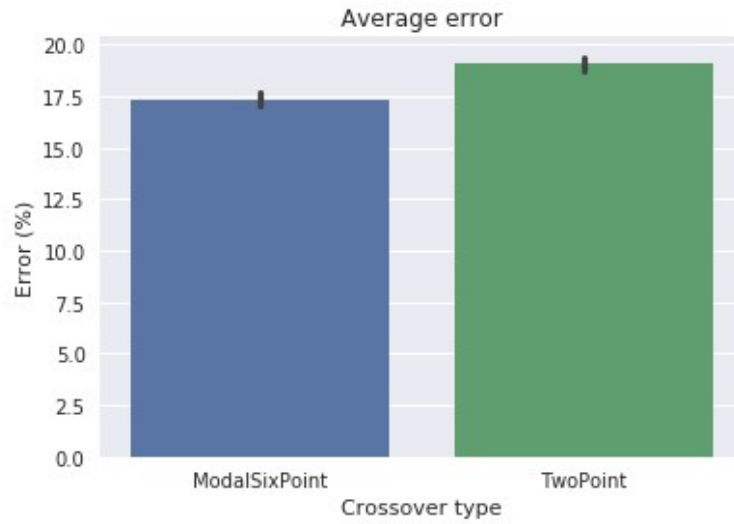


Figure 8: Custom crossover vs standard 2 point crossover

- 3D 800 voxel beam using PLA and Al
- Objective function: Sum of absolute differences to ($f1 = 200$, $df = 600$)
- Using GA for 1 hour
- 6-point crossover performs better than stock 2-point crossover (17.5% vs 19% error)⁴, p-value = $3.8e-8$.

⁴<https://github.com/byi649/Design-Automation-Optimisation-for-Multi-Material-3D-Printing/blob/master/Results/2018-06-29/Statistical%20analysis.ipynb>

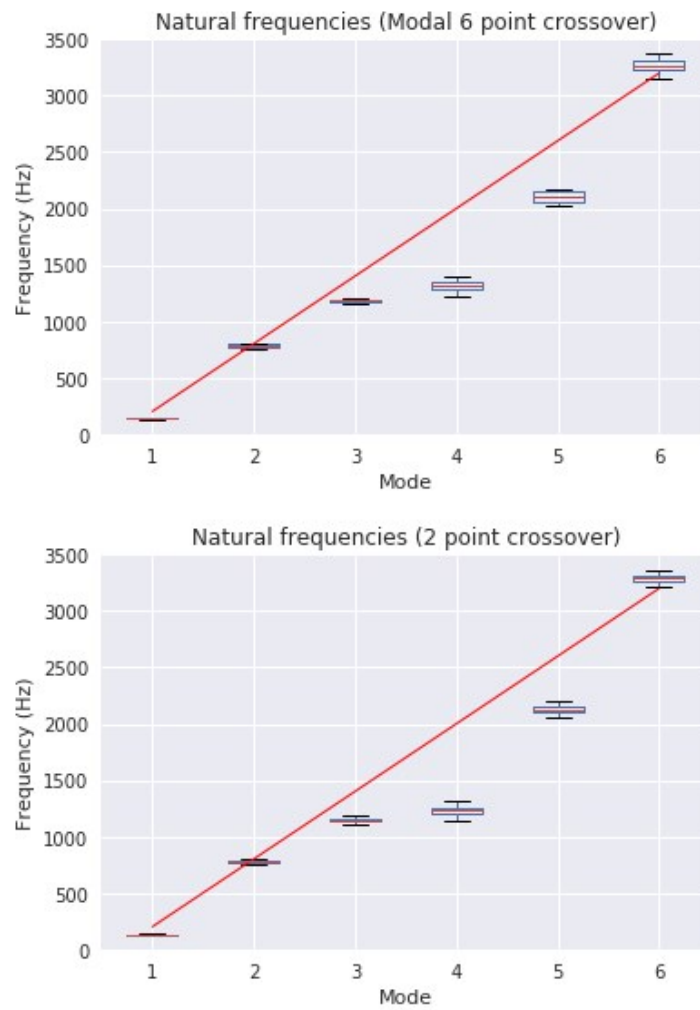


Figure 9: Custom crossover mode frequencies

- Same as above, looking into individual frequencies matched
- Mode 4 perhaps slightly better?

Future work

- Test using flex/PLA/steelfill
- Test using sum of squares as objective function
- More indepth tests on custom crossover usefulness (testing individual frequencies etc)

6 Clustering

Current work

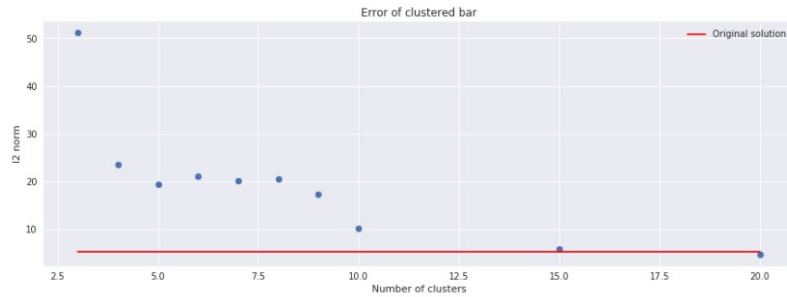


Figure 10: Clustering introduced error

- 2D 40 voxel beam using flex and steelfill
- Objective function: Sum of squares to ($f1 = 200$, $df = 600$)
- Using CMA-ES for one hour, then clustering via scikit-learn
- Breakpoints at $n.clusters = 5, 10, 15$ (final error = 20%, 10%, 5%)

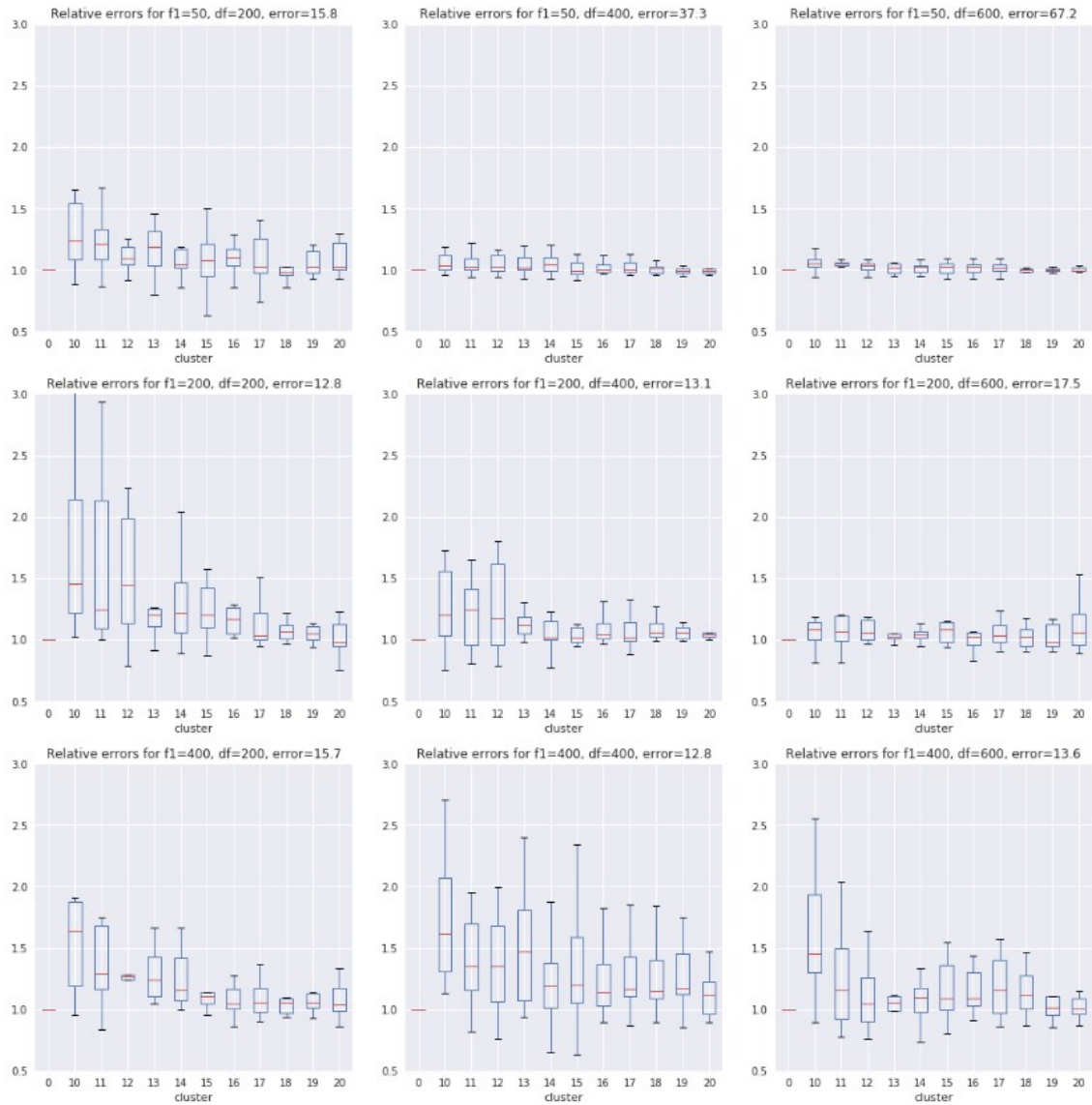


Figure 11: Clustering introduced error for many frequencies

- 2D 40 voxel beam? using ? and ?
- Objective function: Sum of squares to (f_1 , df)
- Using CMA-ES for one hour, then clustering via scikit-learn
- Errors are normalised within each plot

Future work

- Test using flex/PLA/steelfill
- Test using sum of squares as objective function