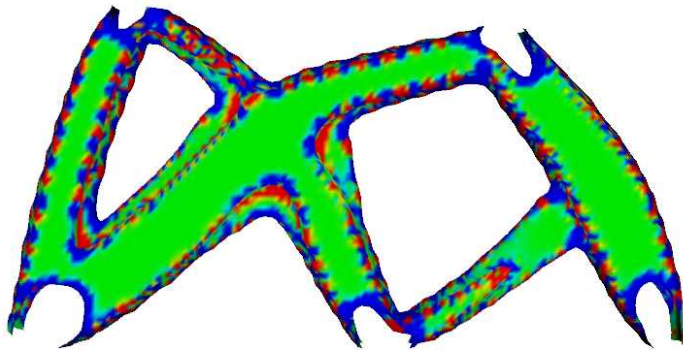


# REMOVING LOW-FREQUENCY ARTEFACTS FROM TOPOLOGY-OPTIMIZED SURFACES

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# Topology-optimized parts



Thermal



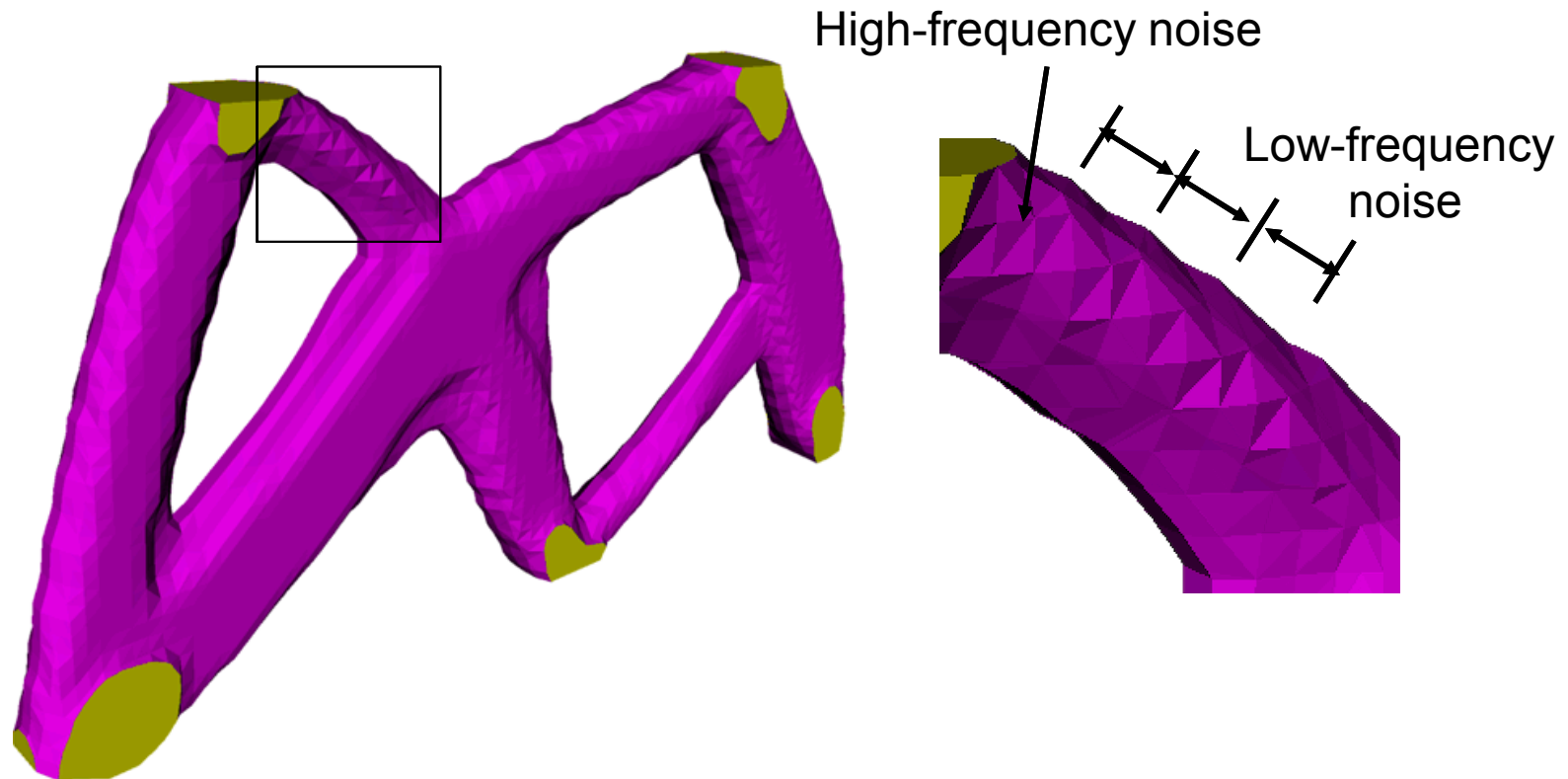
Thermal + Structural



Structural

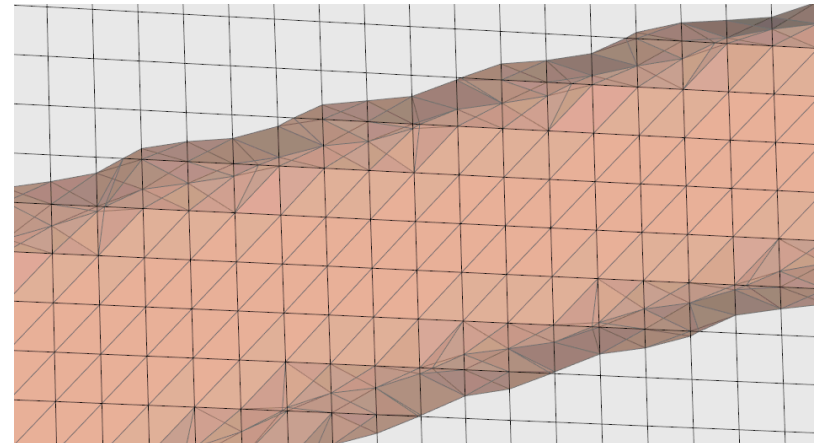
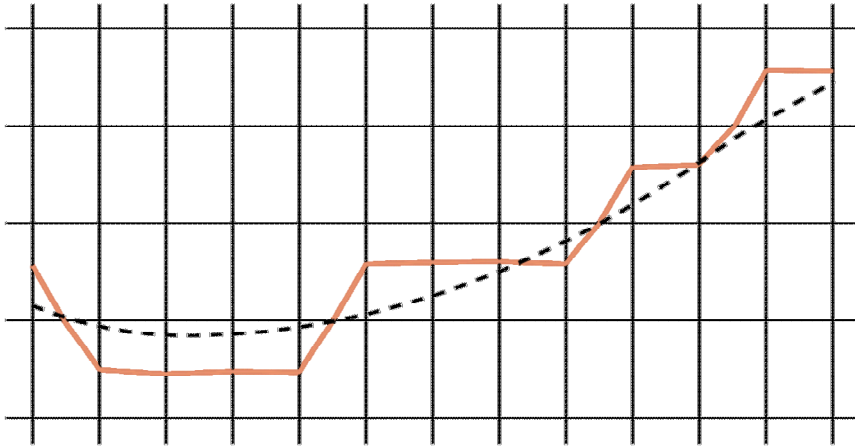
# Error categorization by frequency

- High-frequency: “random noise”
- Low-frequency: “ripples”



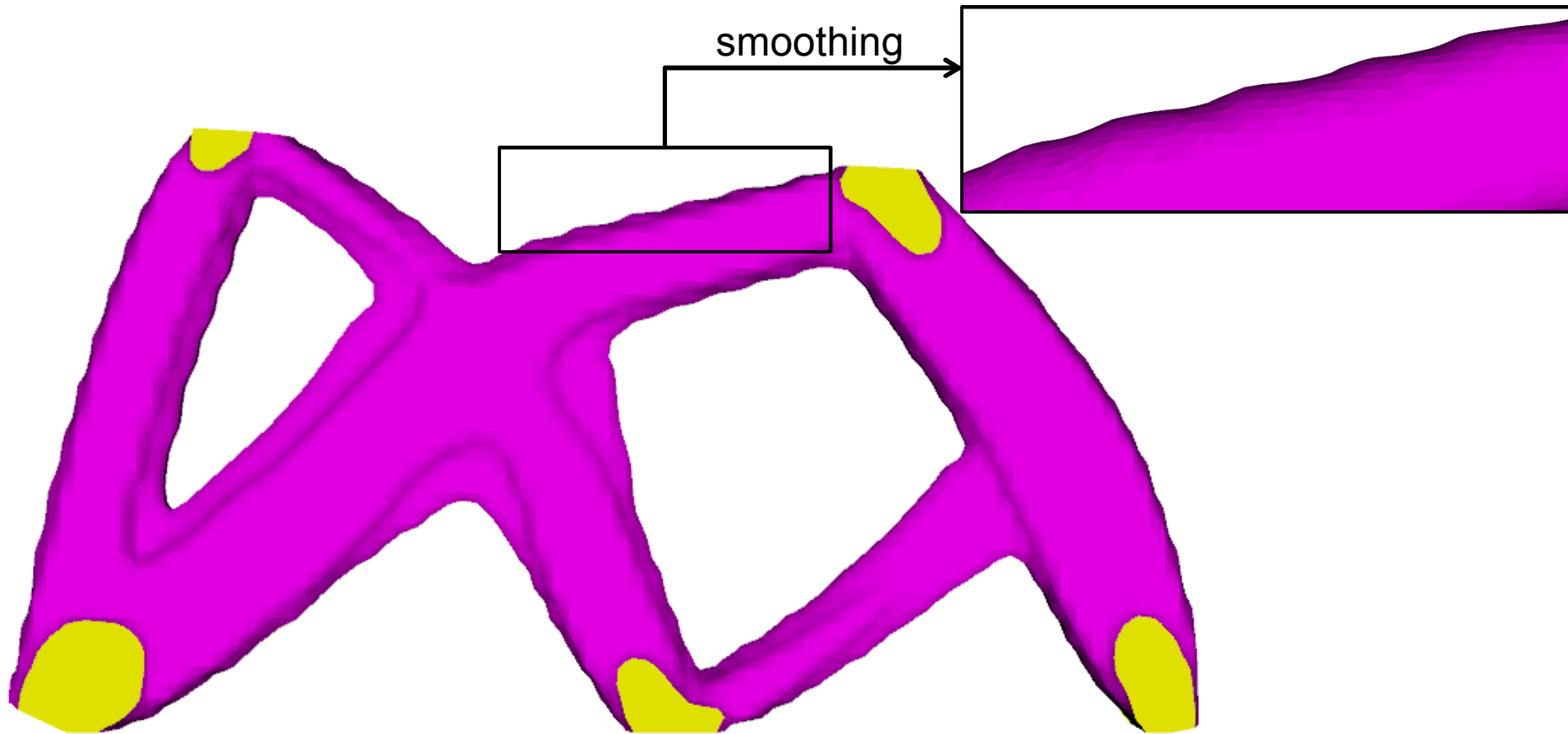
# Density penalization

- Topology optimization is computed on a density field
- Nonbinary densities are penalized, causing “snapping” near the surface



# Persistence of low frequencies

- Local “averaging” methods often preserve ripples



# How to define frequency?

- Signals  $\mathbf{x}$  are column vectors containing a value for each of the  $N$  nodes. Laplacian = matrix  $\mathbf{L}$ .
- Eigenvalue problem:

$$\mathbf{L} \mathbf{v}_i = \lambda_i \mathbf{v}_i, \quad i = 1, 2, \dots, N$$

**Frequencies**

**Normal Modes**

- Differential Equations 101:  $\sin \omega x$  are eigenfunctions of

$$L = -\nabla^2 = -\frac{d^2}{dx^2}$$

# Properties


- **Completeness:**

$$\mathbf{x} = \sum_{i=1}^N c_i \mathbf{v}_i \quad \text{Fourier series}$$

- **Orthogonality:**

If  $\mathbf{L} = \mathbf{L}^T$  then

$$\mathbf{v}_m^T (\mathbf{L} \mathbf{v}_n) = (\mathbf{v}_m^T \mathbf{L}^T) \mathbf{v}_n = (\mathbf{L} \mathbf{v}_m)^T \mathbf{v}_n$$


$$\lambda_n (\mathbf{v}_m^T \mathbf{v}_n) = \lambda_m (\mathbf{v}_m^T \mathbf{v}_n)$$

If the  $\lambda_i$  are distinct, then

$$(\mathbf{v}_m^T \mathbf{v}_n) = \delta_{mn}$$

# Explicit frequency method

- Once  $\mathbf{v}_i$  are known, find  $c_j$  with inner product:

$$\mathbf{x} = \sum_{i=1}^N c_i \mathbf{v}_i$$

$$\mathbf{v}_j^T \mathbf{x} = \sum_{i=1}^N c_i \mathbf{v}_j^T \mathbf{v}_i = \sum_{i=1}^N c_i \delta_{ji} = c_j$$

- Reconstruct  $\mathbf{x}$  applying transfer function  $f$ :

- $\mathbf{x}' = \sum_{i=1}^N f(\lambda_i) c_i \mathbf{v}_i$



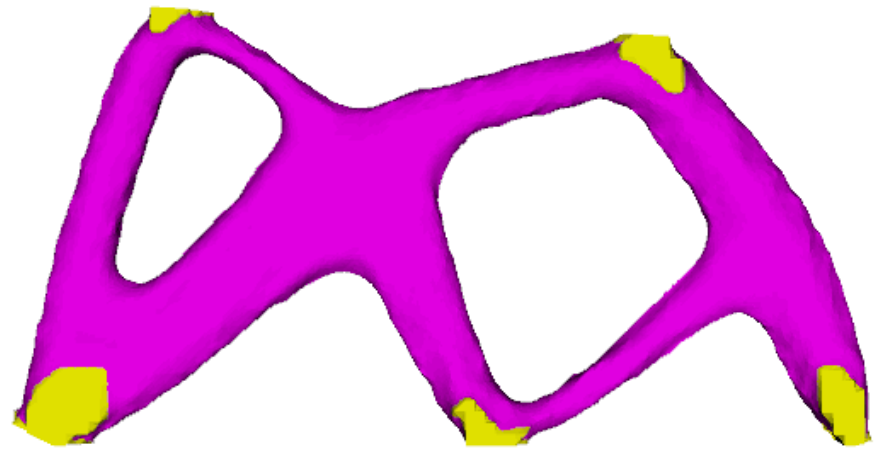
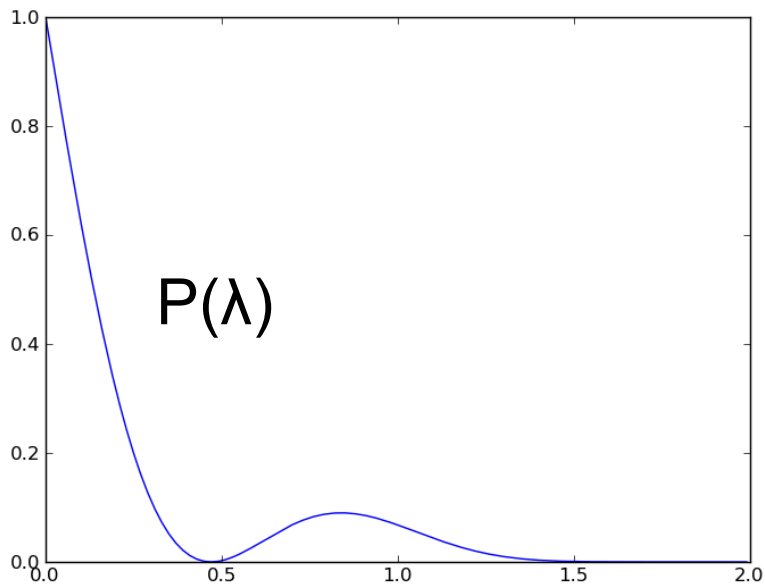
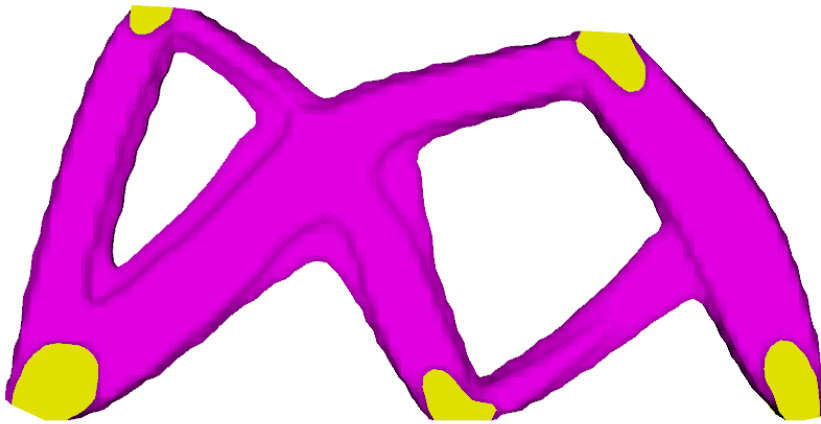
# Polynomial filters

- $P(\mathbf{L})$  has eigenvalues  $P(\lambda_i)$ .

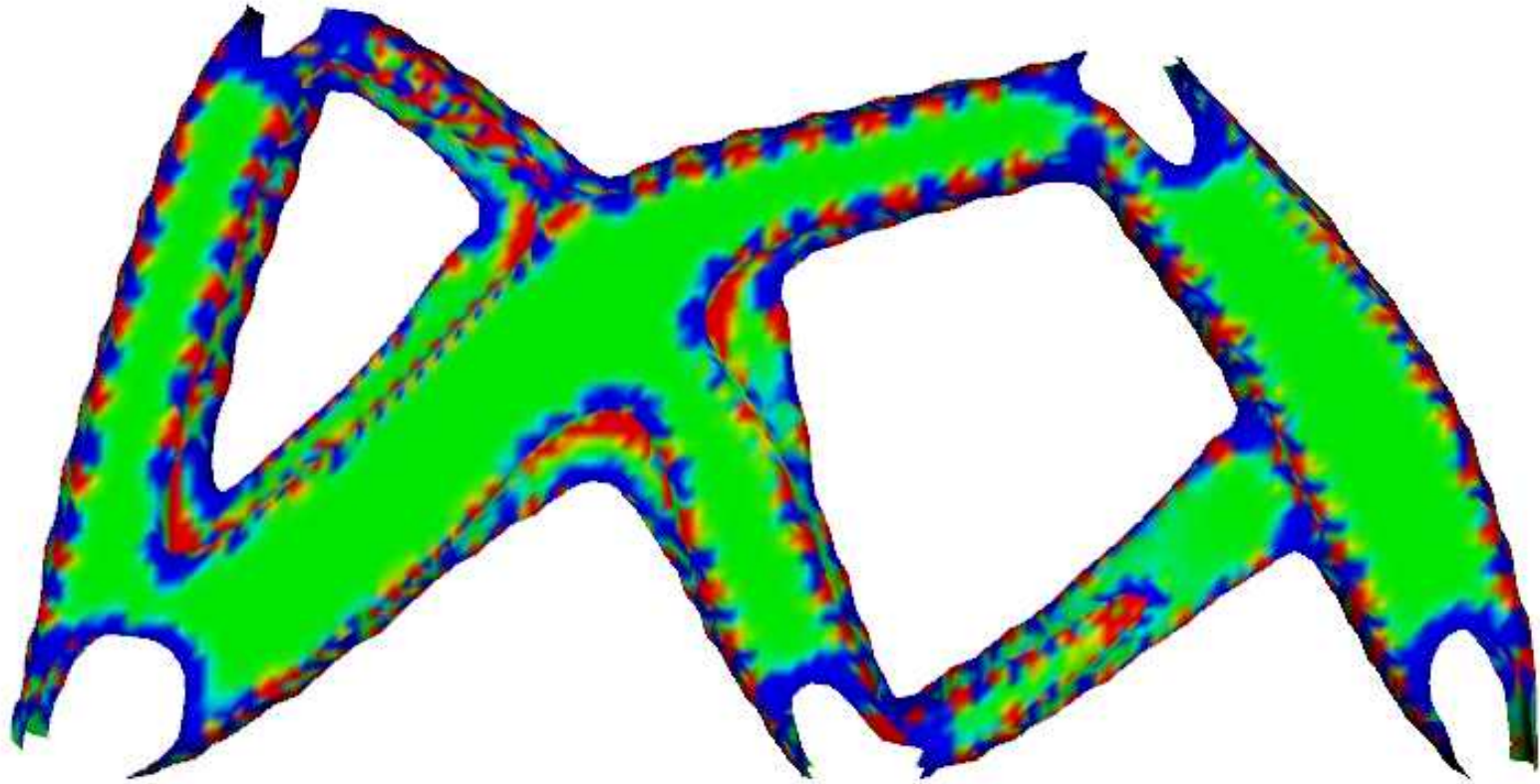
$$\mathbf{x} = \sum_{i=1}^N c_i \mathbf{v}_i$$

$$P(\mathbf{L}) \mathbf{x} = \sum_{i=1}^N c_i P(\mathbf{L}) \mathbf{v}_i = \sum_{i=1}^N P(\lambda_i) c_i \mathbf{v}_i$$

# Polynomials are hard to control!



# Curvature highlights ripples



positive

zero

negative

# Proposed method

1. Construct signal  $\gamma$ . Set  $\gamma = 1$  for red nodes and  $\gamma = 0$  elsewhere.

2. Project into frequency space:

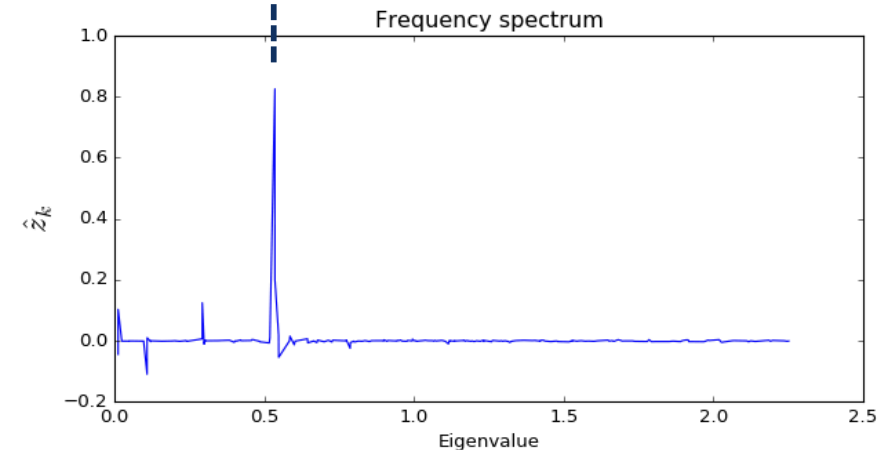
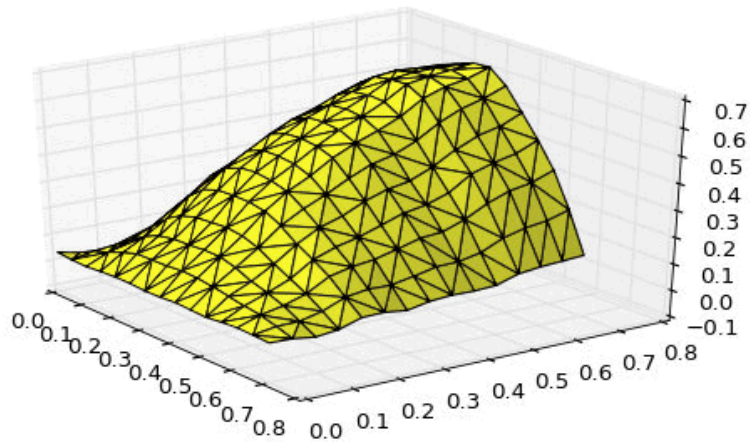
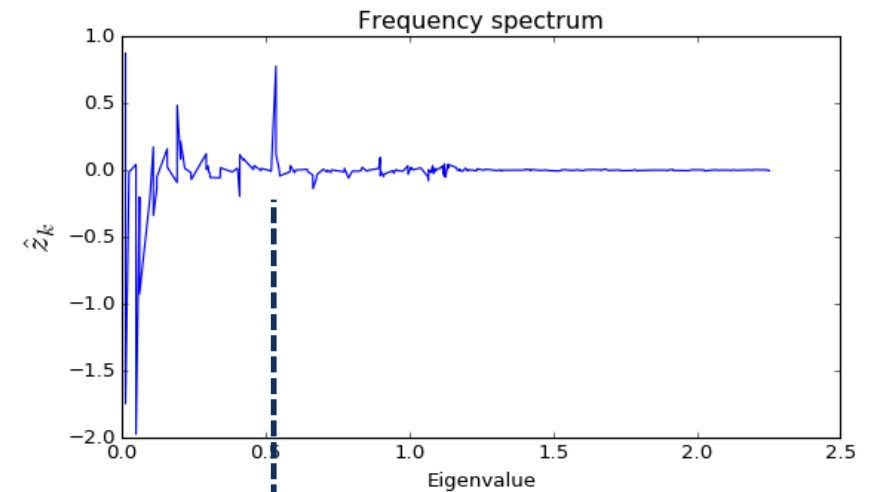
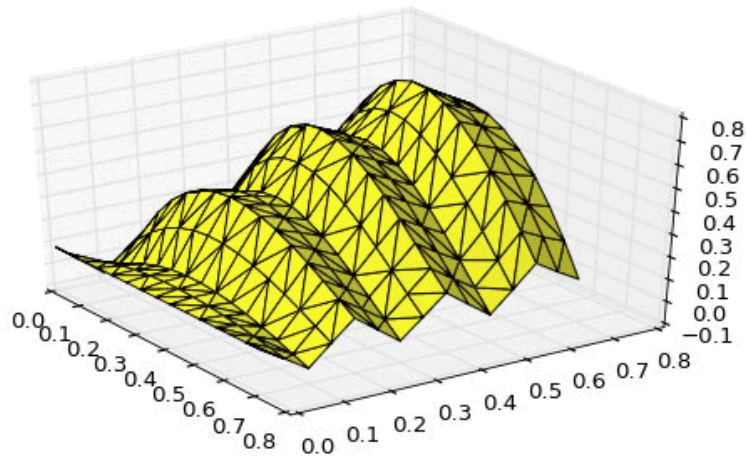
$$\gamma = \sum_{i=1}^N a_i \mathbf{v}_i$$

3. Choose  $f \rightarrow 0$  near any  $\lambda_m$  for which  $a_m$  is large.

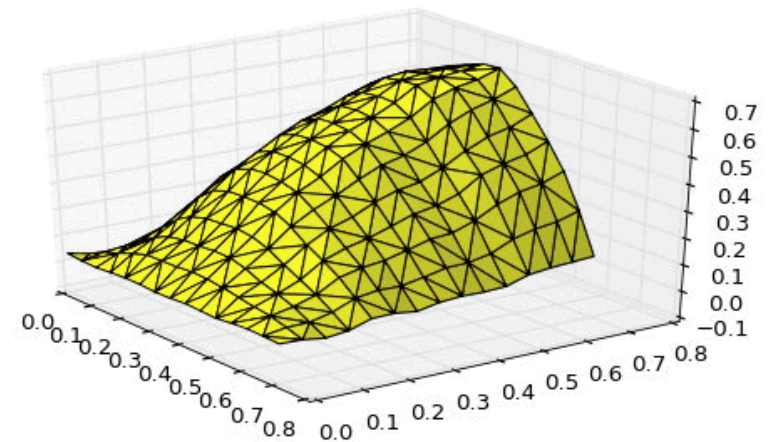
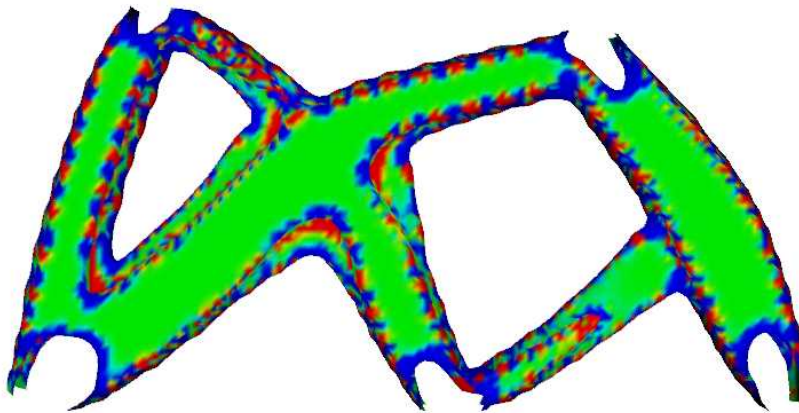
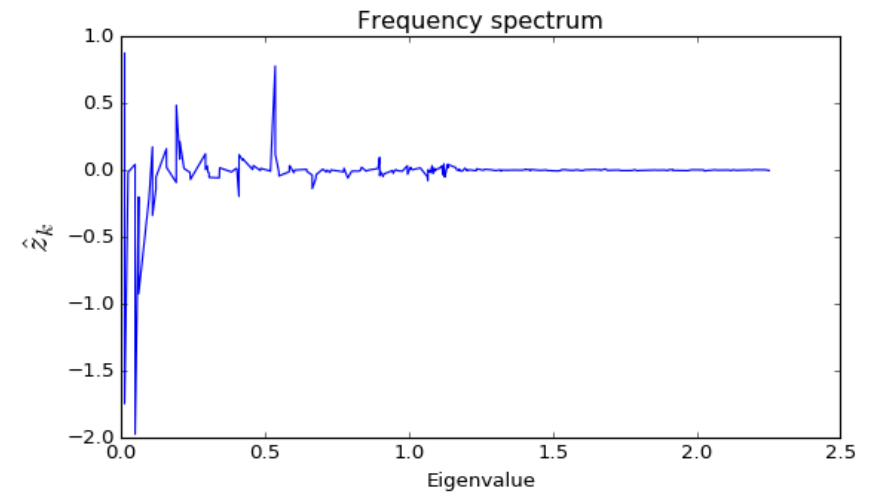
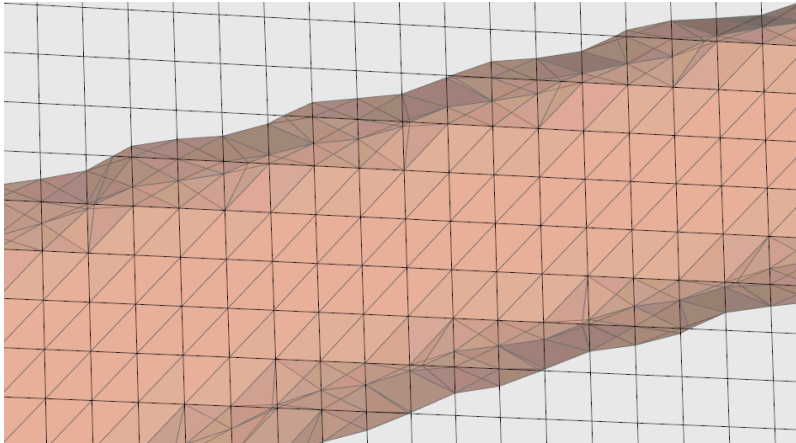
4. Filter shape with  $f$ .

# Mesh spectra are linear

$$\mathbf{x}^{\text{smooth}} + \mathbf{x}^{\text{ripples}} = \sum_{i=1}^N (c_i^{\text{smooth}} + c_i^{\text{ripples}}) \mathbf{v}_i$$



# Summary







Thank You!