

# JAX FDM

A differentiable framework for constrained form-finding

Rafael Pastrana, Sigrid Adriaenssens

[https://github.com/arpastrana/jax\\_fdm](https://github.com/arpastrana/jax_fdm)



Princeton University  
School of Architecture





Schlaich Bergermann, et al. (2018). Trumpf pedestrian footbridge. Ditzingen, Germany



Skidmore, Owens and Merrill, et al. (2021). The roof of the Moynihan train hall. New York City, New York

# Form-finding is shape optimization

A form-finding method computes a shape in static equilibrium



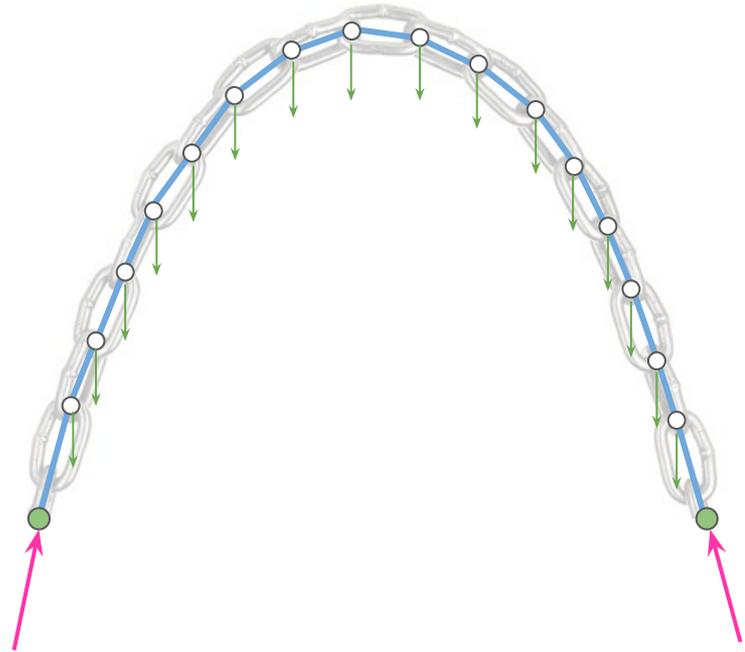
Tension-only



Compression-only

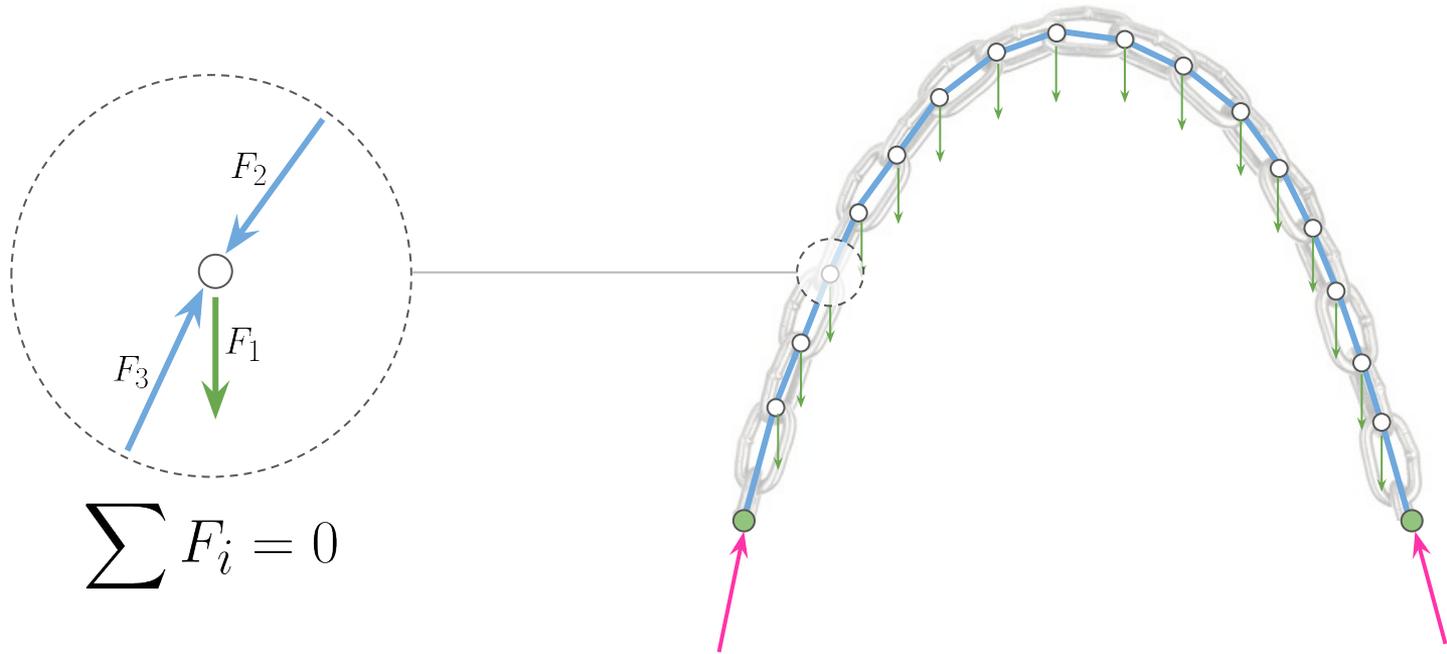
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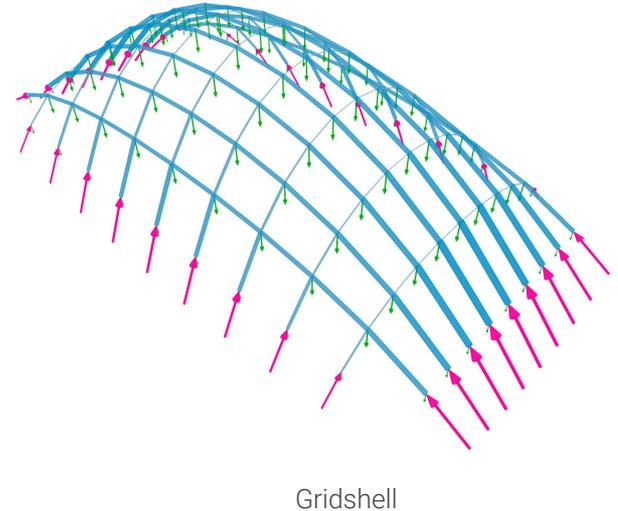
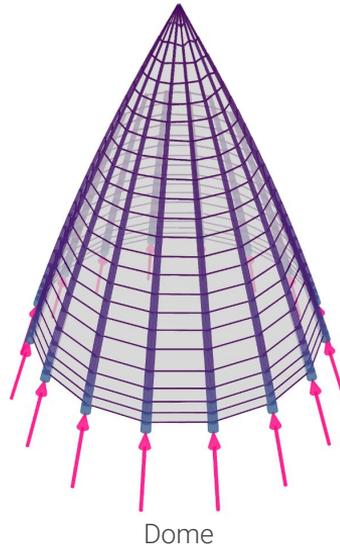
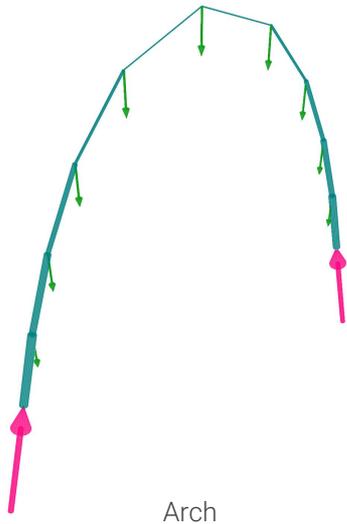
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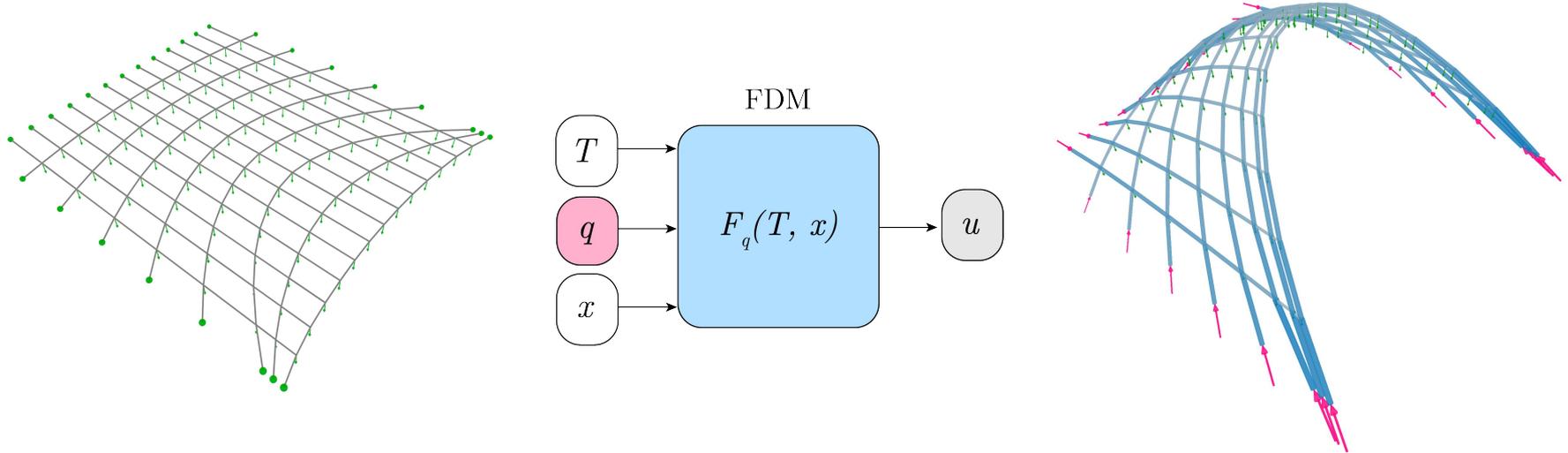
# The force density method (FDM)

A numerical form-finding method for vaults and cable-nets



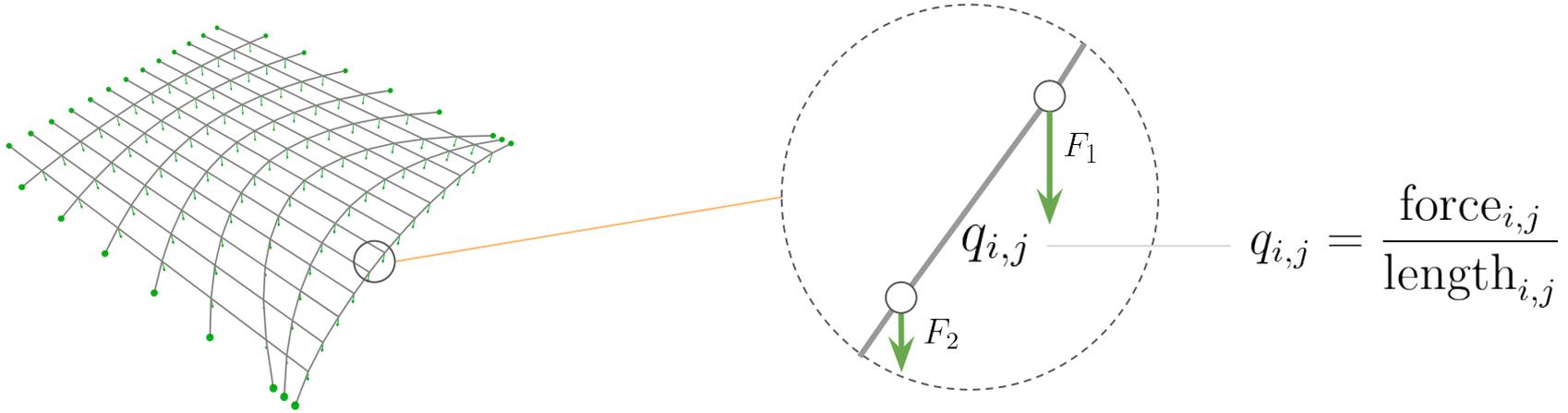
# The force density method (FDM)

How does it work? Forward



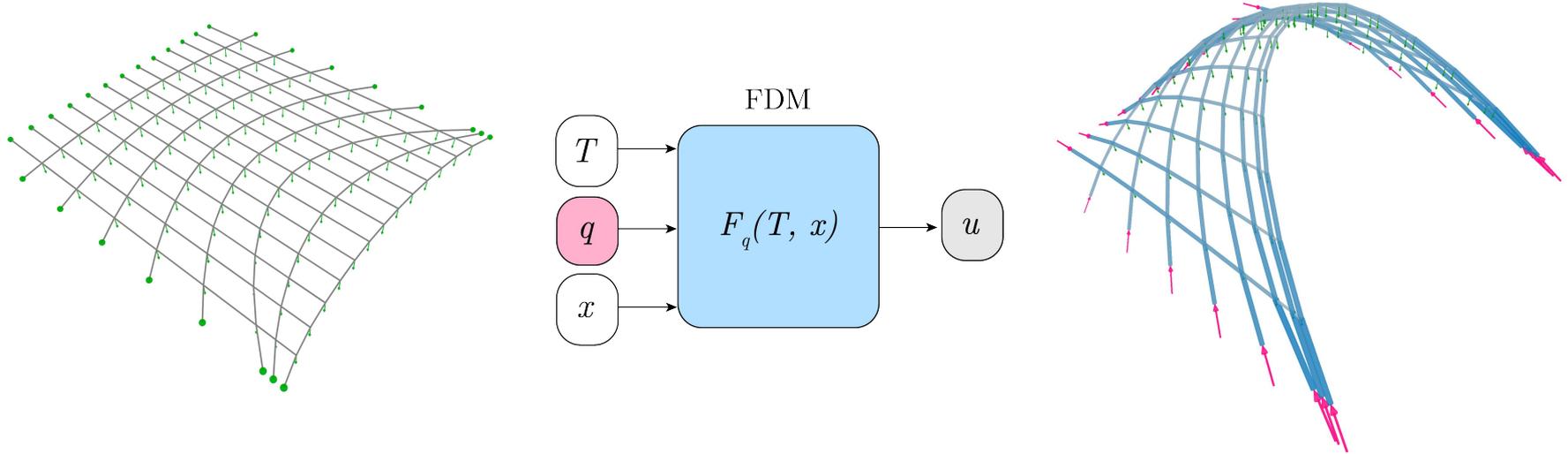
# The force density method (FDM)

How does it work? Define the force density of the bars



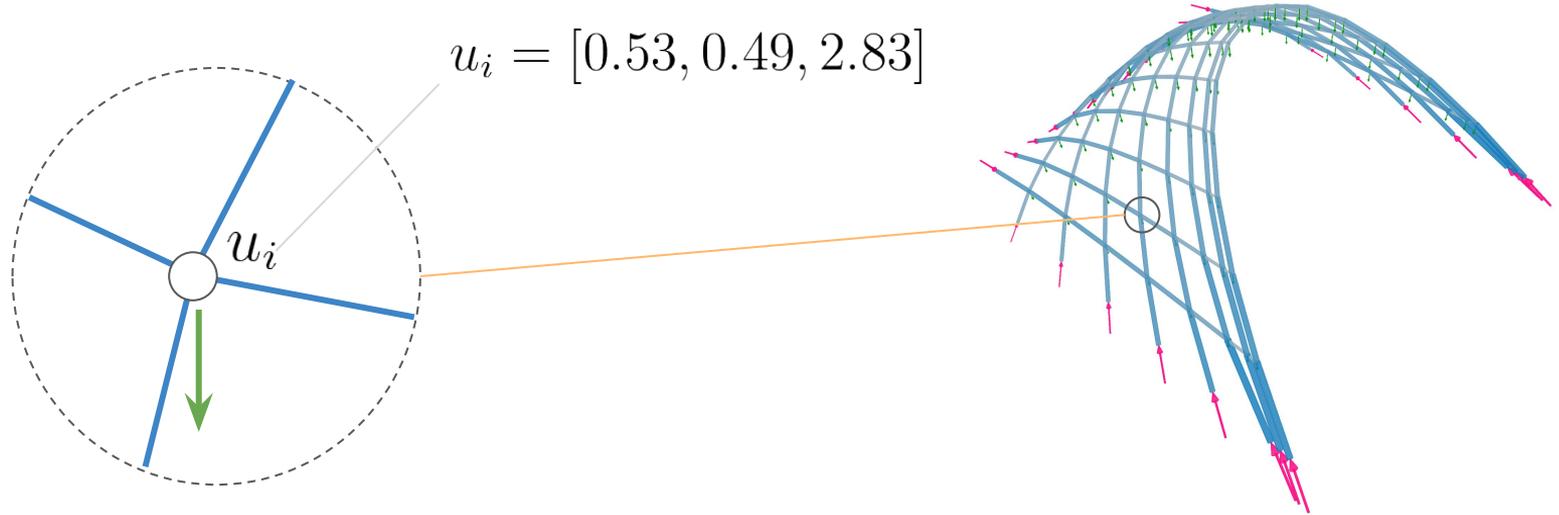
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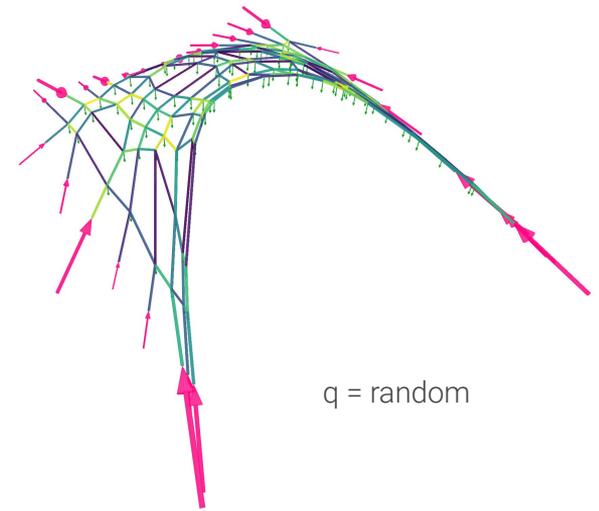
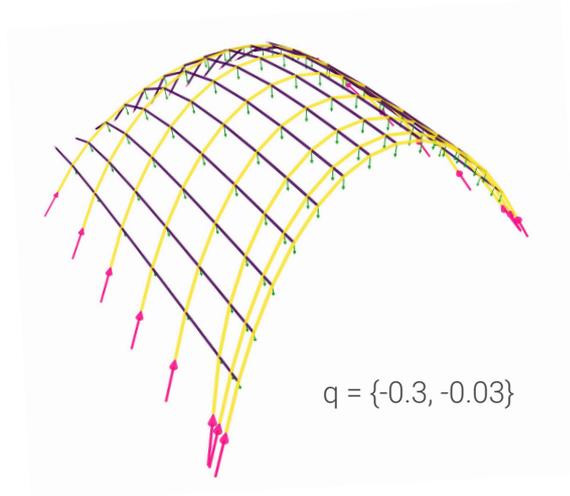
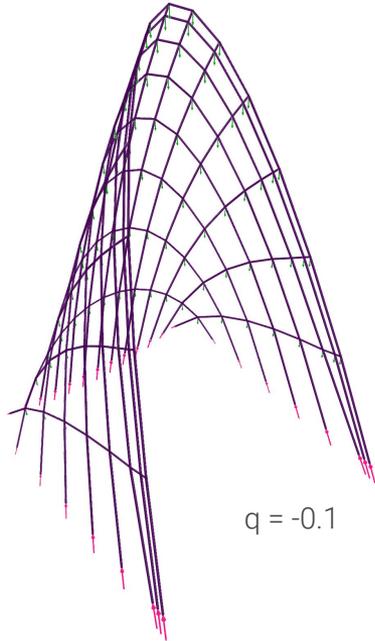
# The force density method (FDM)

How does it work? Get the XYZ coordinates of the nodes



# The force density method (FDM)

A plethora of shapes in static equilibrium for different force densities



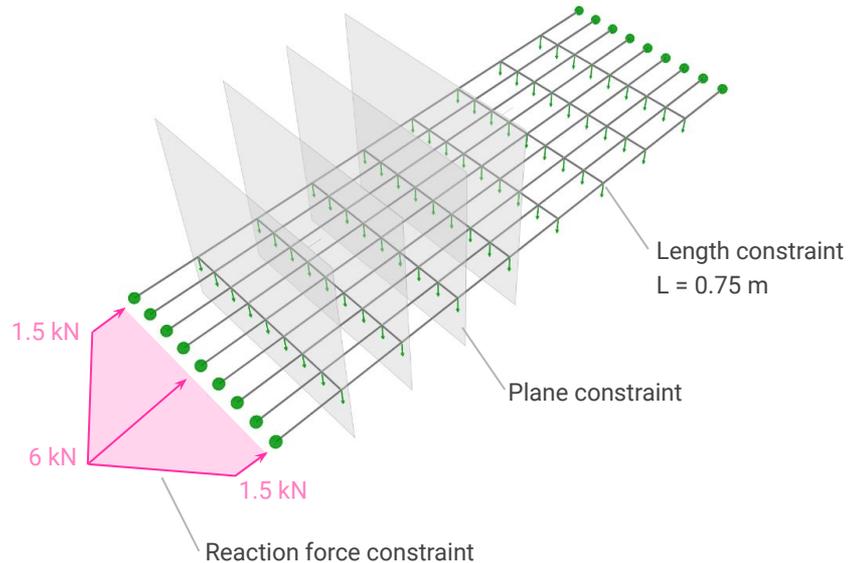
# The challenges of form-finding

Viable shapes in static equilibrium meet additional constraints: fabrication, structural, aesthetic, and environmental.

No two structures are (or want to be) alike.

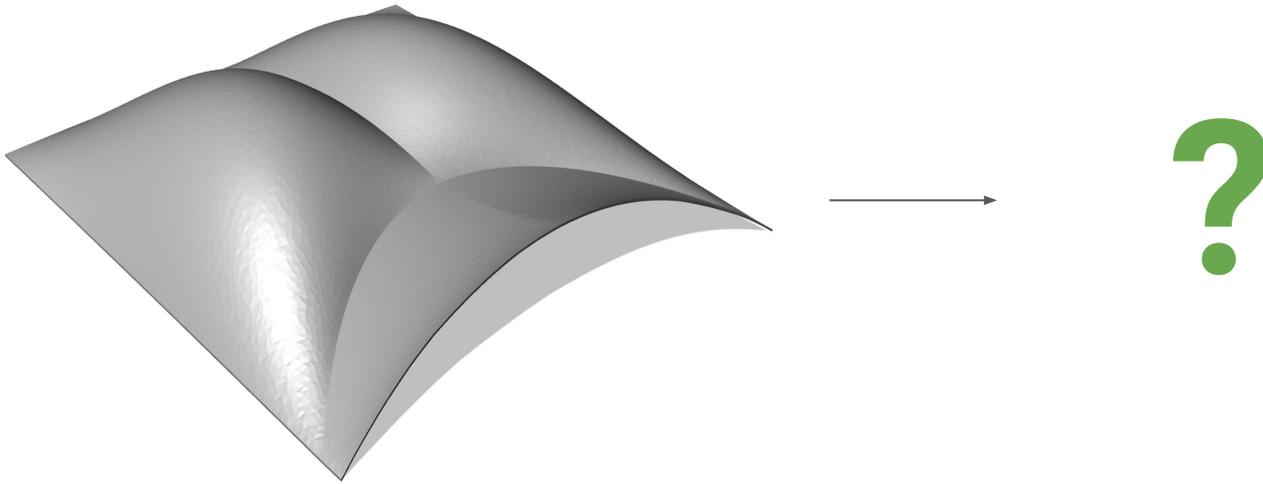
# The challenges of form-finding

Shapes in equilibrium conditioned on fabrication constraints



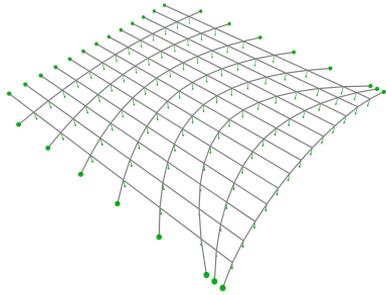
# The challenges of form-finding

Capturing architectural design intent

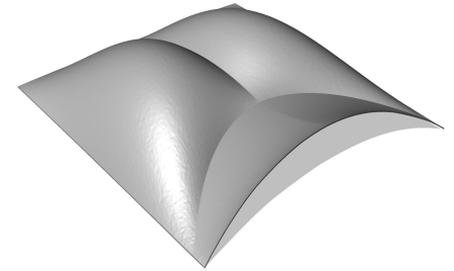
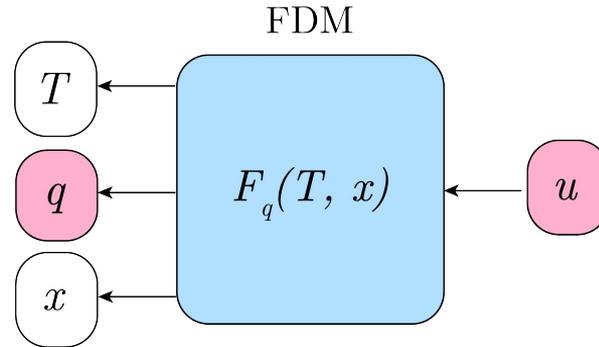


# Constrained form-finding is the challenge

We actually want to solve an inverse problem



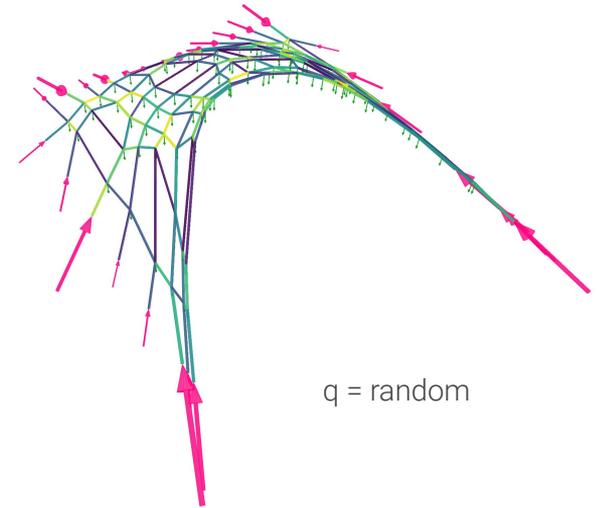
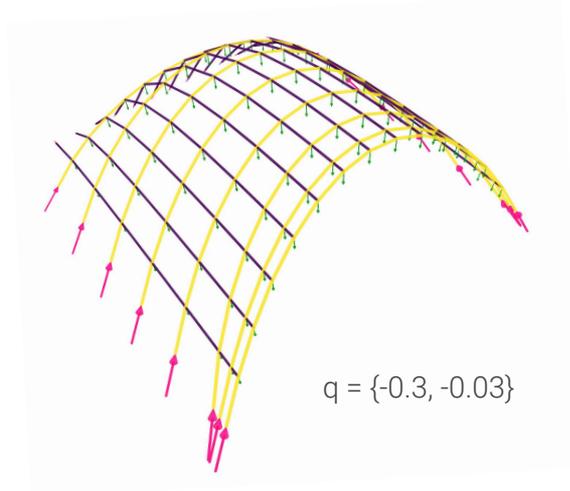
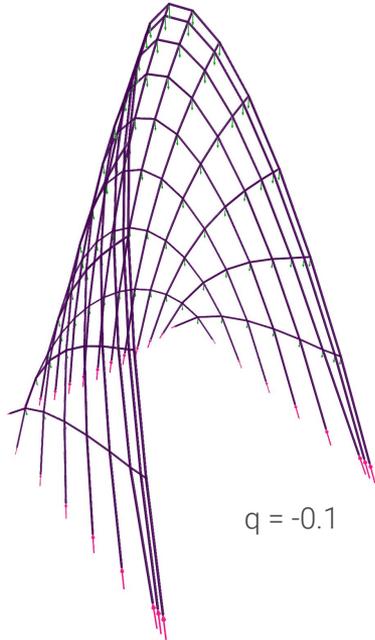
*“What is the set of force densities  $q$  that are best conducive to it?”*



*“A shape in static equilibrium which approximates this other shape”*

# Constrained form-finding by hand?

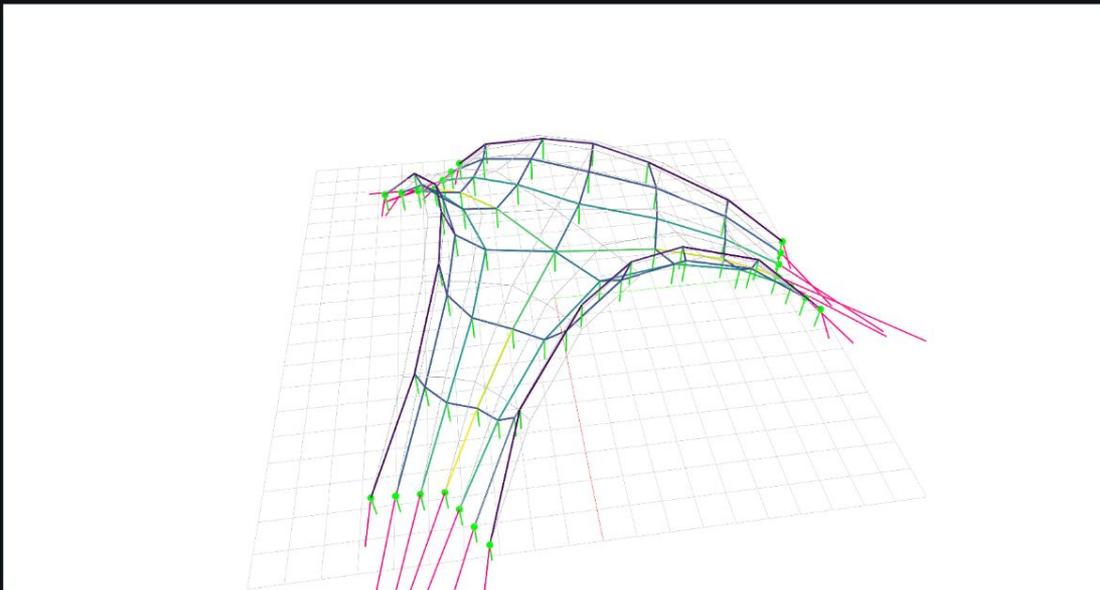
Constrained form-finding by hand is laborious and error prone



# JAX FDM

A differentiable, hardware-accelerated framework for constrained form-finding in structural design.

Crafted with care in the [Form-Finding Lab](#) at [Princeton University](#) ❤️🇺🇸

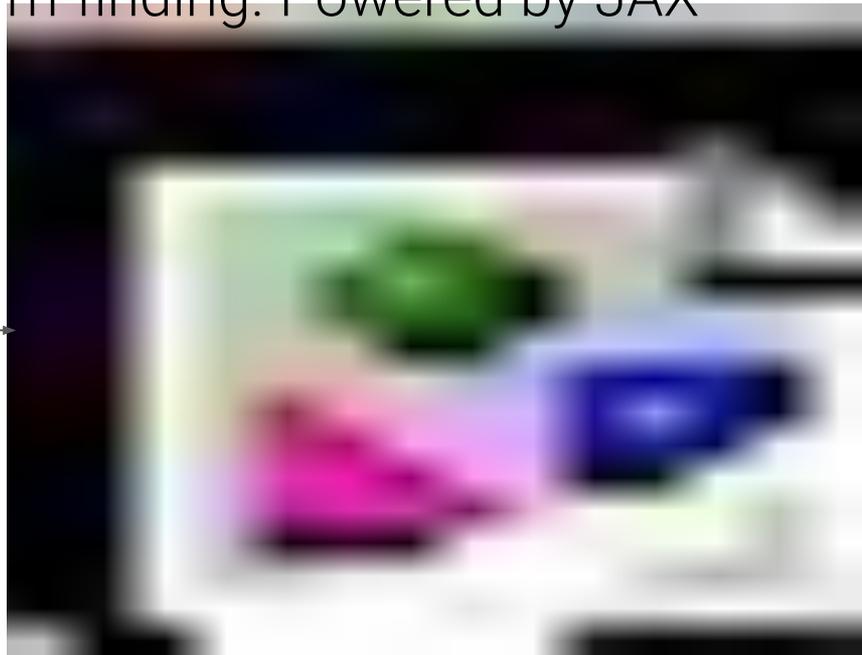
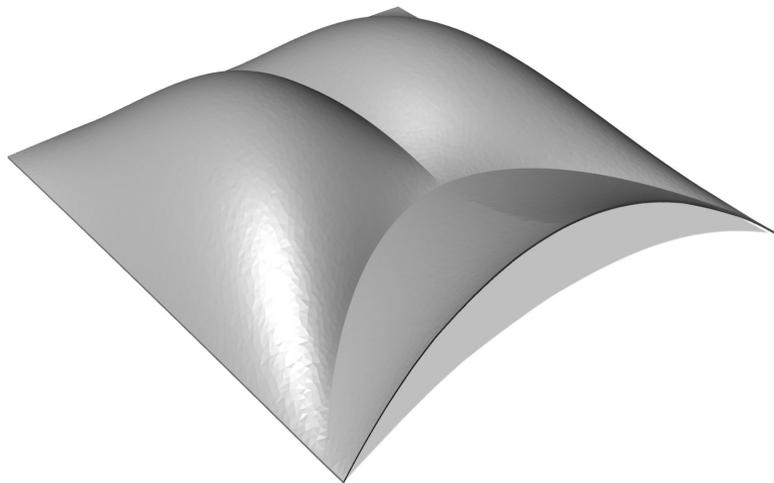


JAX FDM enables the solution of inverse form-finding problems for discrete force networks using the force density method (FDM) and gradient-based optimization. It streamlines the integration of form-finding simulations into deep learning models for machine learning research.

## Key features

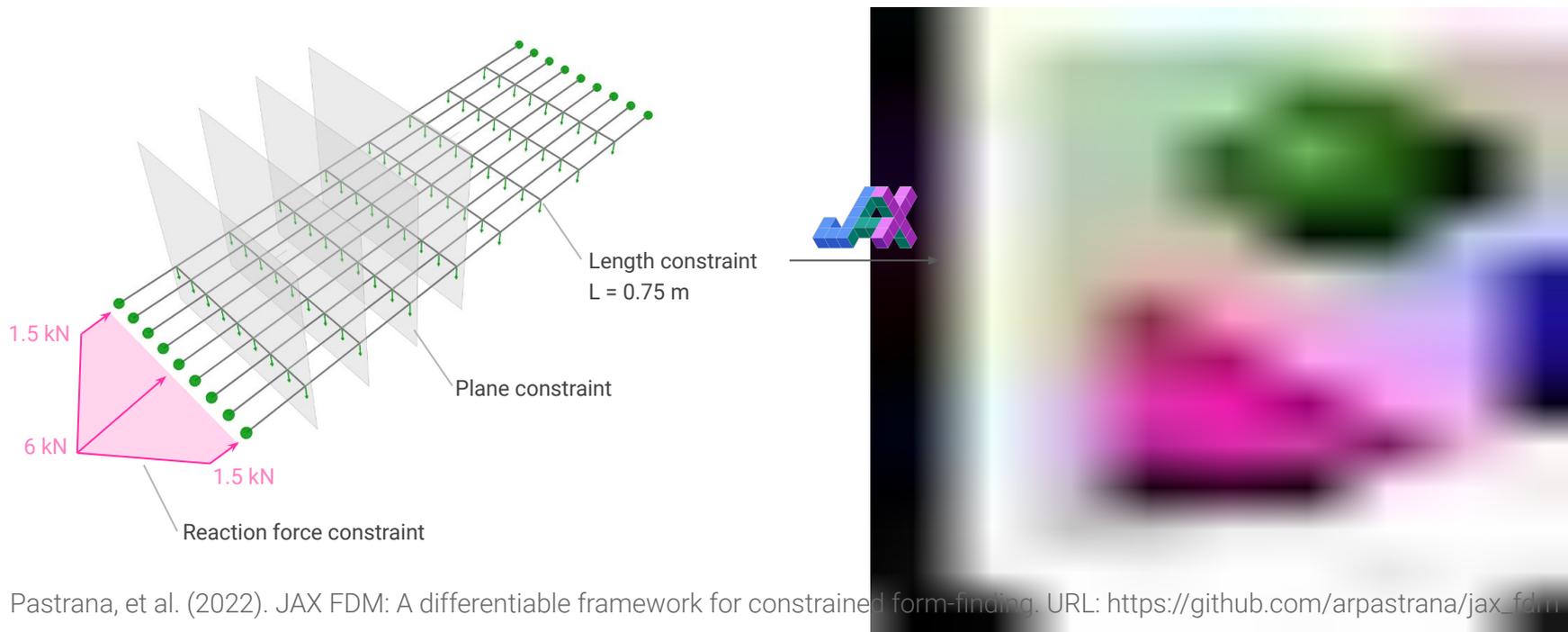
# JAX FDM A solution

A differentiable tool for constrained form-finding. Powered by JAX



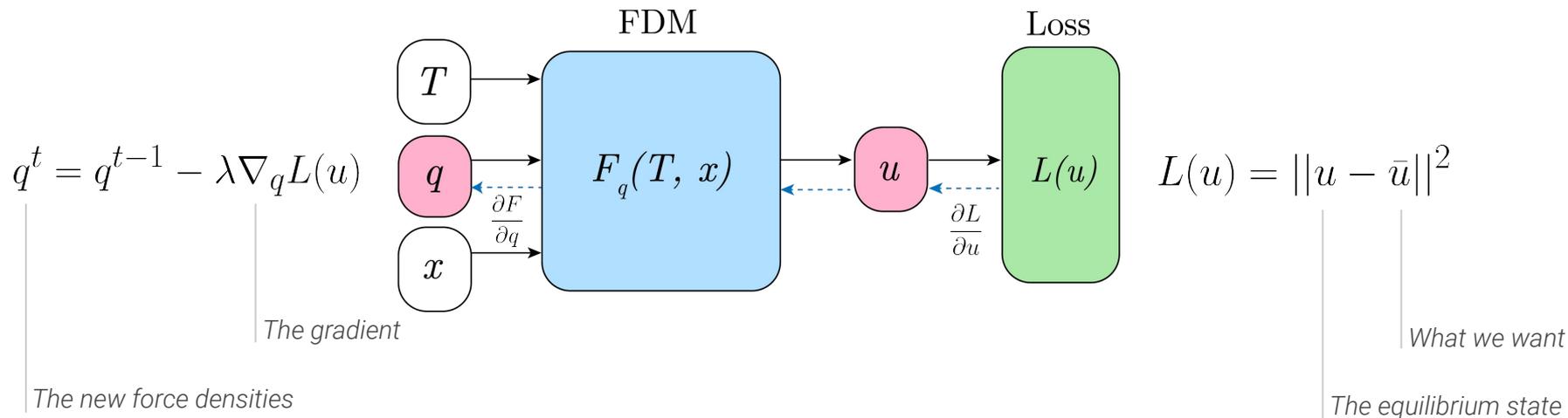
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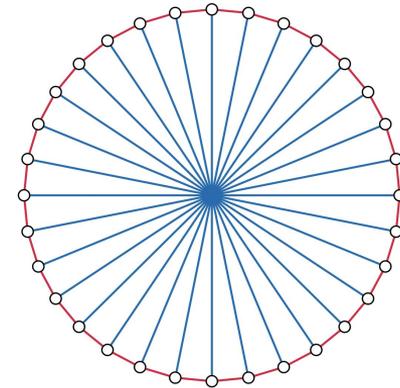
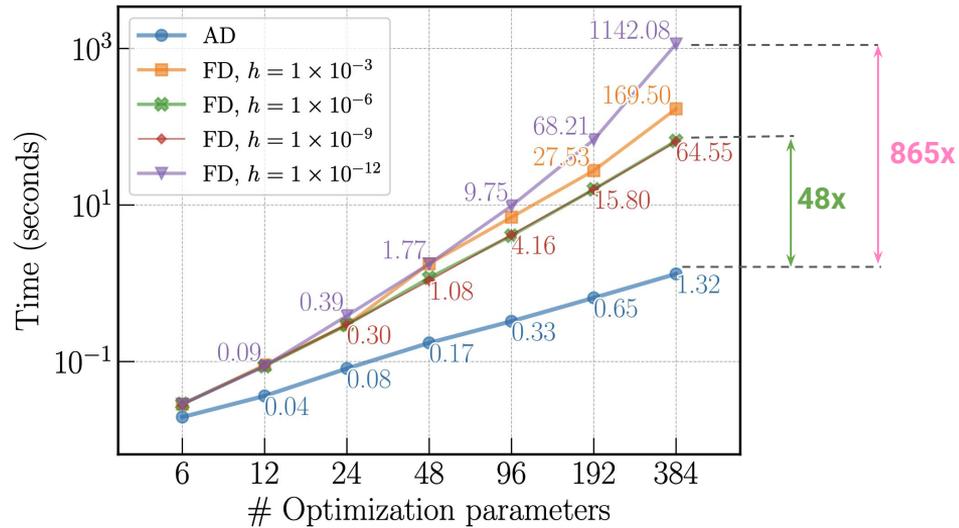
# JAX FDM Tackling constrained form-finding

We apply backpropagation for the inverse design of 3D structures



# Constrained form-finding Backpropagation

Inverse design with backpropagation is faster and more stable



48 parameters

```
from jax import grad
import jax.numpy as jnp
from jax_fdm.datastructures import FDNetwork
from jax_fdm.equilibrium import EquilibriumModel

# create the FDM model
pattern, supports, loads = FDNetwork.from_json('arch.json')
model = EquilibriumModel(pattern, supports, loads)

# define the loss function
def loss(q, target_length=1.5):
    eq_state = model(q)
    return jnp.mean((eq_state.edge_lengths - target_length) ** 2)

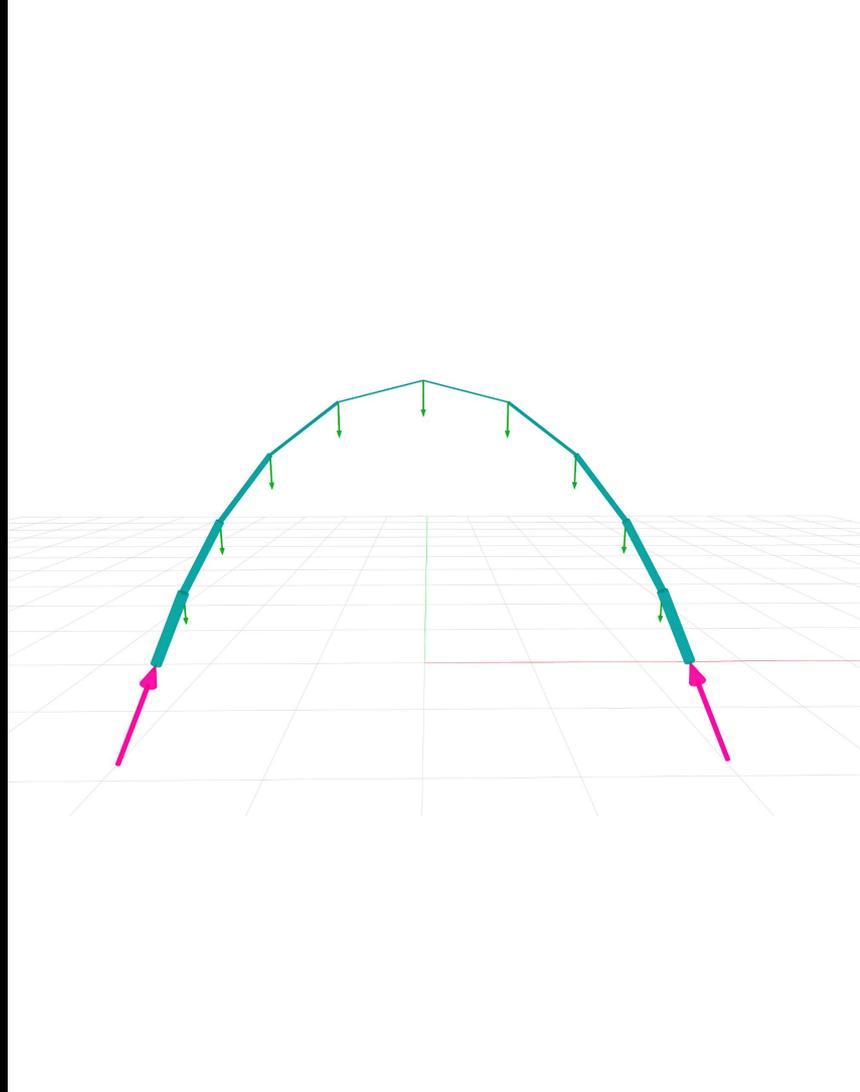
# vanilla gradient descent
q = jnp.ones(10) * -1.0 # initial guess
lr = 0.1 # step size
for i in range(1000):
    loss_value = loss(q)
    q = q - lr * grad(loss)(q)
```

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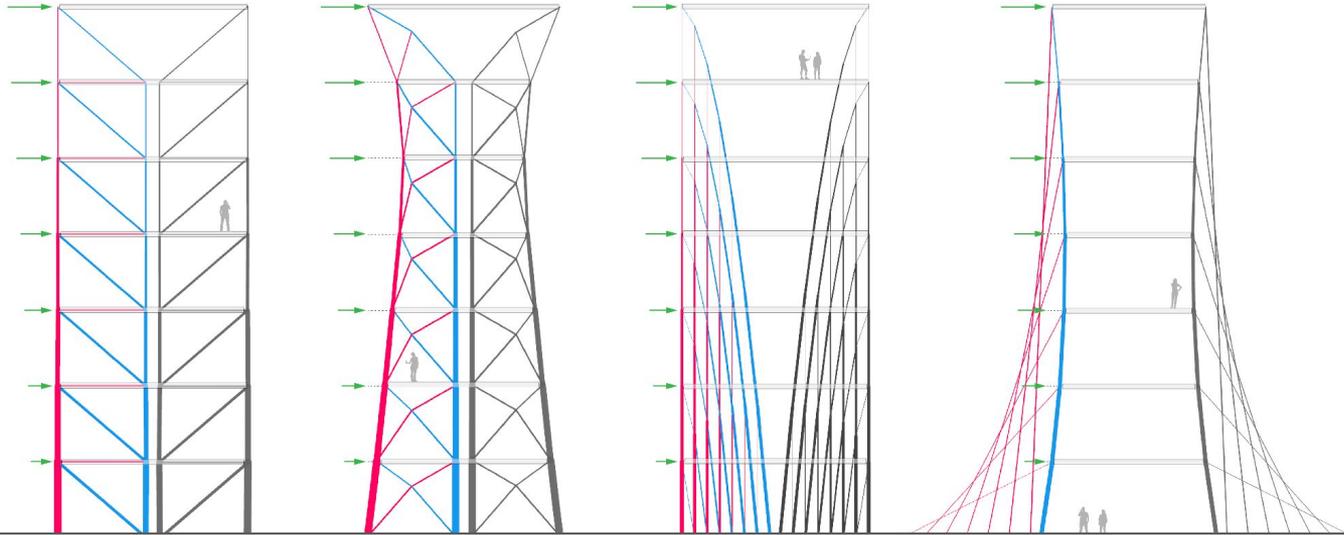
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# Constrained form-finding of tall buildings

Leveraging static equilibrium to design new vertical structures



# Constrained form-finding to reduce waste

Vaults that require minimal scaffolding during construction



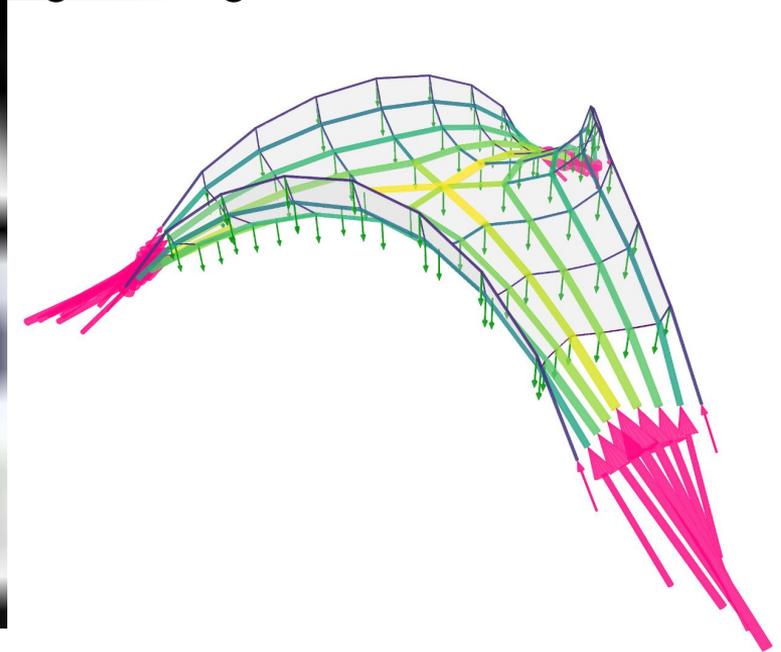
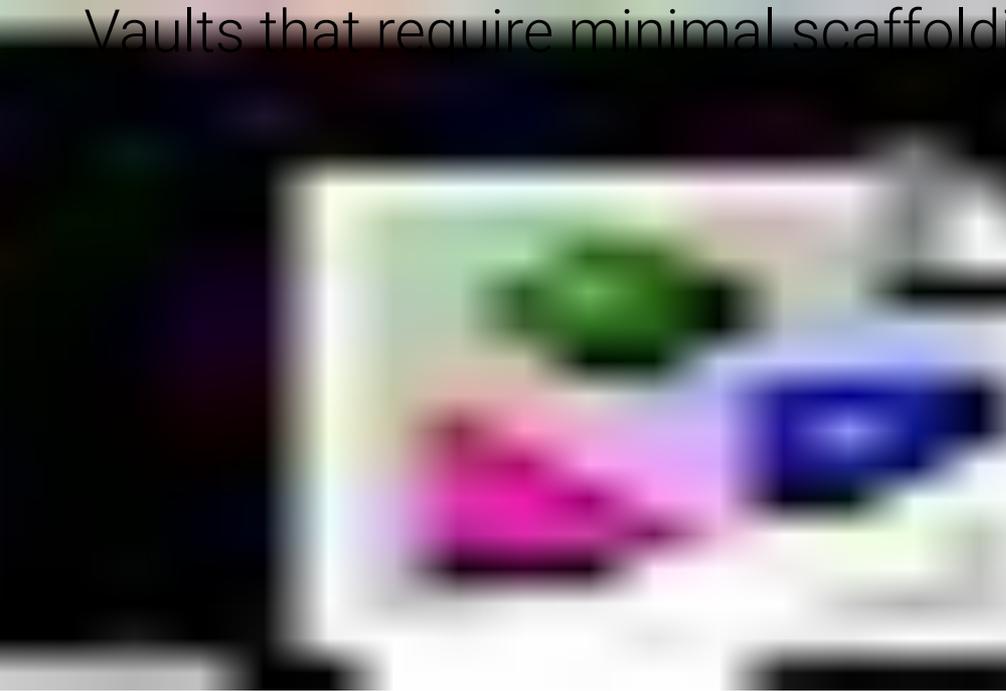
Mexican architect



Mexican builder

# Constrained form-finding to reduce waste

Vaults that require minimal scaffolding during construction



# JAX FDM What's next?

Combining machine learning and differentiable form-finding

## Learning to solve the inverse problem

JAX FDM as a layer in a neural network. 10x speed-up?

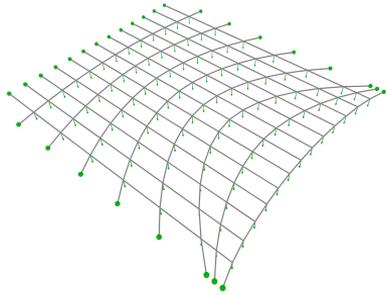
## Learning continuous representations of graphs

Can we optimize patterns and force densities jointly?

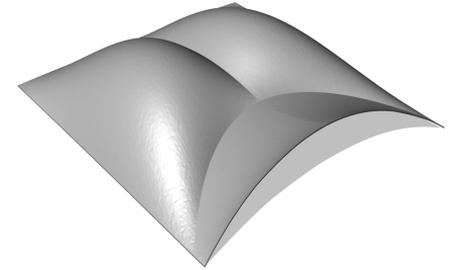
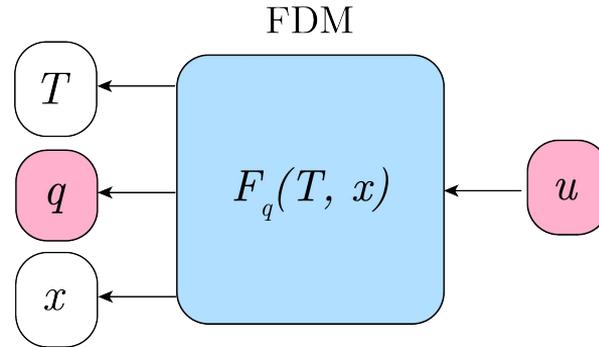


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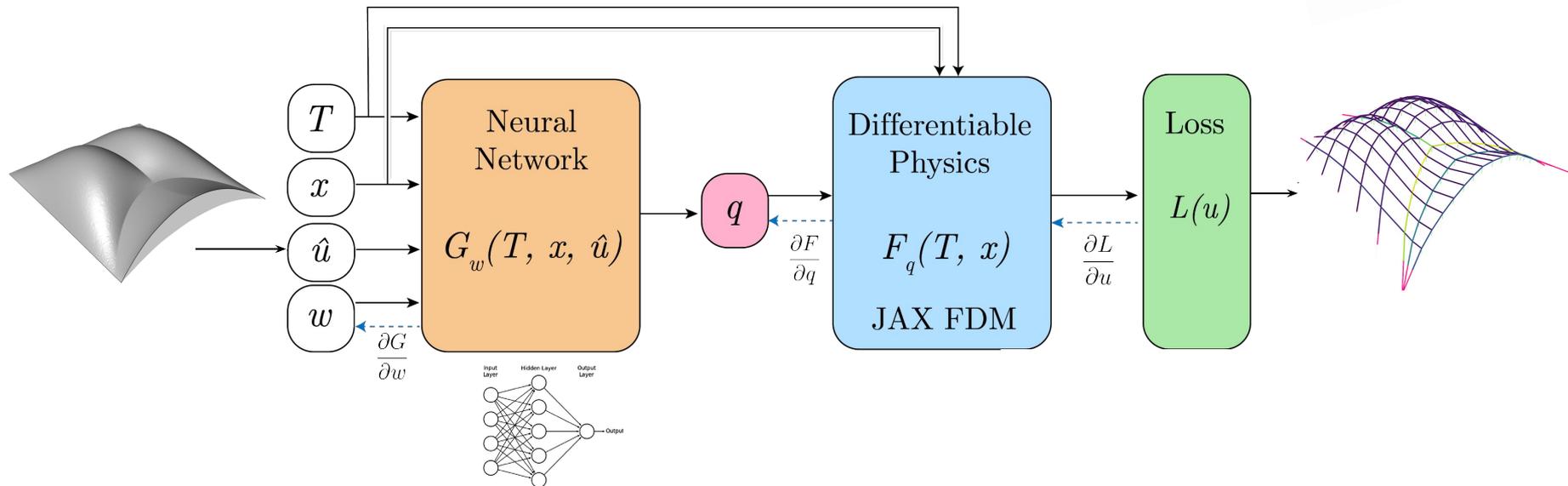
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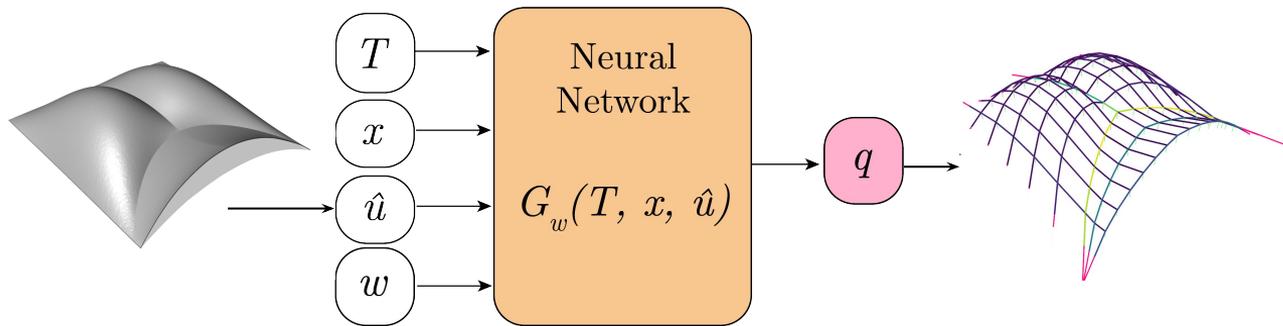
# Next steps Learning to solve the inverse problem

Differentiable form-finding as a layer in a neural network



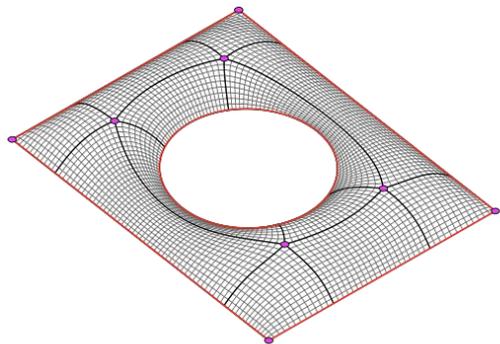
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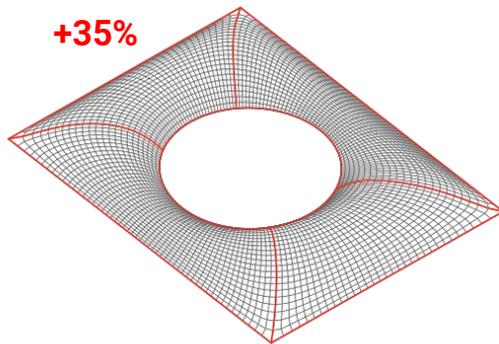


10x speed-up?

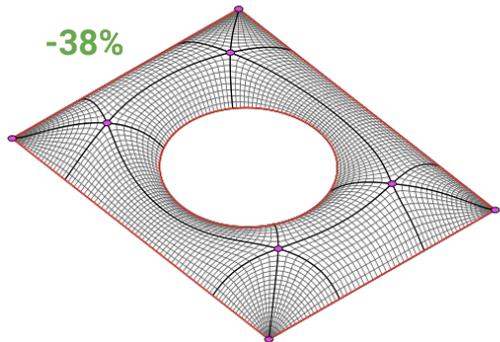
# Next steps Generative model of patterns



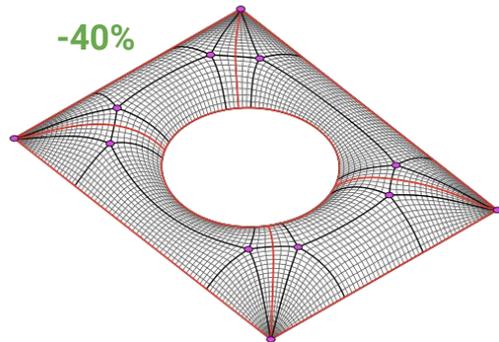
+35%



-38%



-40%



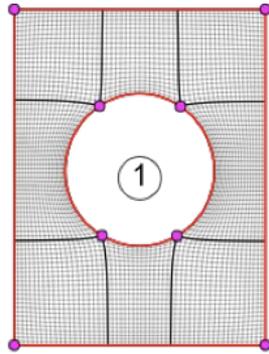
Oval (2019). Topology finding of patterns for structural design. Ph.D. thesis



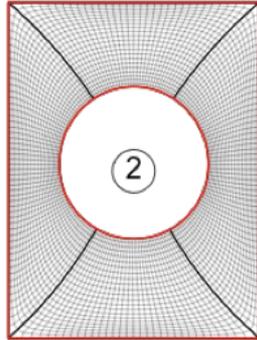
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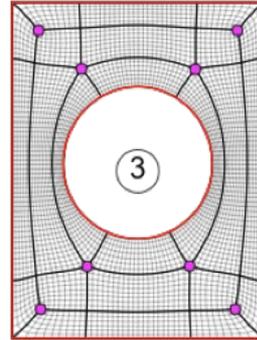
Learning continuous representations of graphs  
Cast as a **NLP** task?



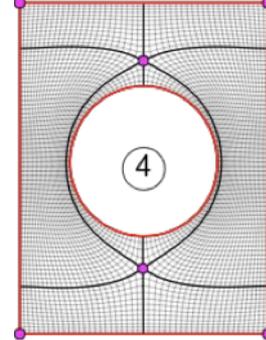
ATPA°TAPA°



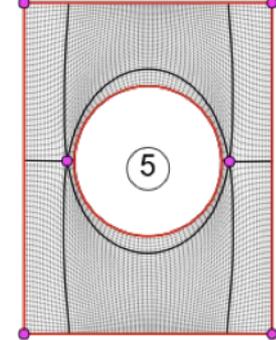
ATTPDA°



ATA°ATAPA°ATDPA°



APA°APA°TD

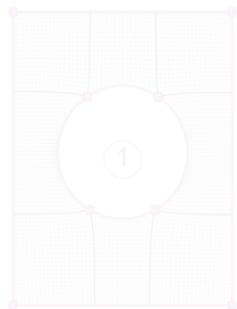


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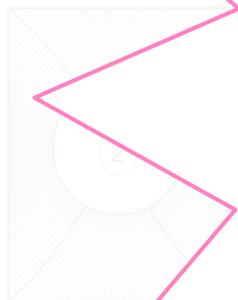
# Next steps Generative model of patterns



Learning continuous representations

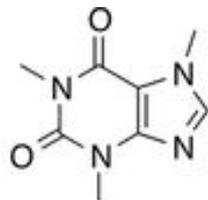


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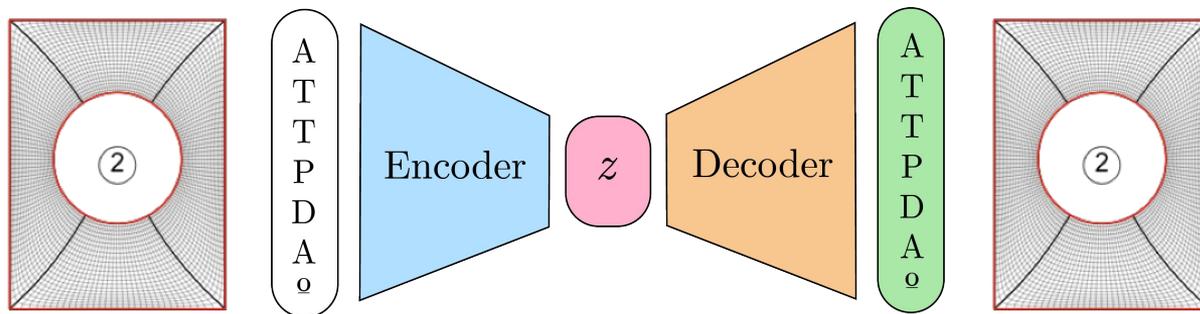
Caffeine

CN1c2ncn(C)c2C(=O)N(C)C1=O

# Next steps Generative model of patterns

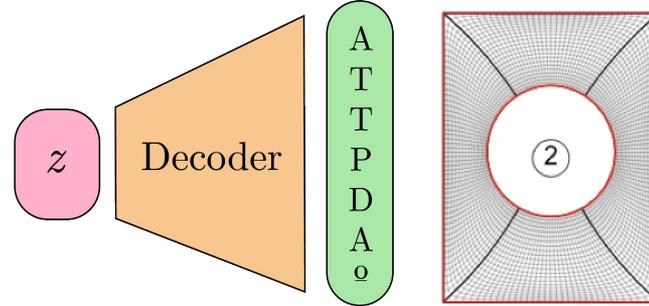


Learning continuous representations of graphs  
Cast as a **NLP** task?



# Next steps Generative model of patterns

Learning continuous representations of graphs  
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