

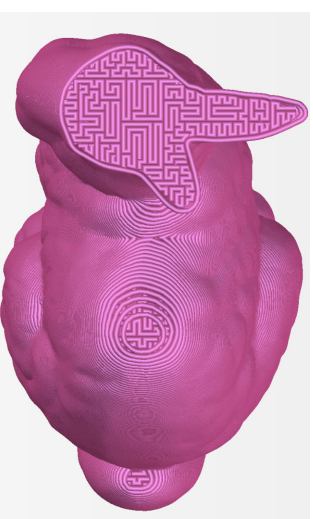
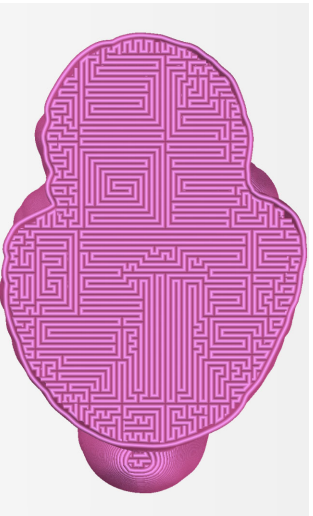
SFCD DECOMP: MULTICRITERIA OPTIMIZED TOOLPATH PLANNING IN 3D PRINTING

*Prashant Gupta, Yiran Guo, Narasimha Boddeti,
Bala Krishnamoorthy*

Washington State University

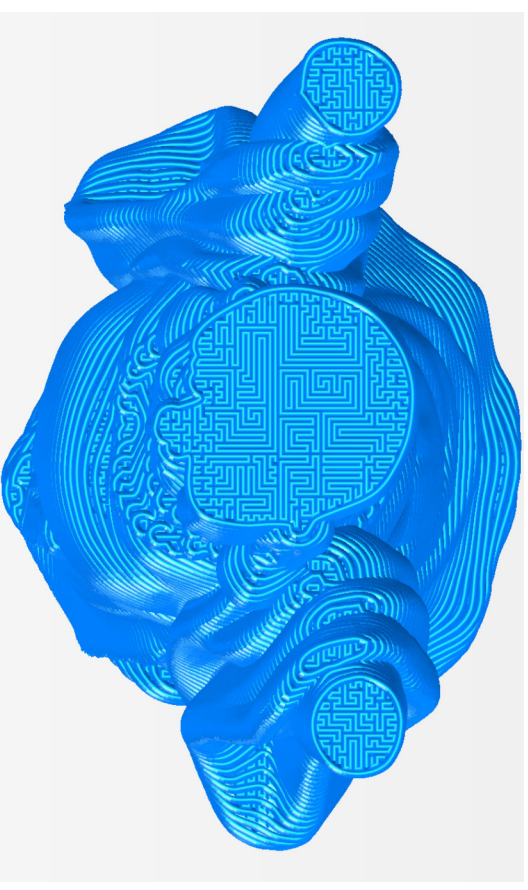
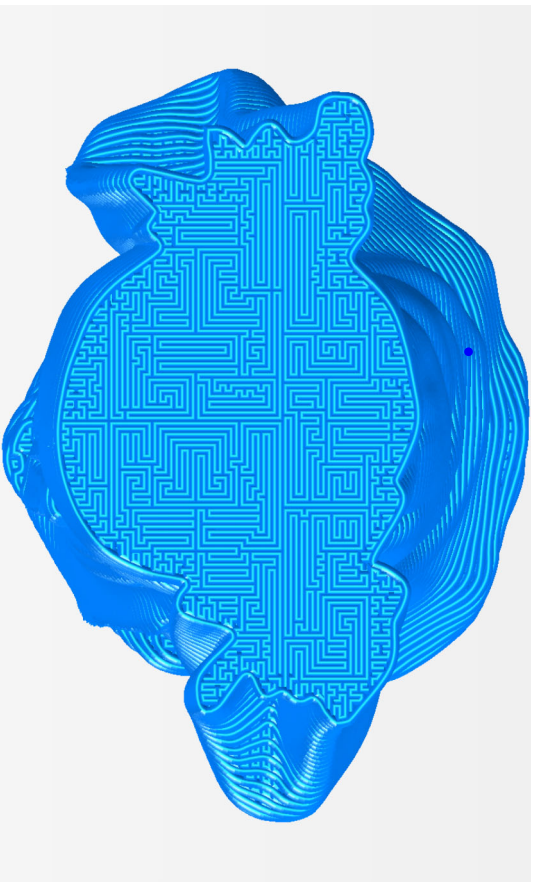
ISCGA

arXiv: 2109.01769



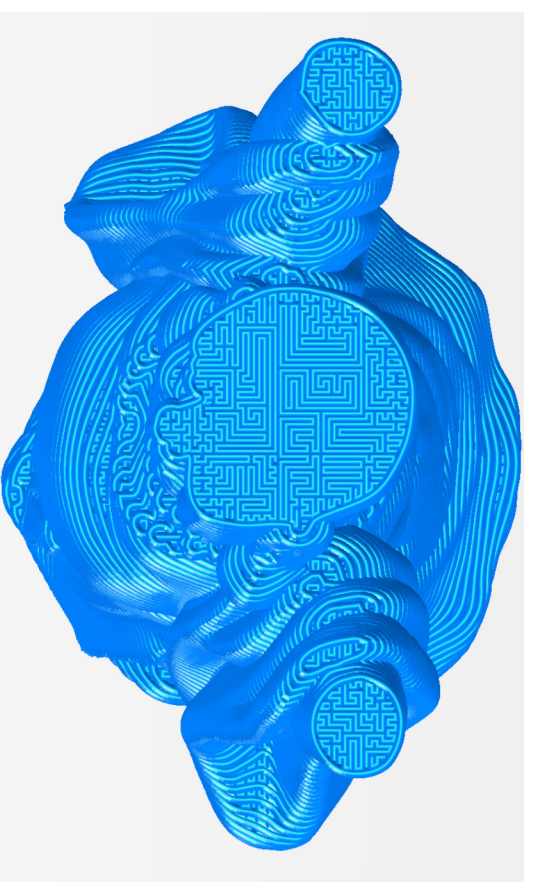
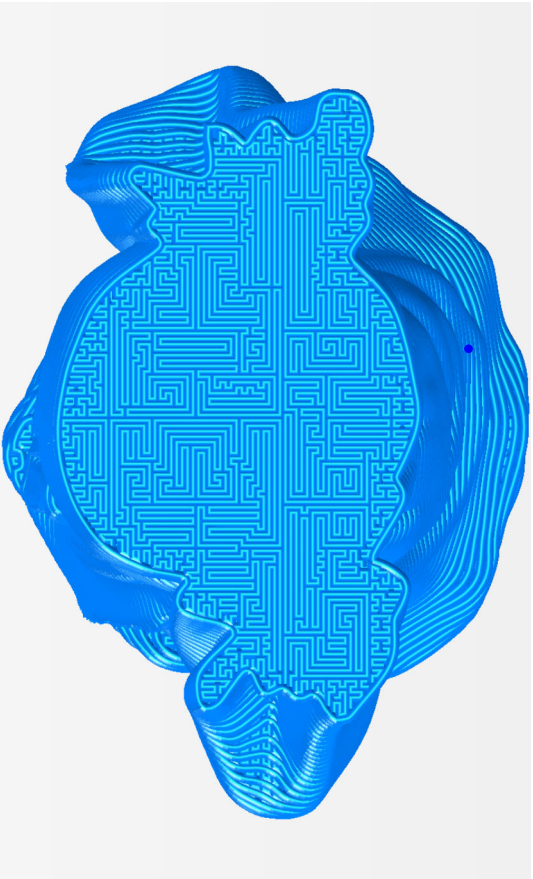
3D PRINTING: TOOL PATH OPTIM.

* layer-by-layer, dense infill 3D printing



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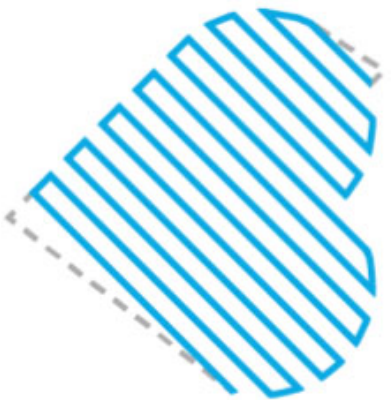
* layer-by-layer, dense infill 3D printing



- * optimize sequence of extruder movement (tool path) for
 - continuity
 - turn costs
 - smoothness
 - local orientation
 - ...

TOOL PATH GEOMETRY

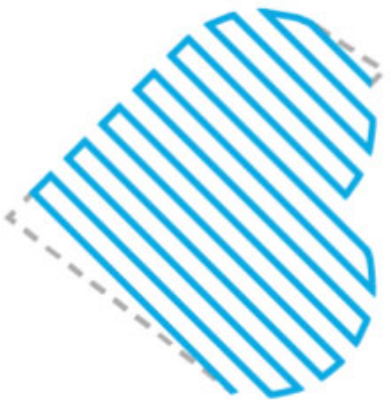
* Standard: zigzag, contour parallel



img.fornizable.com

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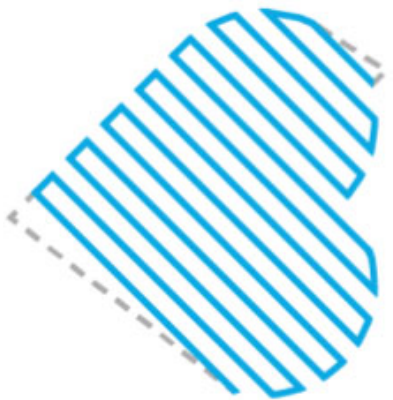


img: formizable.com

X suffer from directional biases

TOOL PATH GEOMETRY

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img: formizable.com

- X suffer from directional biases
- X cannot crossweave easily across layers

TOOL PATH GEOMETRY

- * Zhao et al. (2016): Fermat spiral
 - smooth tool path optimized for continuity
 - curved path approximated by PL line segments in practice



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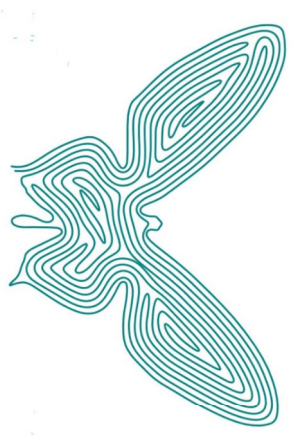
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- X hard to crossweave across layers
 - X cannot apply to subdomains



QUESTIONS

? Optimize total paths by decomposing into subdomains?

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- ? Process subdomains in parallel?

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- ? Process subdomains in parallel?
- ? Optimize for multiple criteria?
 - continuity
 - turn costs
 - maximize rectilinear movement
 - edge overlap across layers?

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✓ min turn 3DPP is NP-hard

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- ✓ optimize edge overlap between layers if it
 - mechanical testing.

NP-HARDNESS OF 3DPP

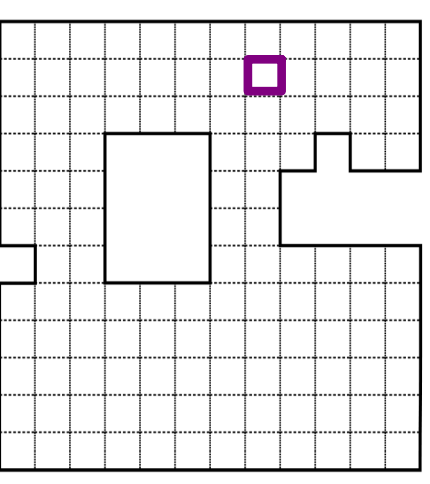
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extruder: unit square

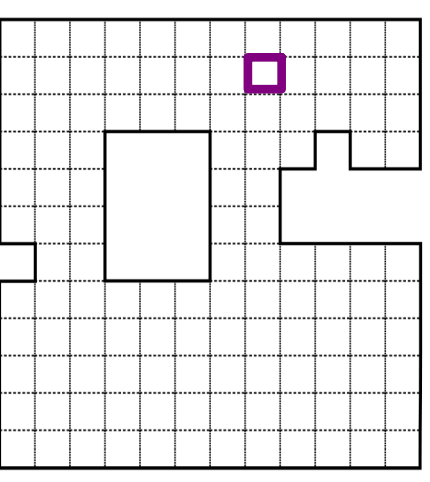


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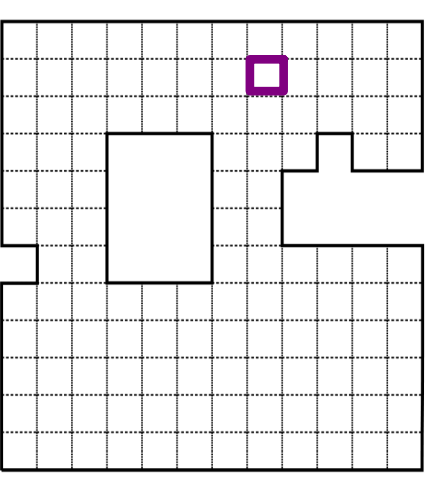
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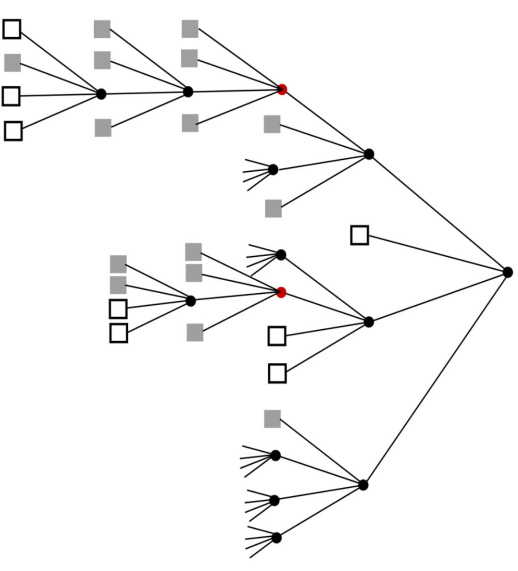
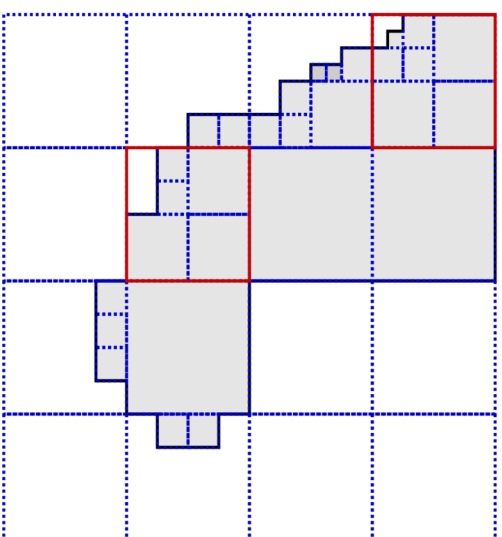
Theorem Min turn cost 3DPP is NP-hard.



SFC DECOMP: DOMAIN DECOMPOSITION

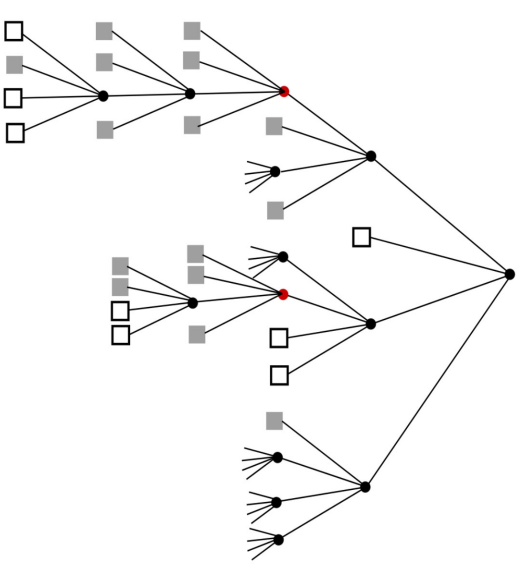
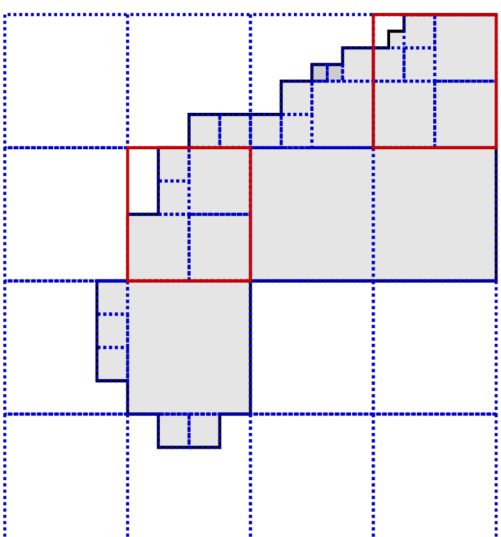
* quadtree representation of print domain

- leaf squares in domain are kept
- solve leaf cells in parallel

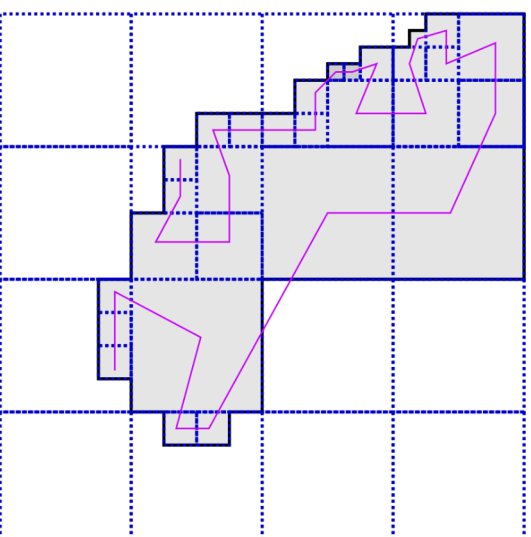


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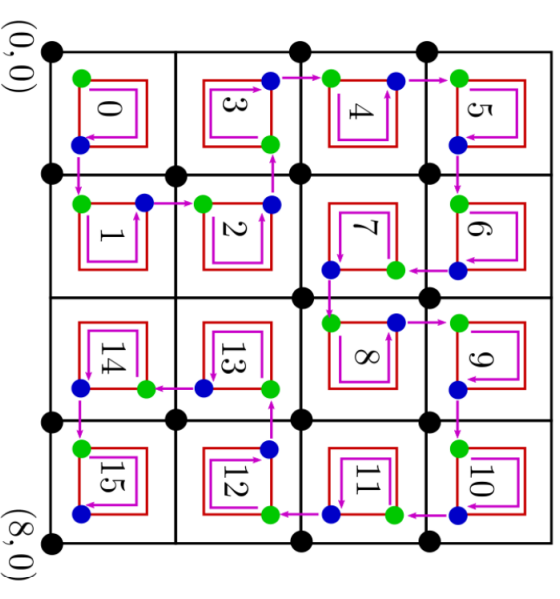
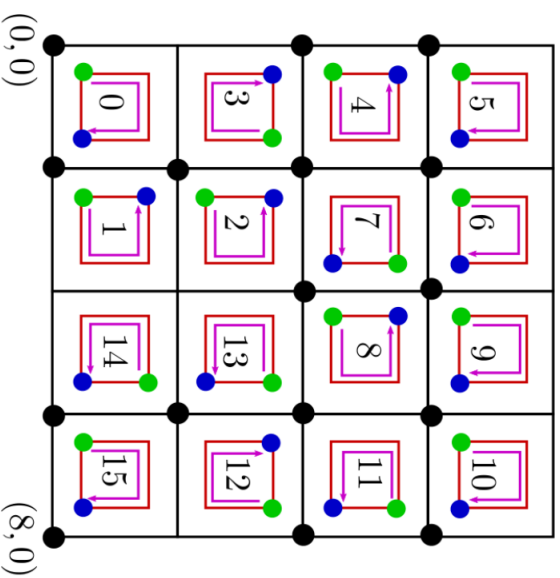
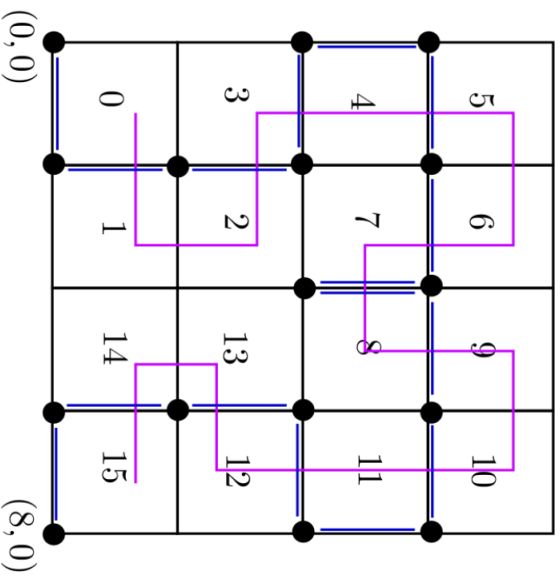


- * connect leaf domains using a Hilbert SFC



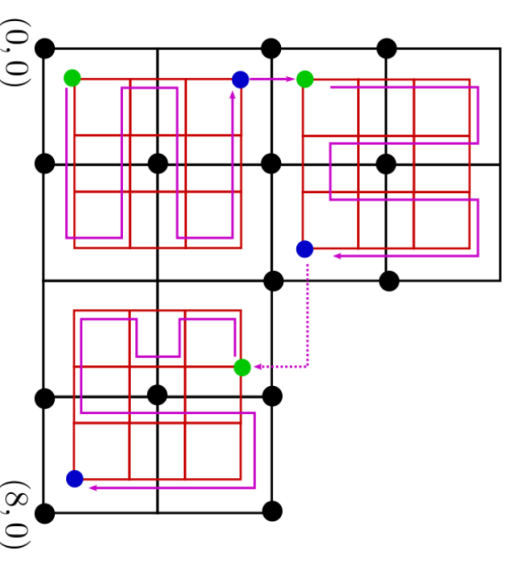
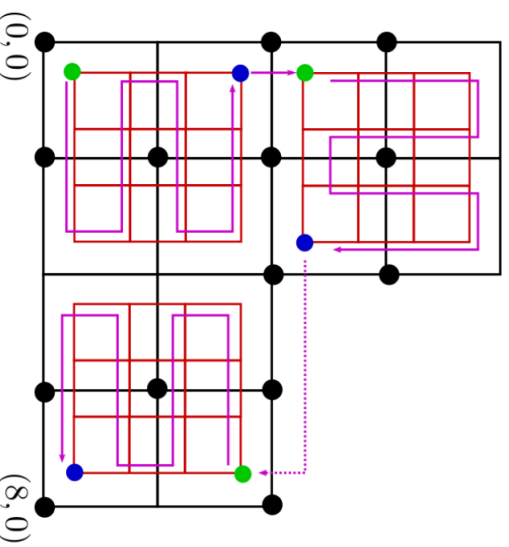
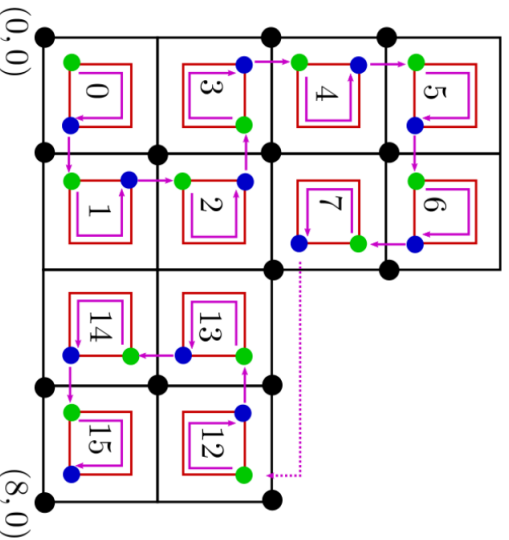
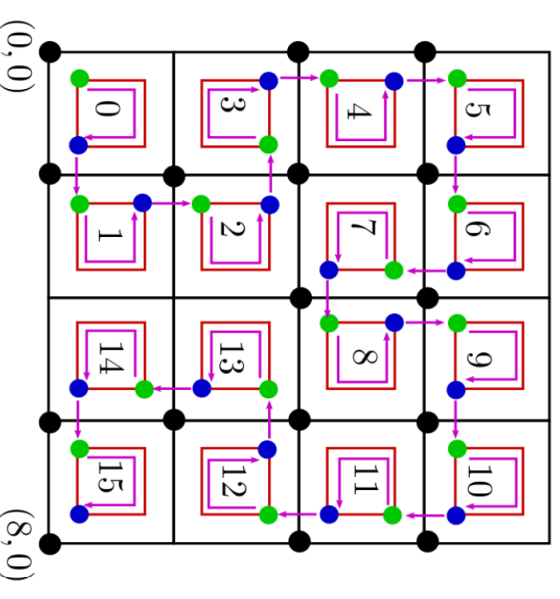
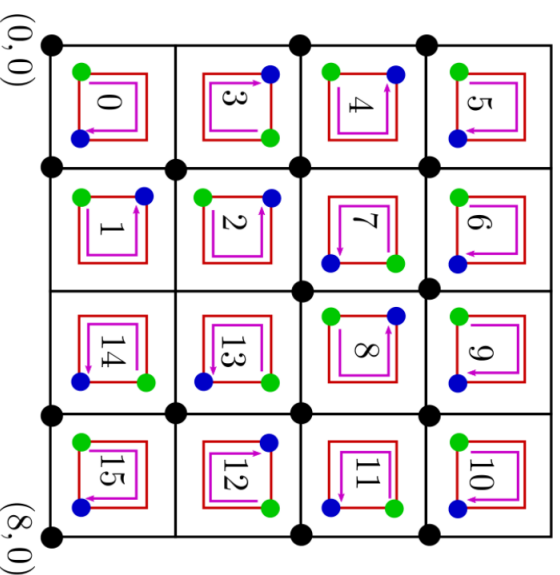
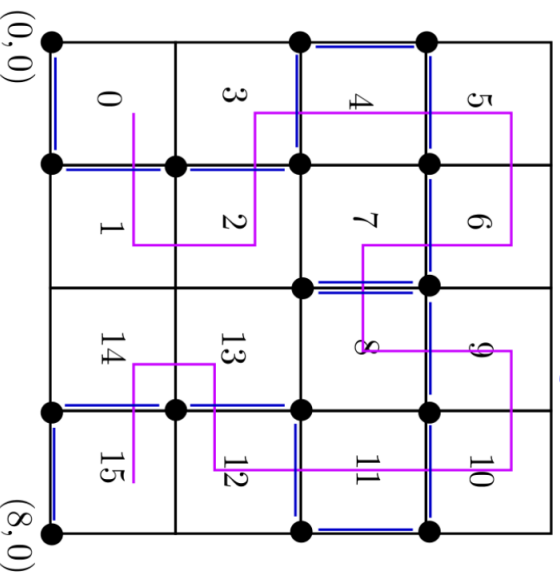
SFC DECOMP: DOMAIN DECOMPOSITION

** can merge square cells into bigger subdomains*



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- Lemma Hamiltonian path is guaranteed to exist
- idle movement can be minimized
 - if subdomains are solved to optimality, any discontinuities (idle movements) occur only at boundary of print domain

SUBDOMAIN TOOL PATH: MIP

* Miller-Tucker-Zemlin (MTZ) -type formulation

— turn cost g_j at node j :

$$g_j \geq A_{ijk} (x_{ij} + x_{jk} - 1), \quad g_j \geq 0$$

\swarrow \searrow \downarrow if $(i,j) \perp (j,k)$

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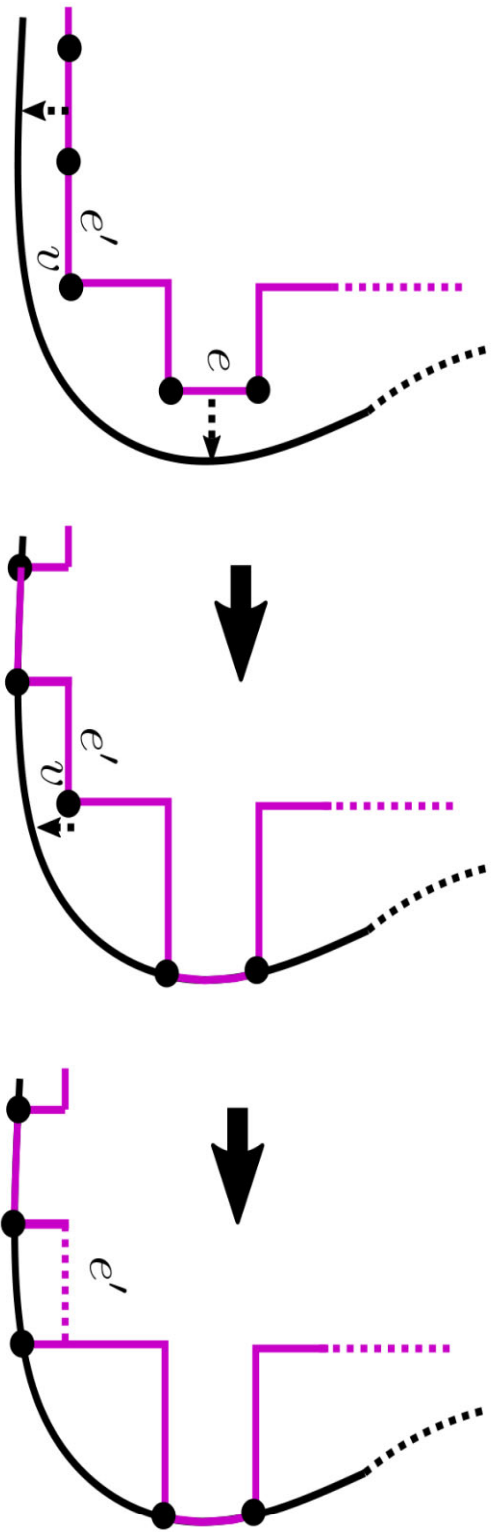
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- * solve large instances (approx):
 - relax subtour constraints
 - merge subtours using 2-opt + min-span.forest.
 - found Hamiltonian path in all 10,000+ instances except one

GENERAL GEOMETRY

* project print boundary vertices of IOB to the general polygon (\Rightarrow IOB)

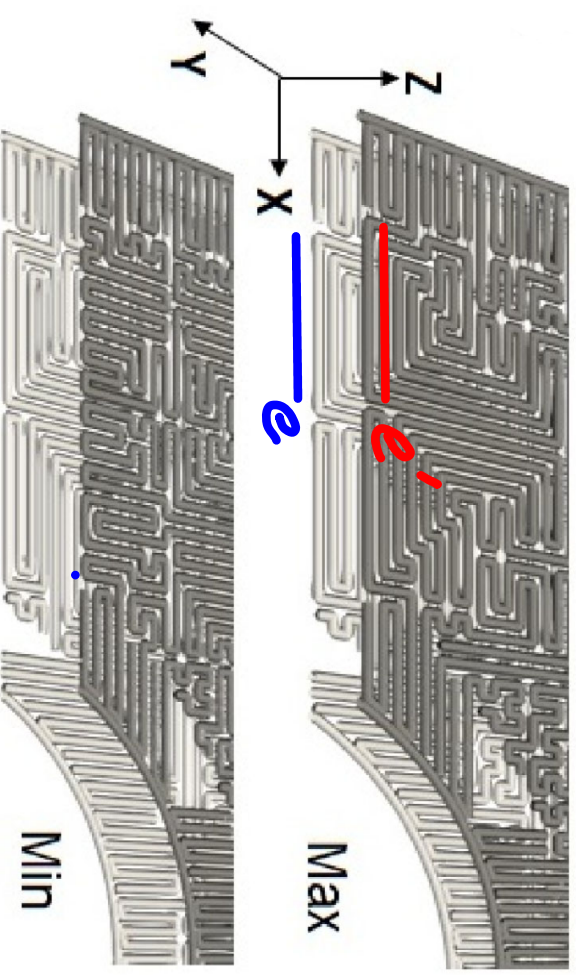


OPTIMIZING EDGE OVERLAPS

* overlap between edges

e and e' is

- max if $e = e'$



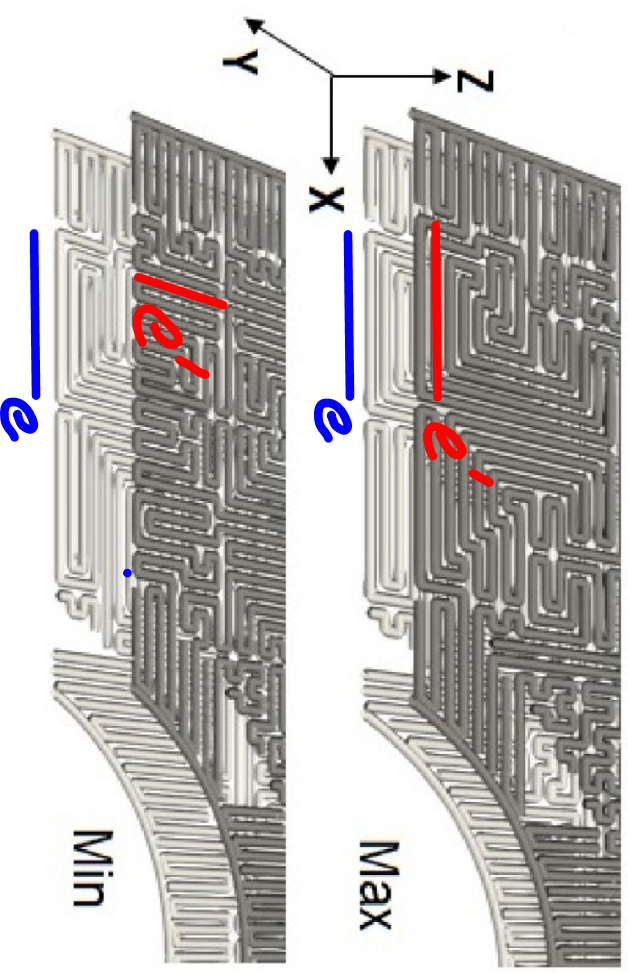
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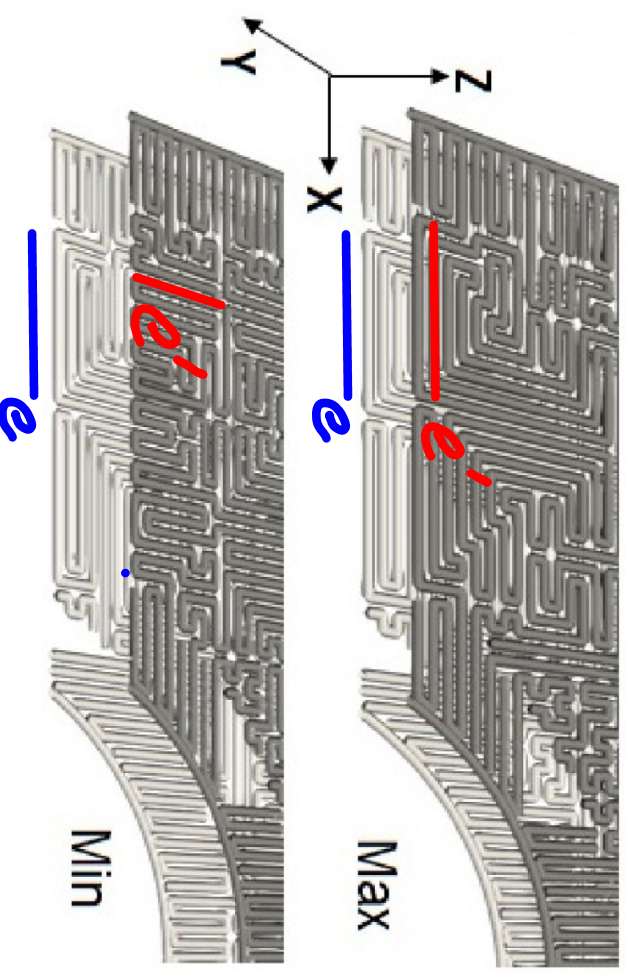
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* Set weight of e' lower (higher) for max (min) overlap

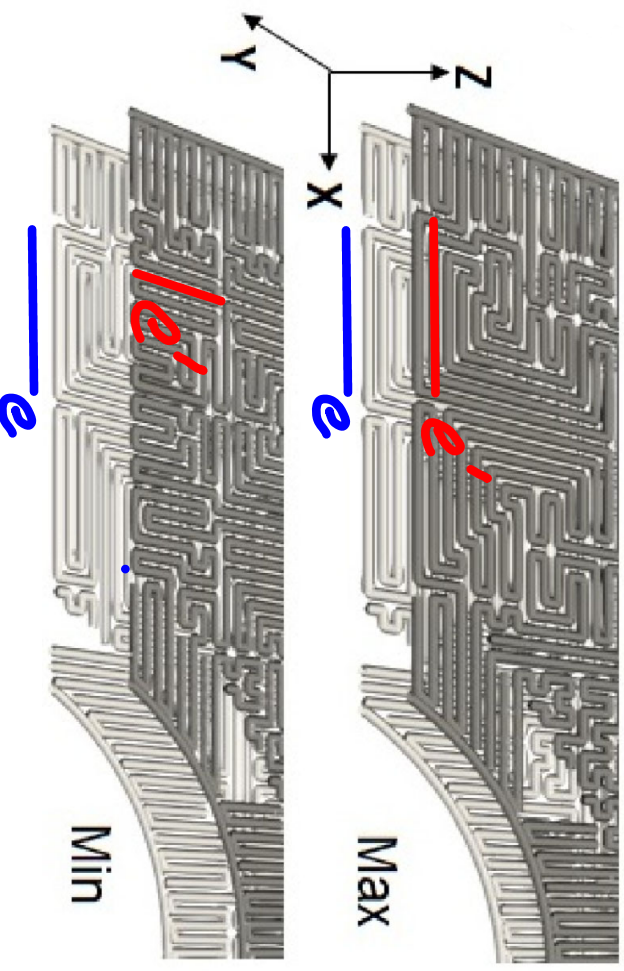
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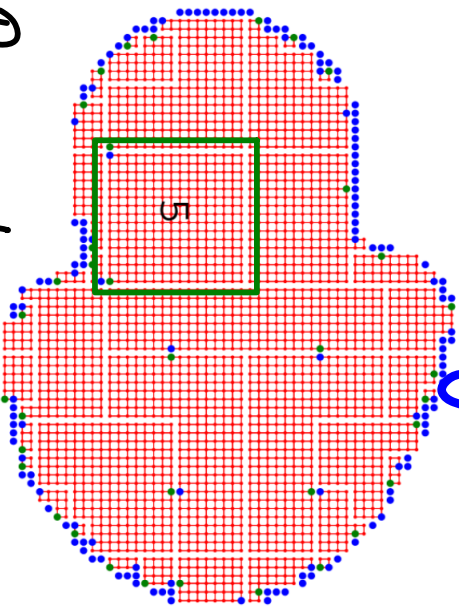
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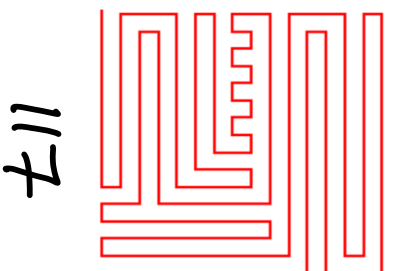
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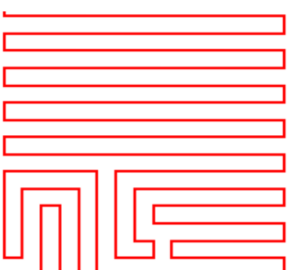
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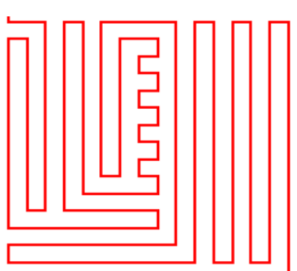
Bunny: layer 117



117



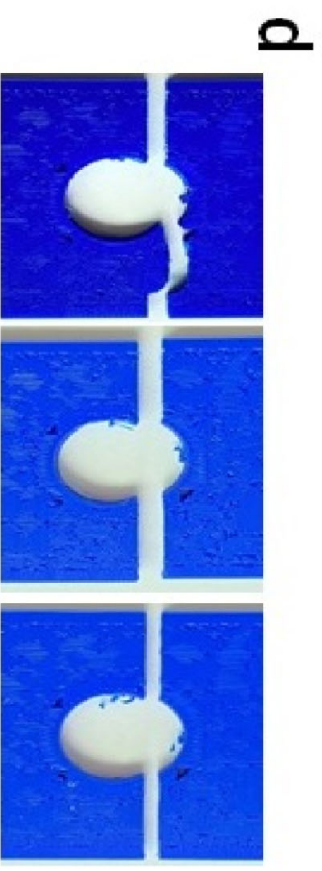
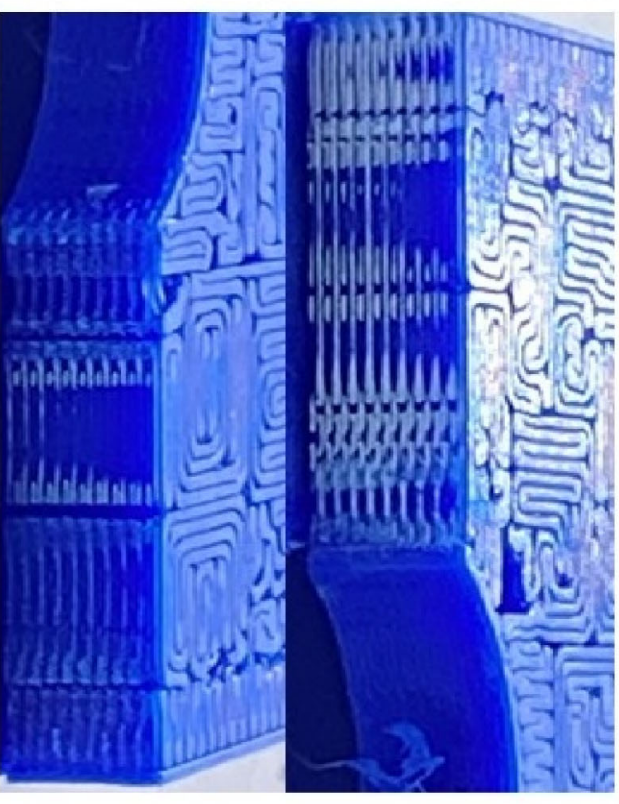
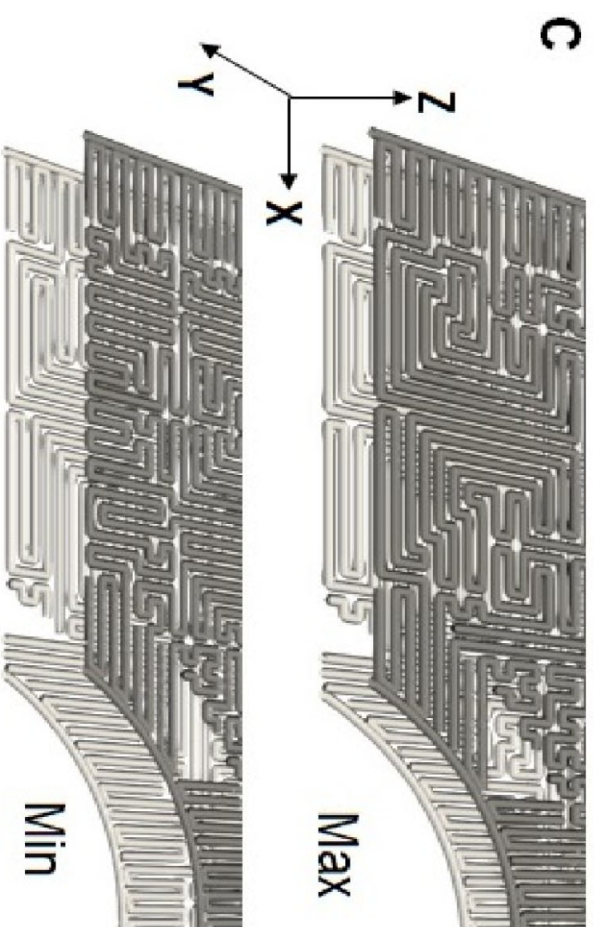
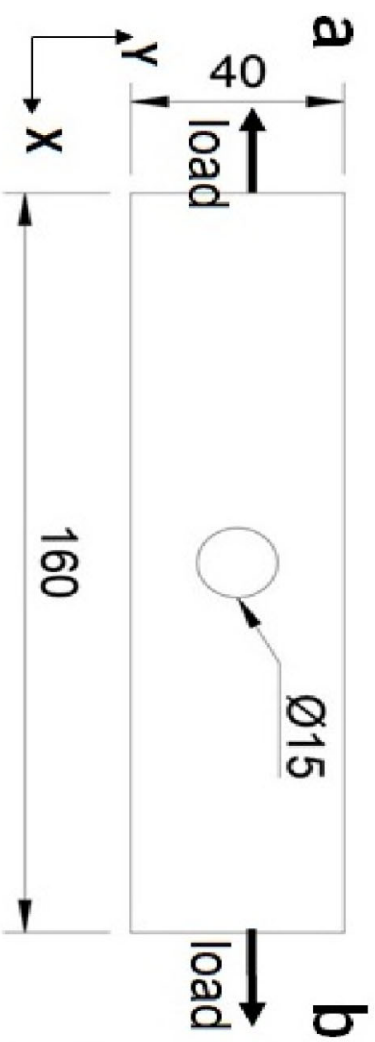
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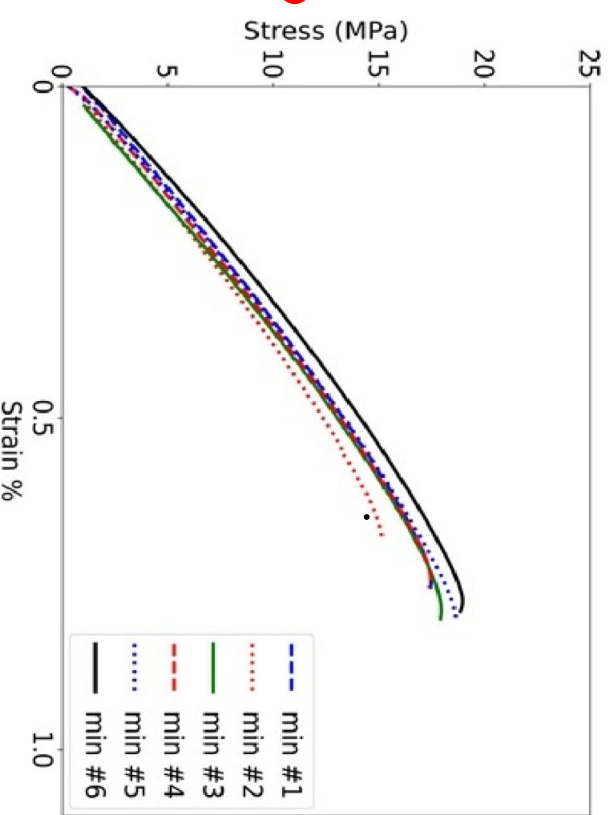
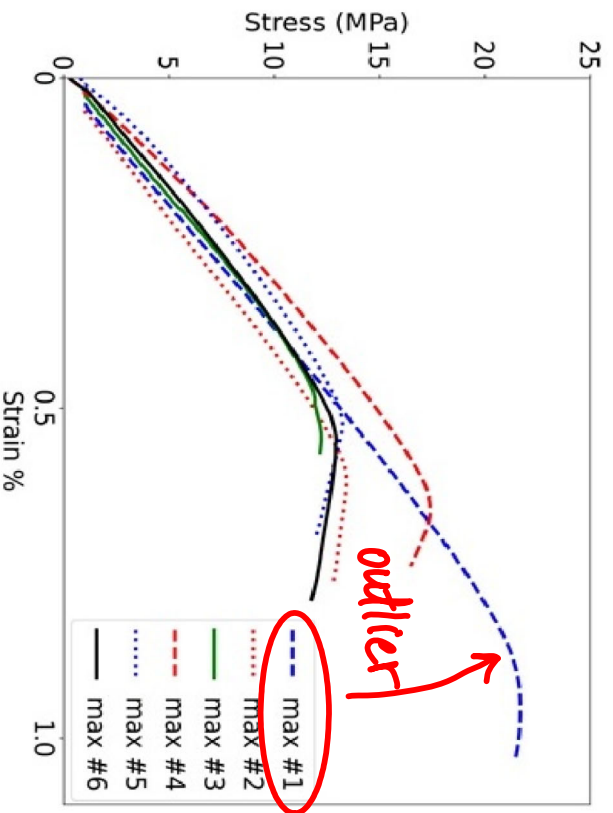
119

min edge overlap

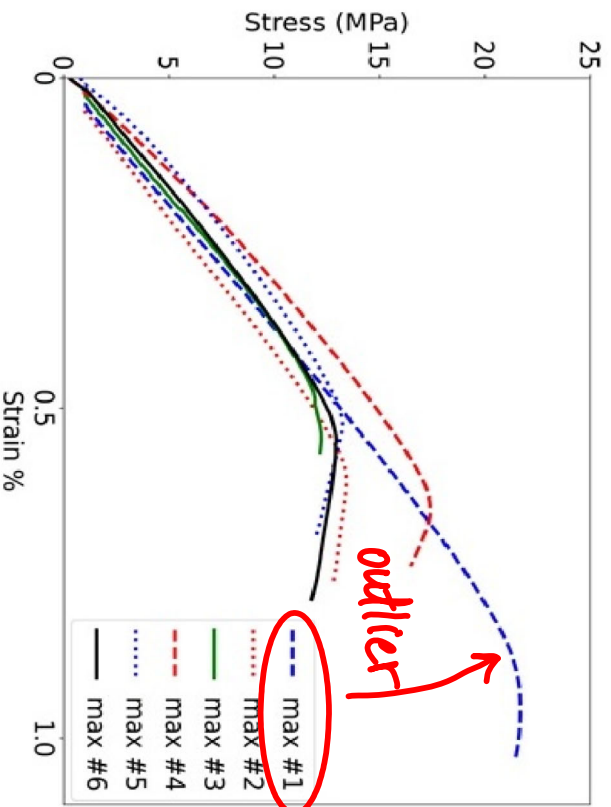
MECHANICAL TESTING



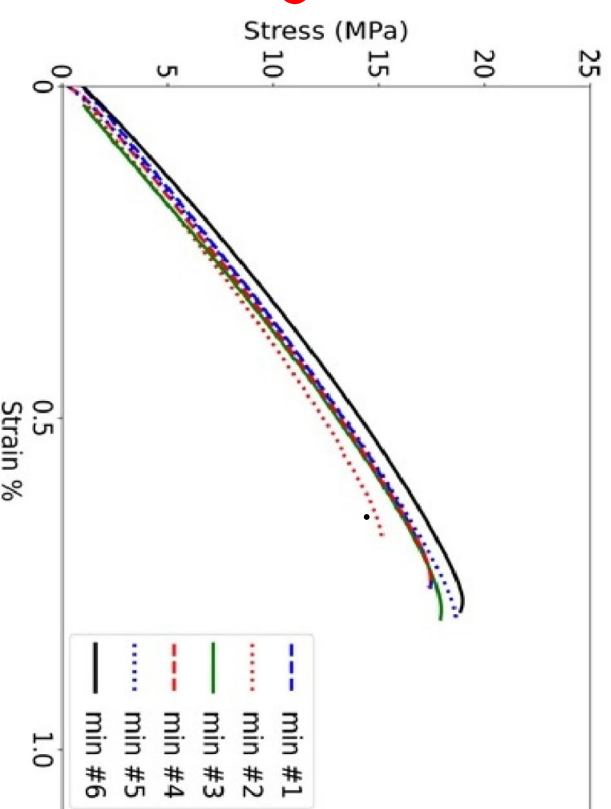
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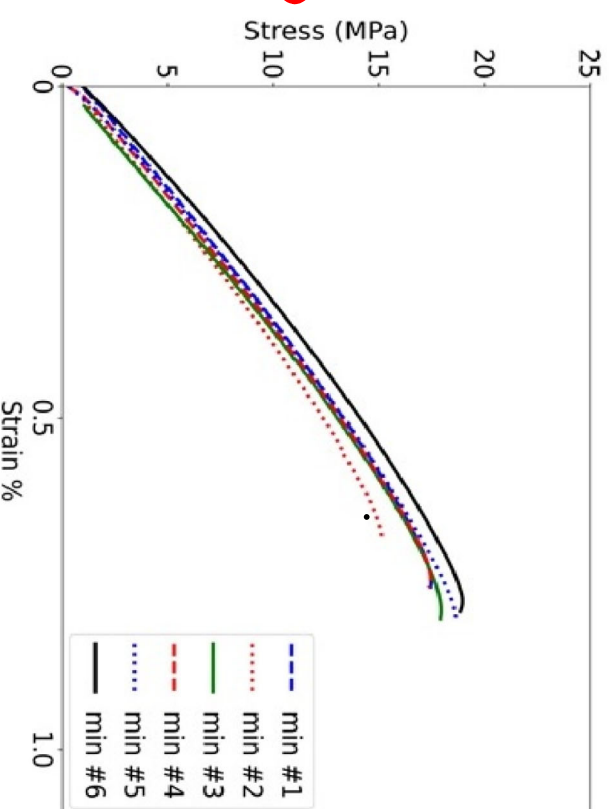
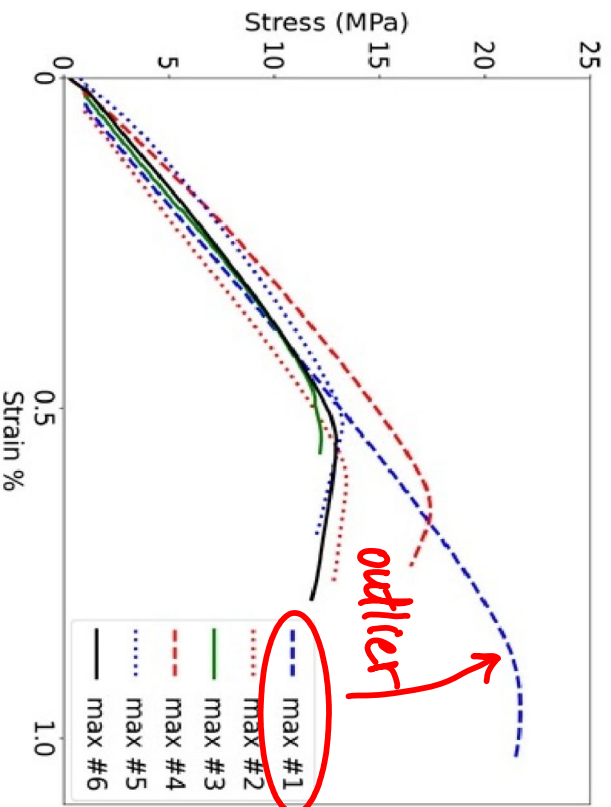


max overlap
avg. strength = 13.84 MPa
(w/o outlier)



min overlap
avg. strength = 17.60 MPa

MECHANICAL TESTING



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* min edge overlap had better strength

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