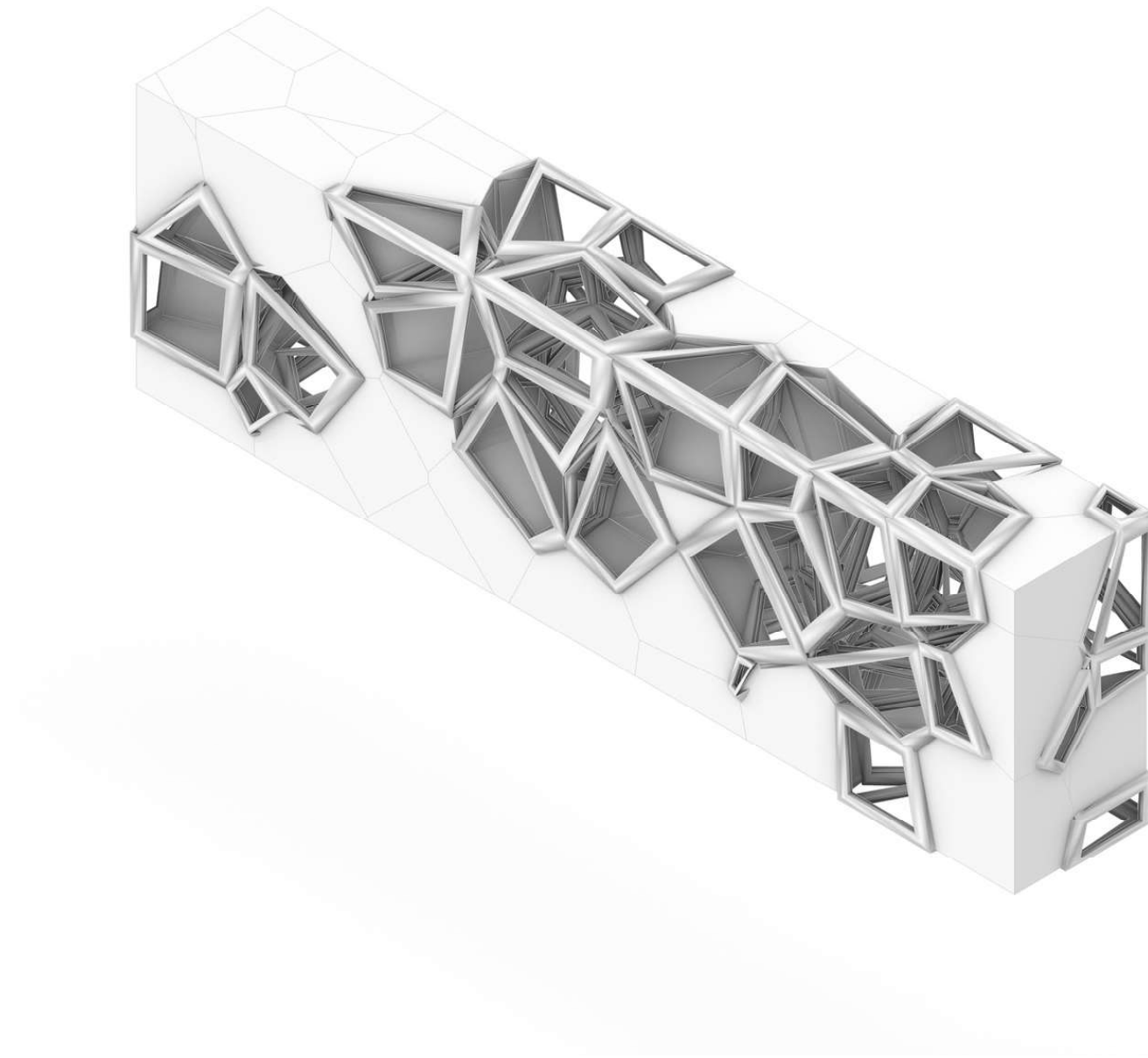
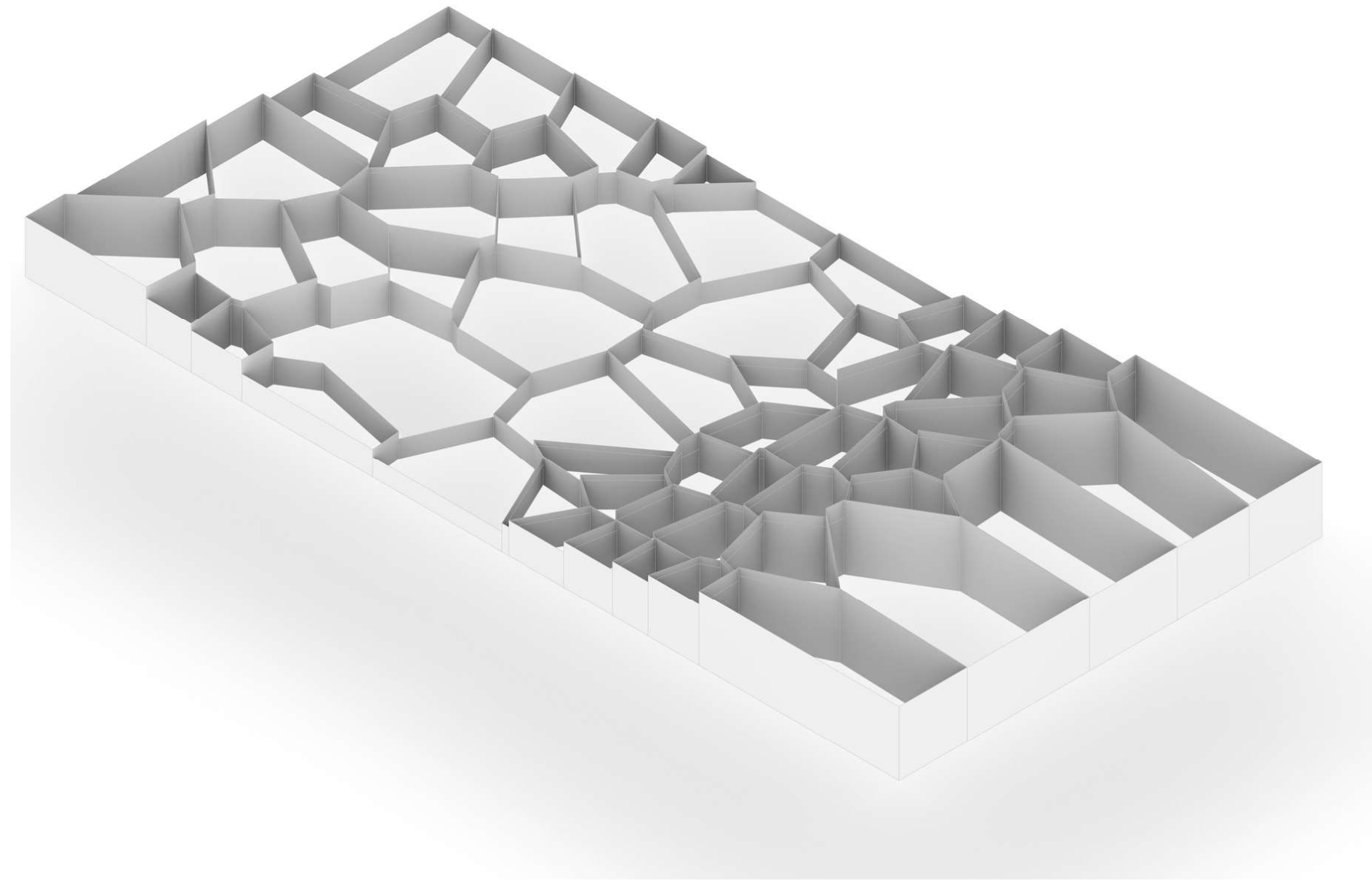
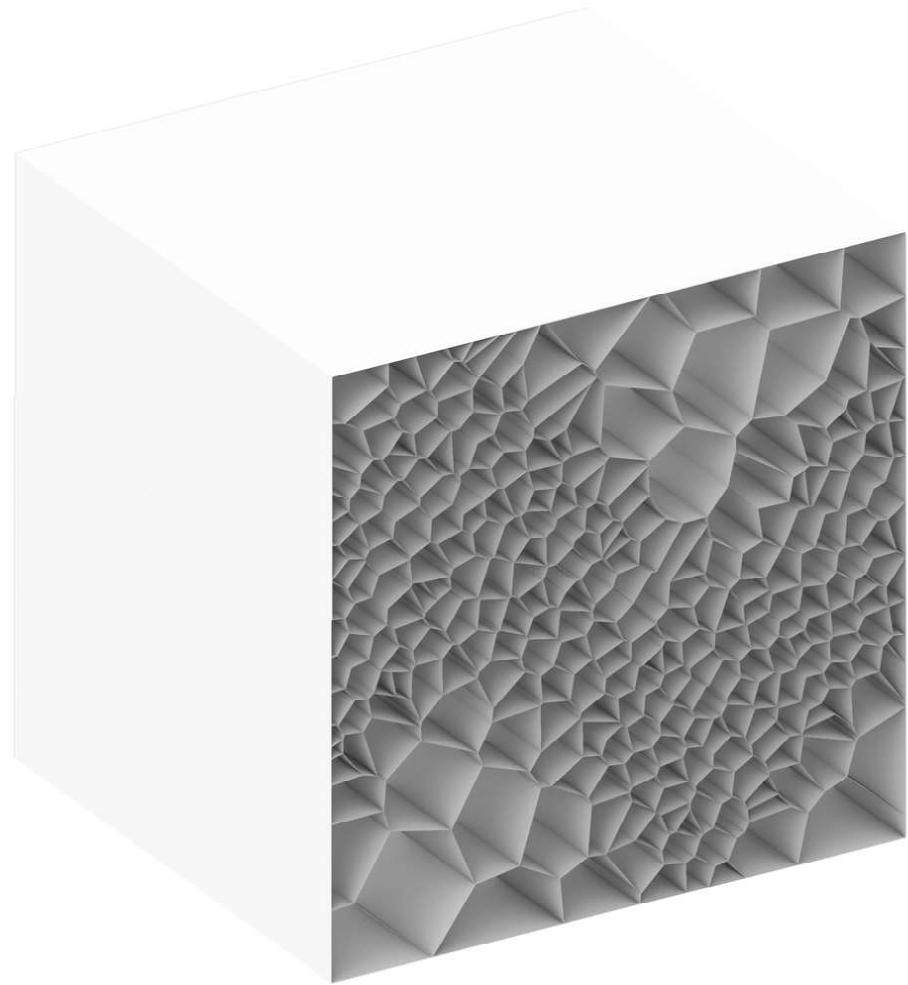


FRAGMENT

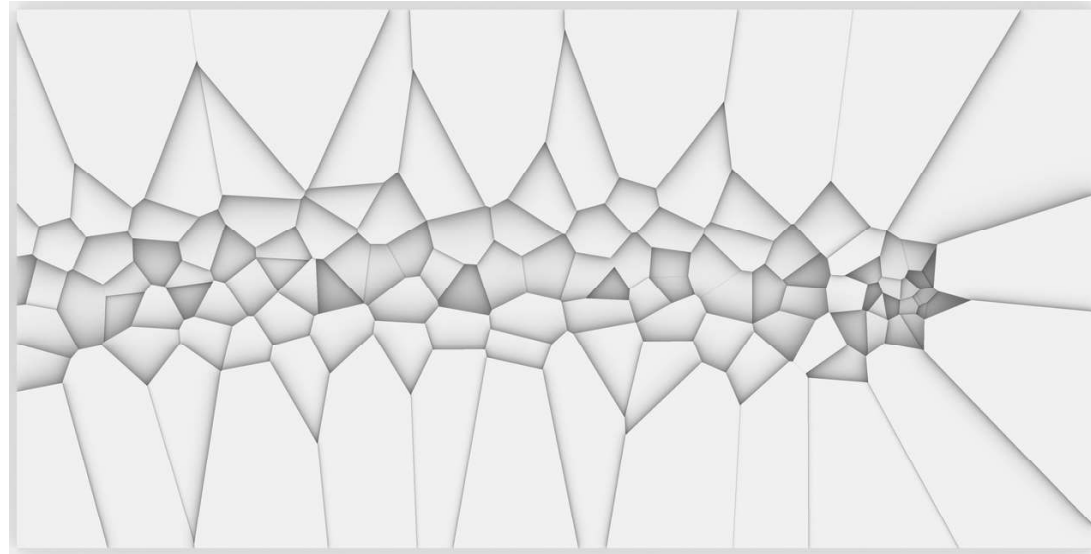
early stages &
exploration



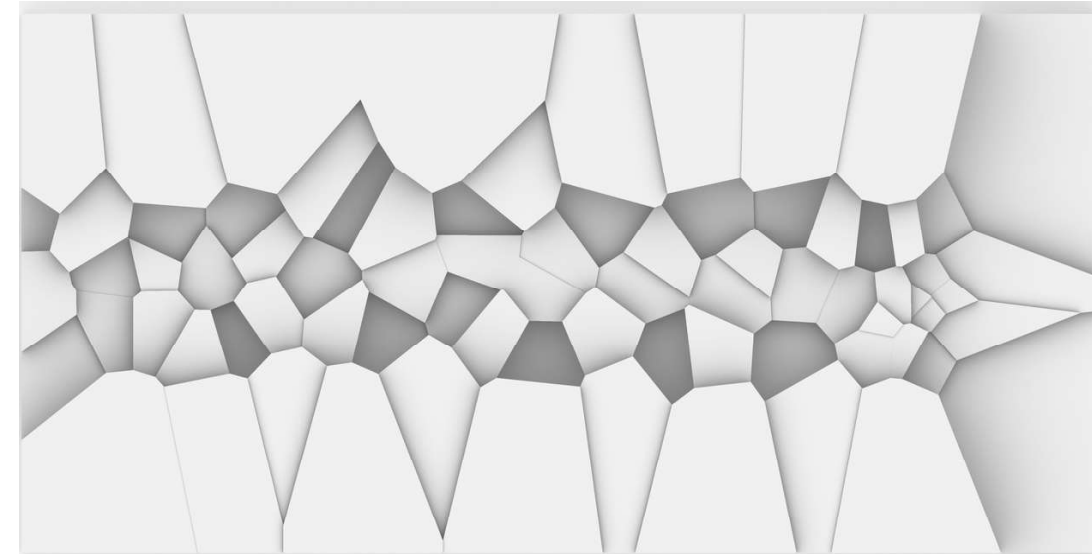
VORONOI SIZING & DEPTH



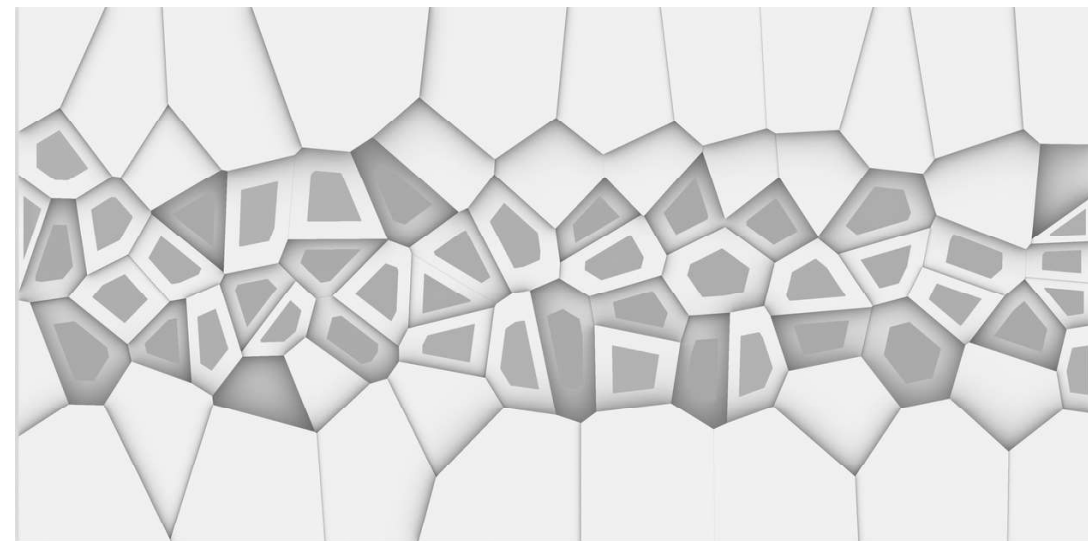
WALL APPLICATIONS & EXPERIMENTATIONS



DEPTH



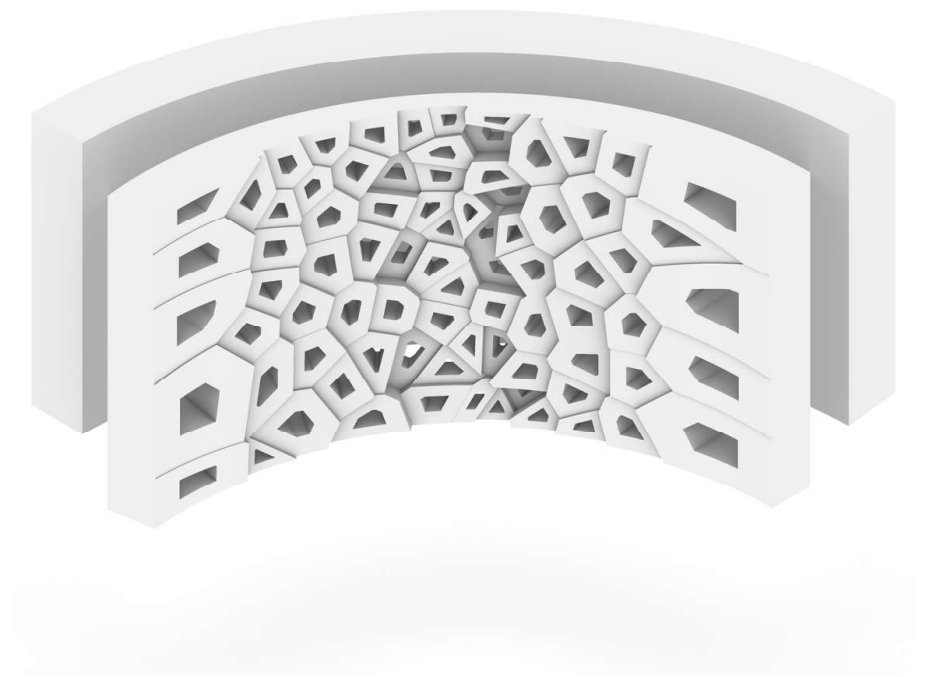
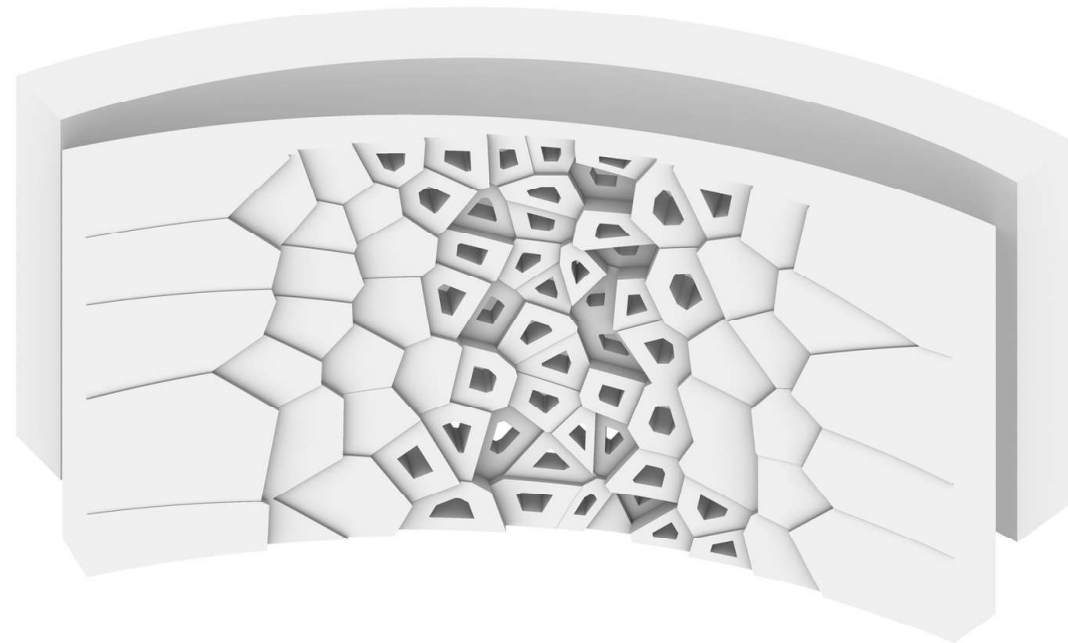
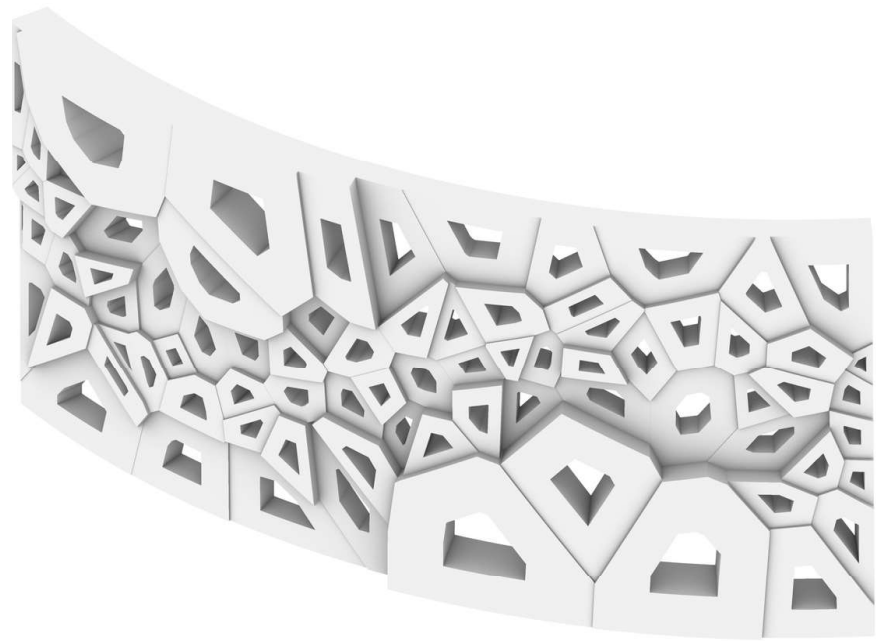
DEPTH & OPENINGS



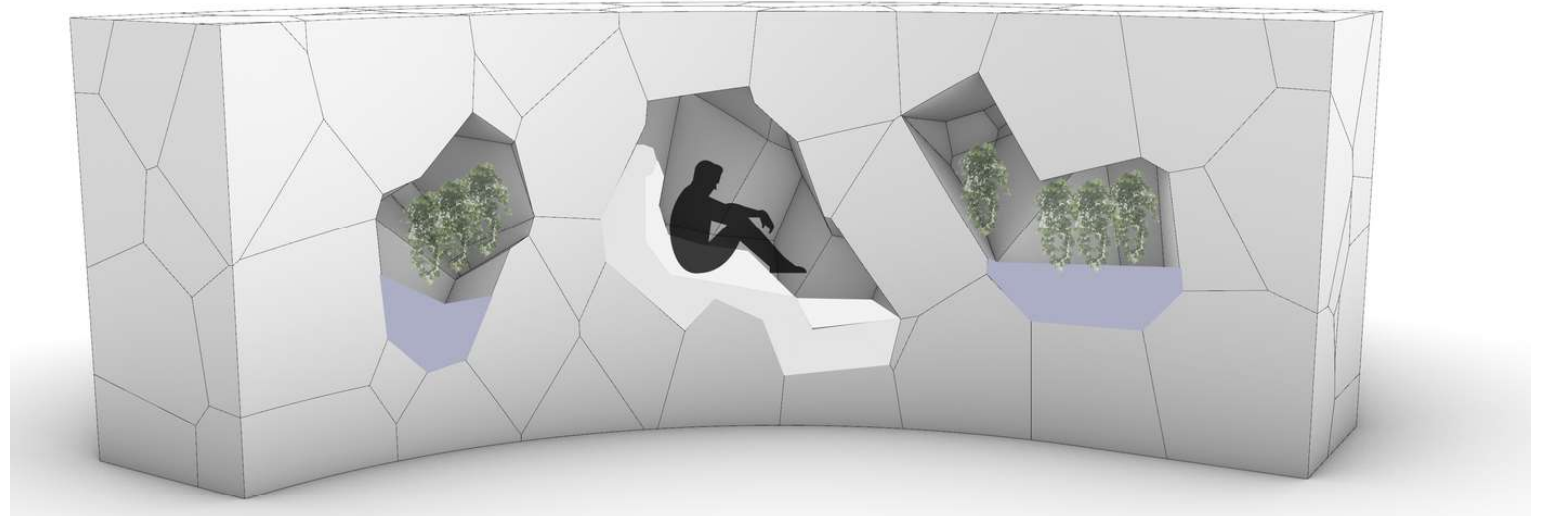
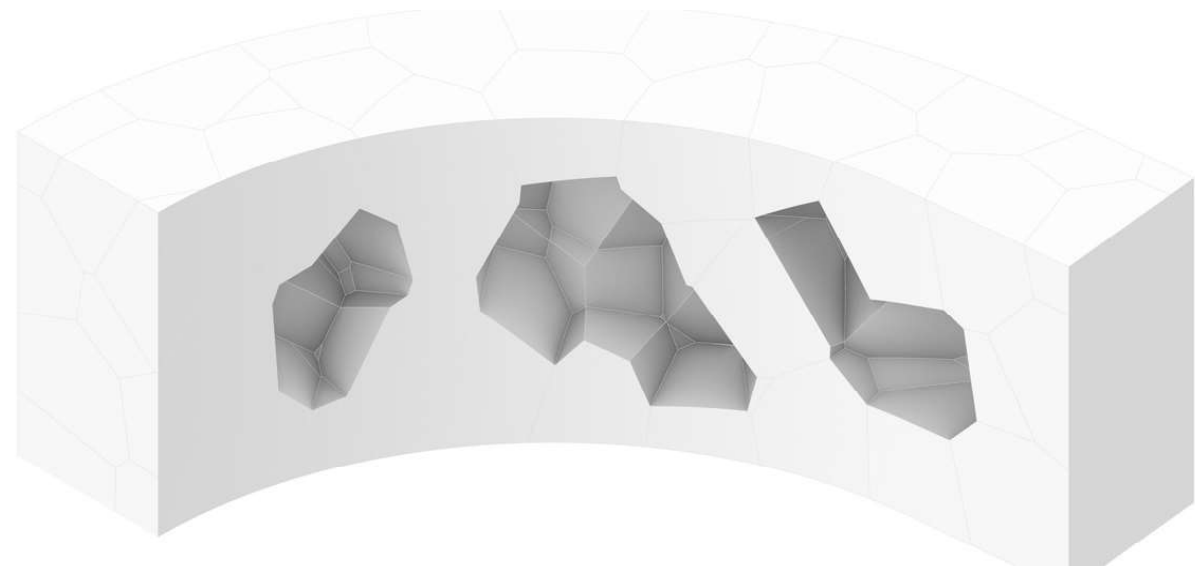
DEPTH & CELL OPENINGS

WALL APPLICATIONS & EXPERIMENTATIONS

DEVELOPING WALL DESIGN CONCEPTS



EXPLORING SPATIAL QUALITIES



FRAGMENT

THE OPPORTUNITIES

Implemented



STRUCTURE



AQUAPONICS



ACOUSTICS

Explored



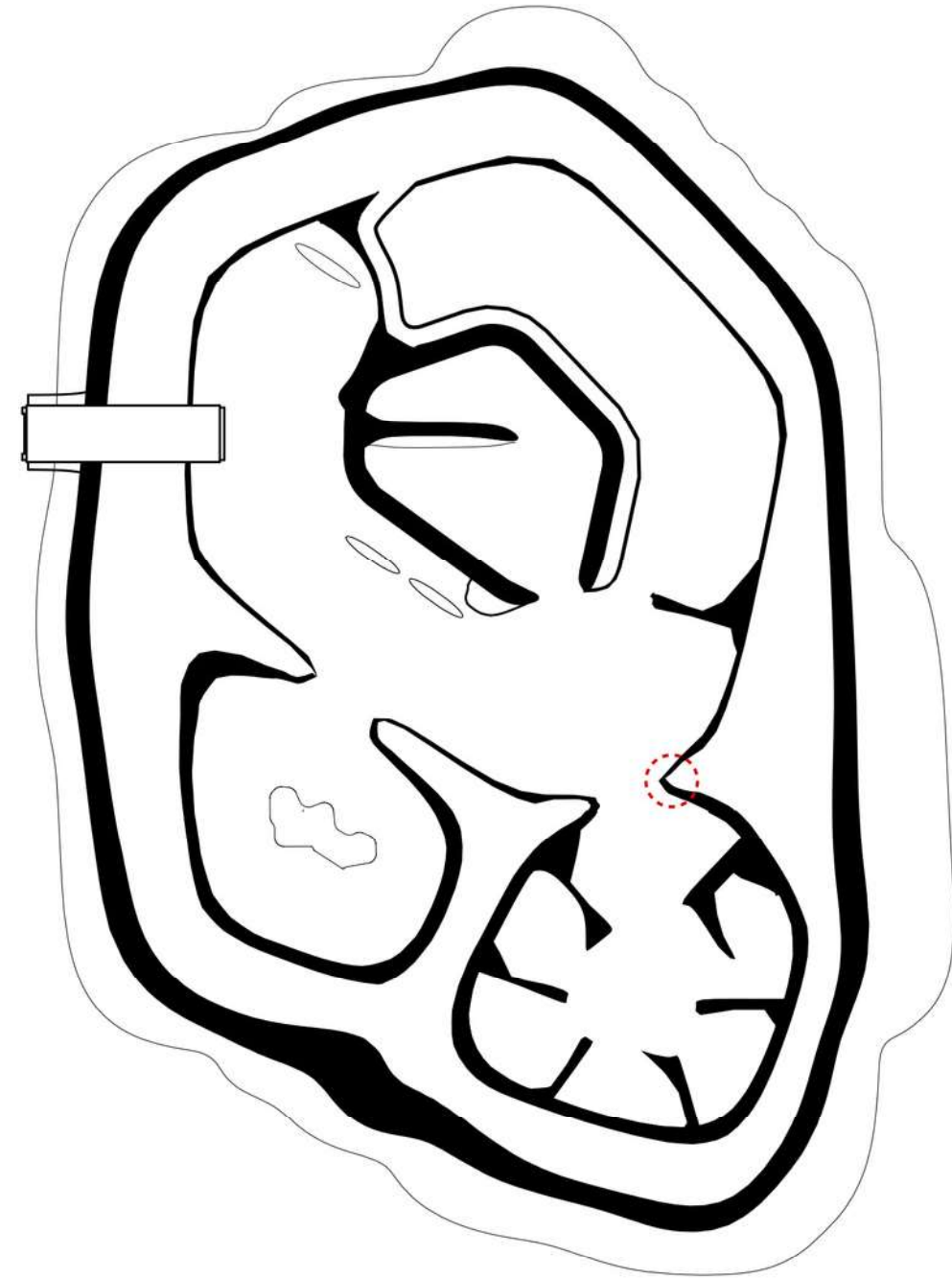
FURNITURE



LIGHTING

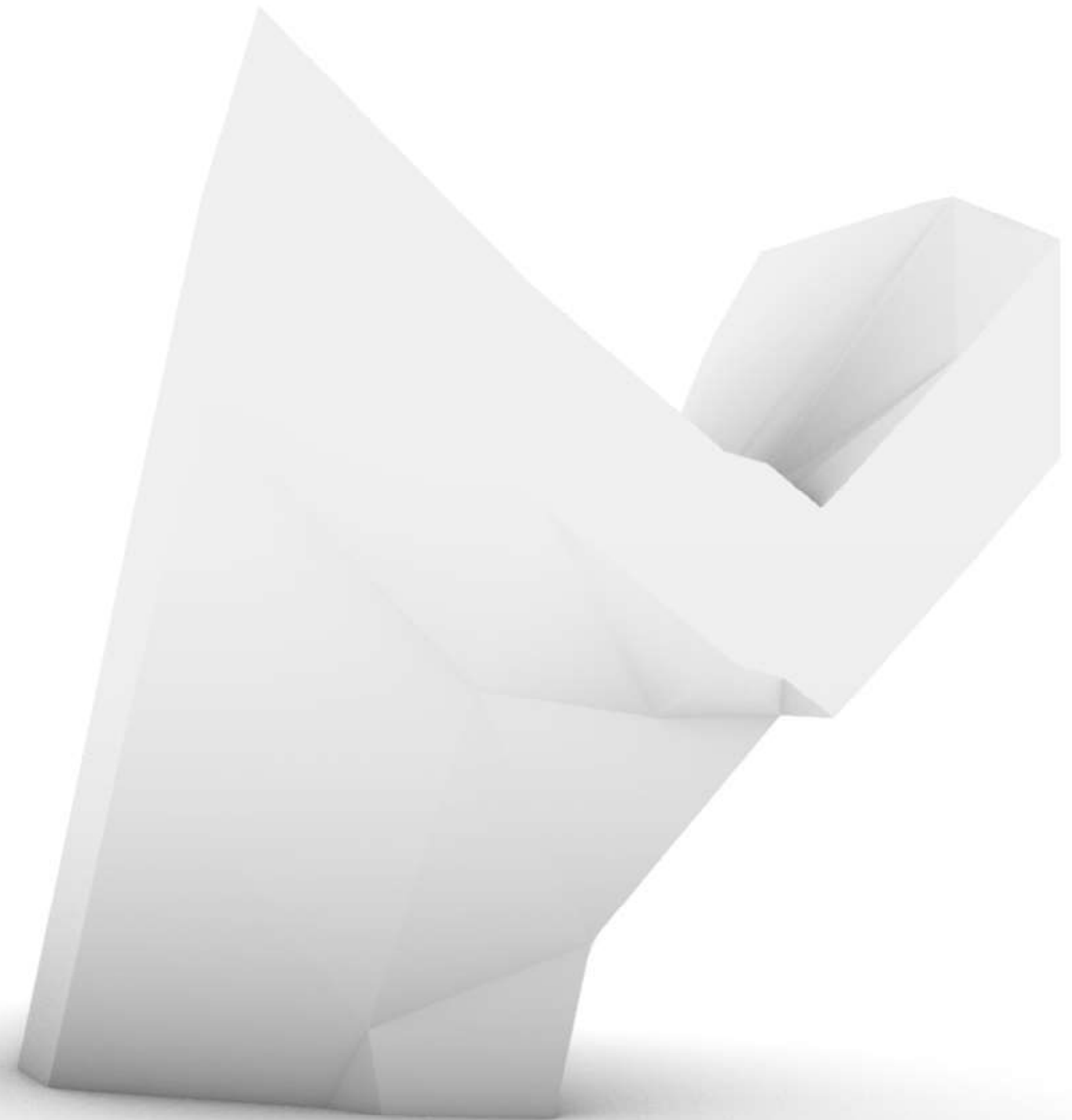
FRAGMENT

IT IS LOCATED AT AN INTERSECTION OF TWO CATENARY SHELLS TO SHOWCASE THE WAY FORCE DISTRIBUTION OF A COMPRESSIVE STRUCTURE WORKS IN SUCH A SCENARIO



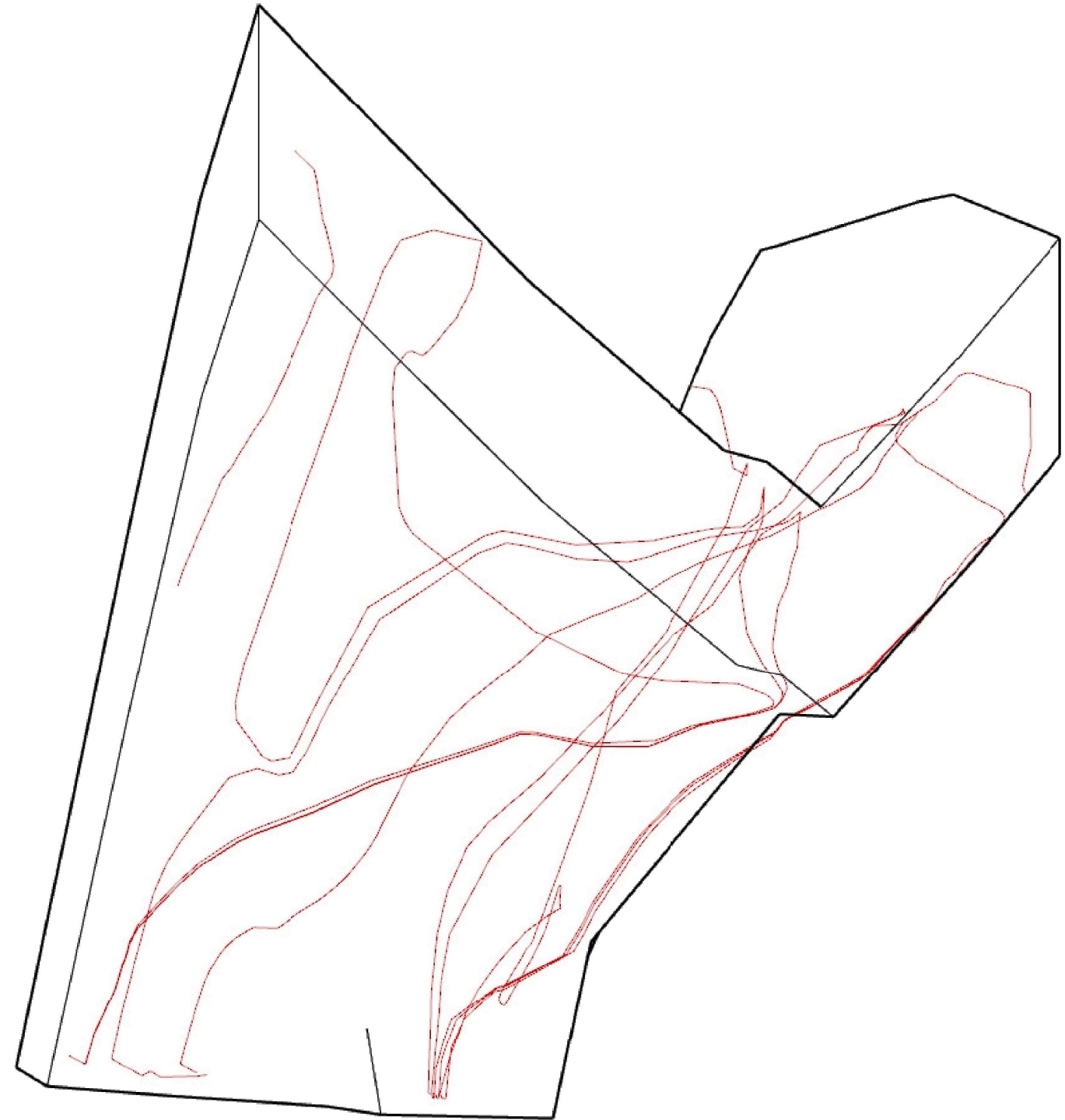
FRAGMENT

A FRAGMENT IS CHOSEN TO
FURTHER ELABORATE THE
STRUCTURAL CONDITIONS AND
ALLOWS ELABORATING
ADDITIONAL ELEMENTS ON A
SMALLER SCALE



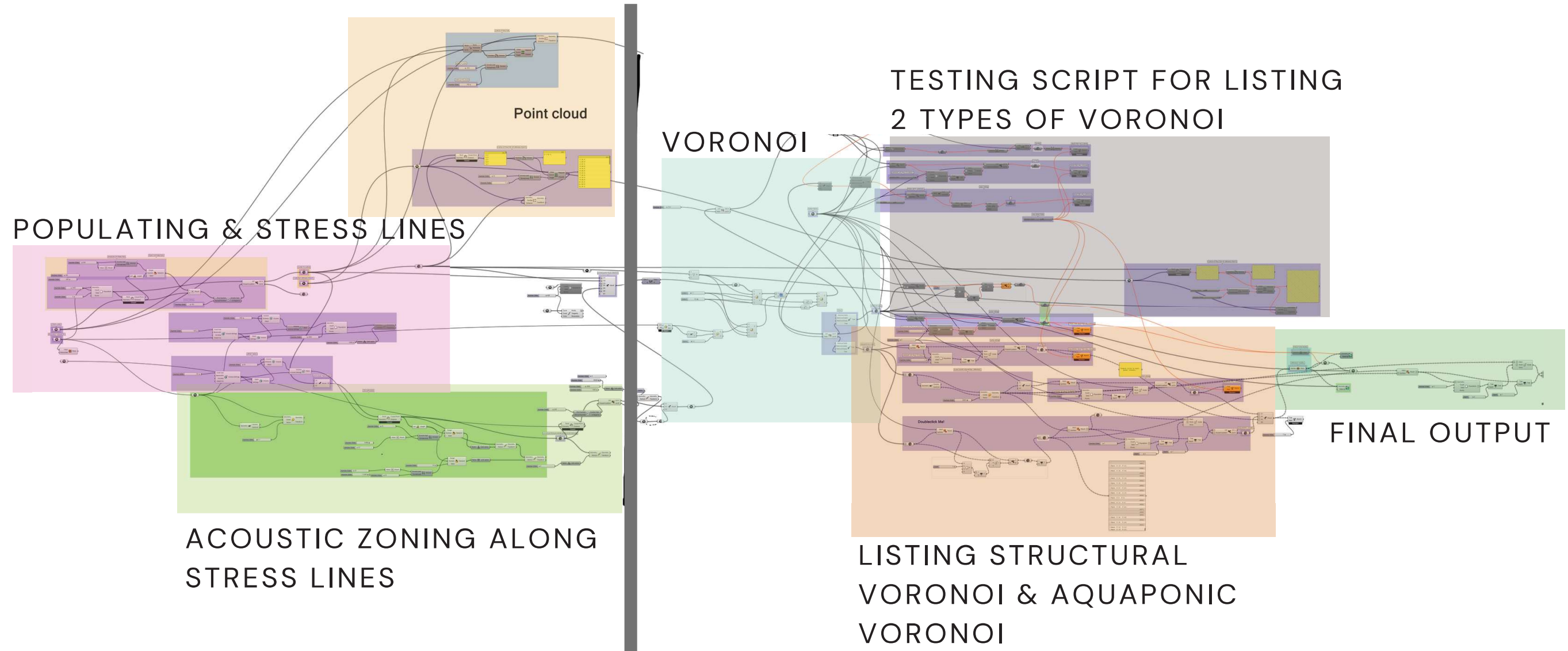
FORCE ANALYSIS

PICKING AN INTERSECTION OF A
SEPARATION OF TWO SPACES
SHOWCASES THE OVERLAY OF
FORCES WORKING IN
COMPRESSION COMING FROM
THE PRINCIPLE OF A FLIPPED
CATENARY



THE SCRIPT

CURVE ATTRACTOR - POINT ADJUSTING

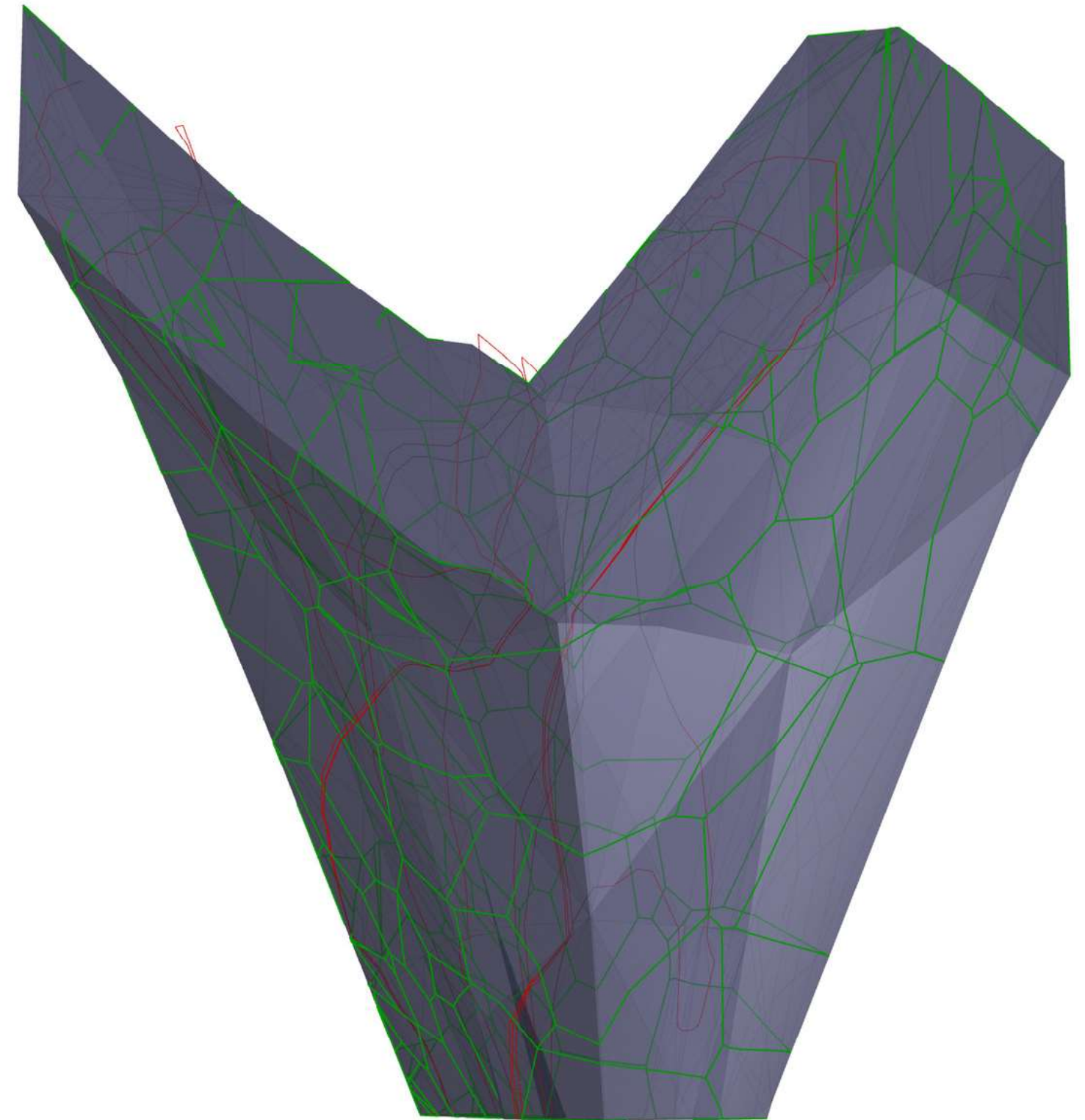


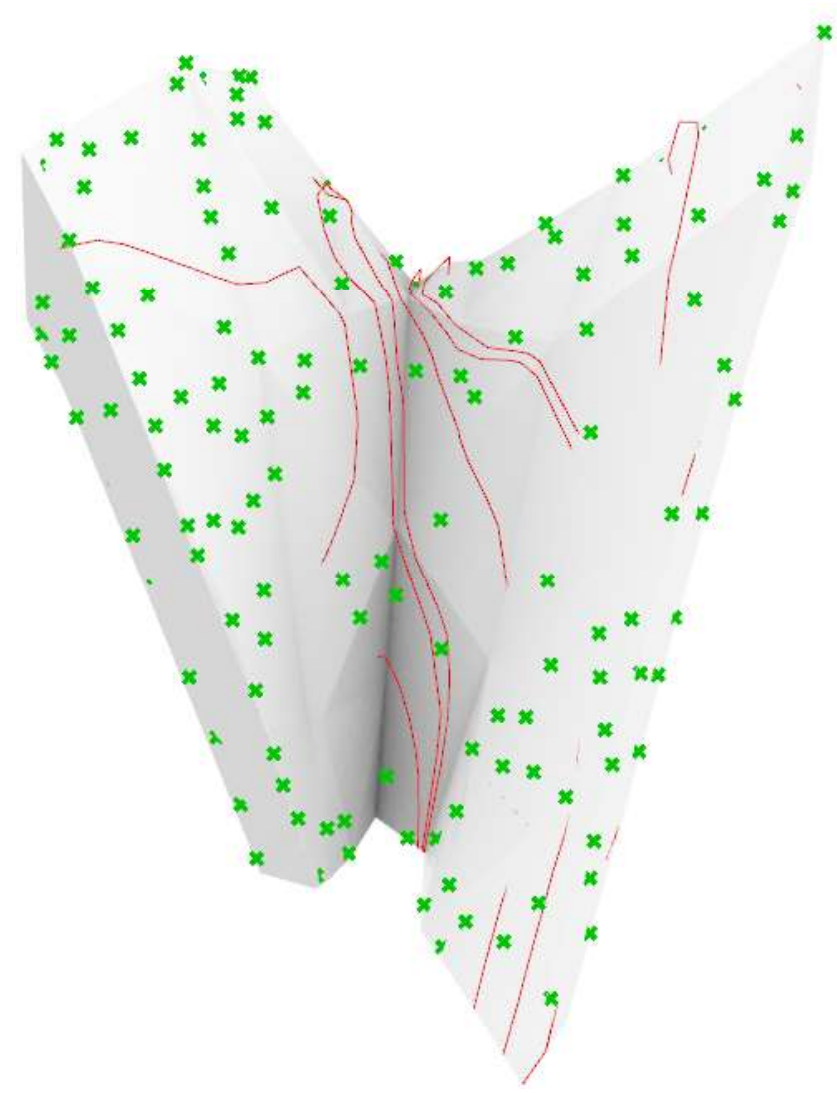
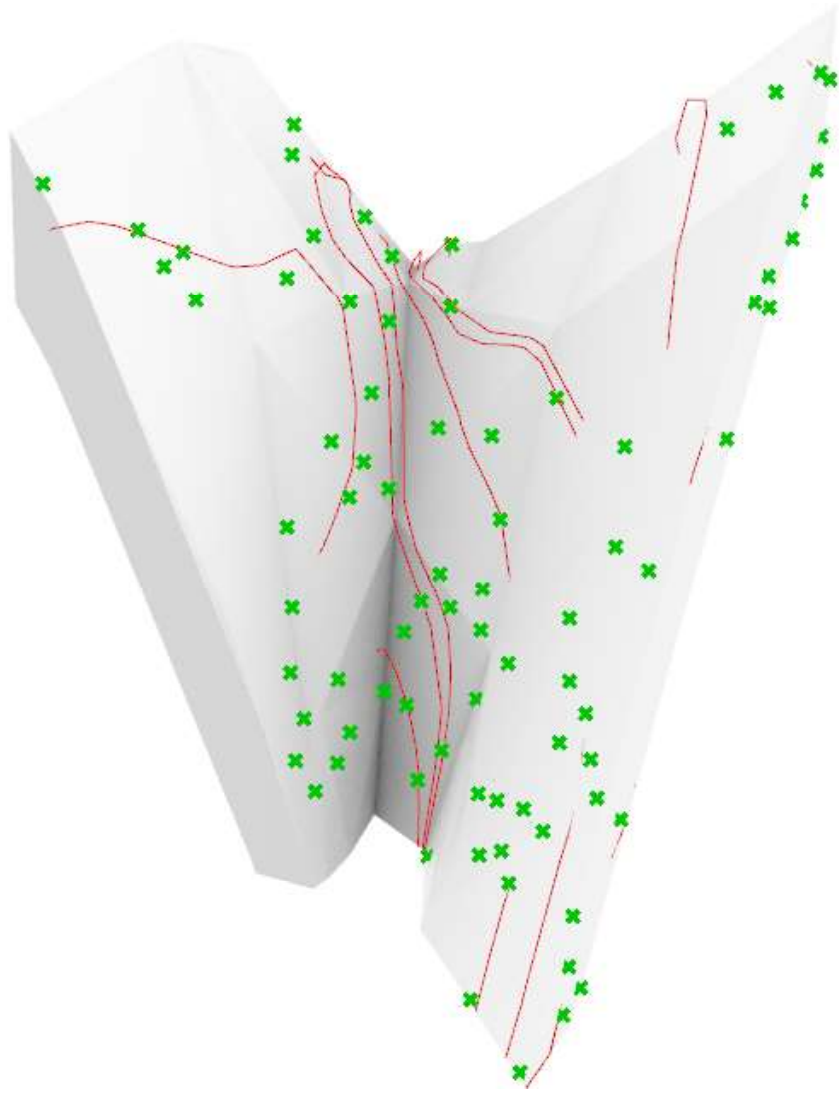
POINT ADJUSTMENT, ZONING, OVERALL
DESIGN INPUT

CREATING VORONOI, MESH SPLIT FOR
OPENINGS, VORONOI TYPES & THEIR
OUTPUT

VORONOI ZONING

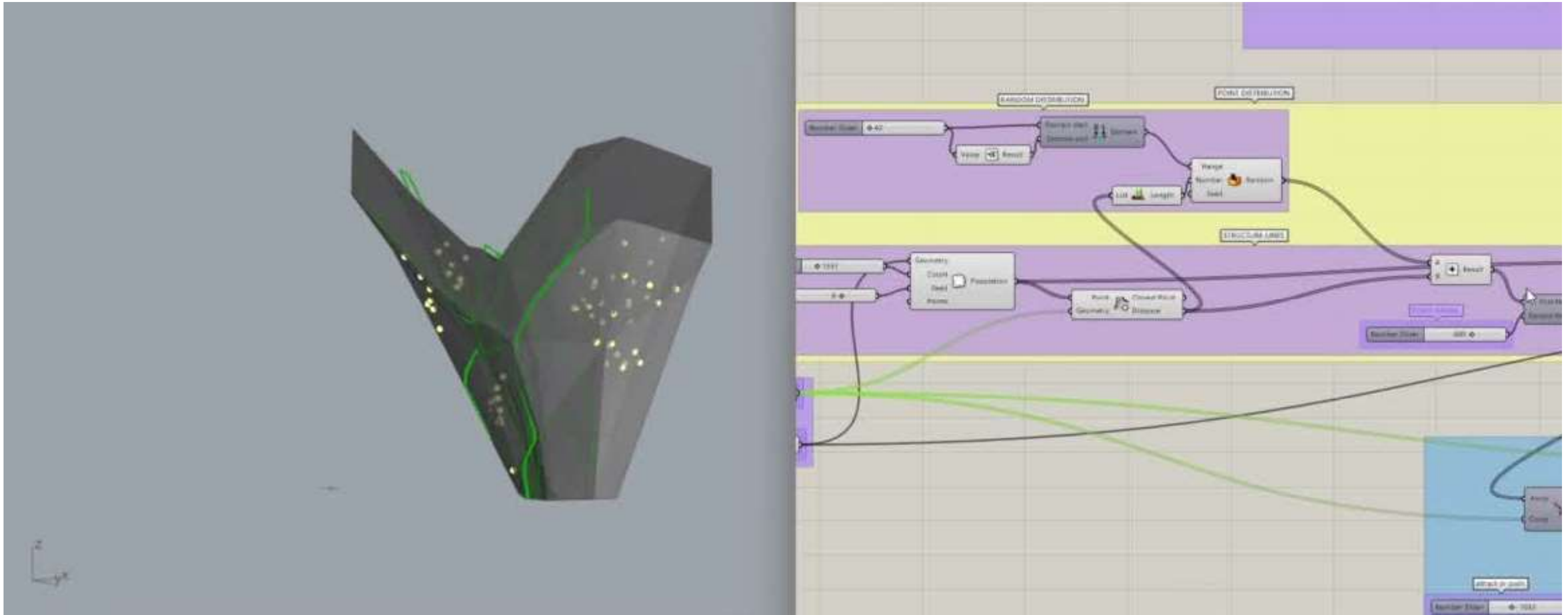
Using curve attractor to zone voronoi. The distance between the cells are adjustable and go along the stresslines.



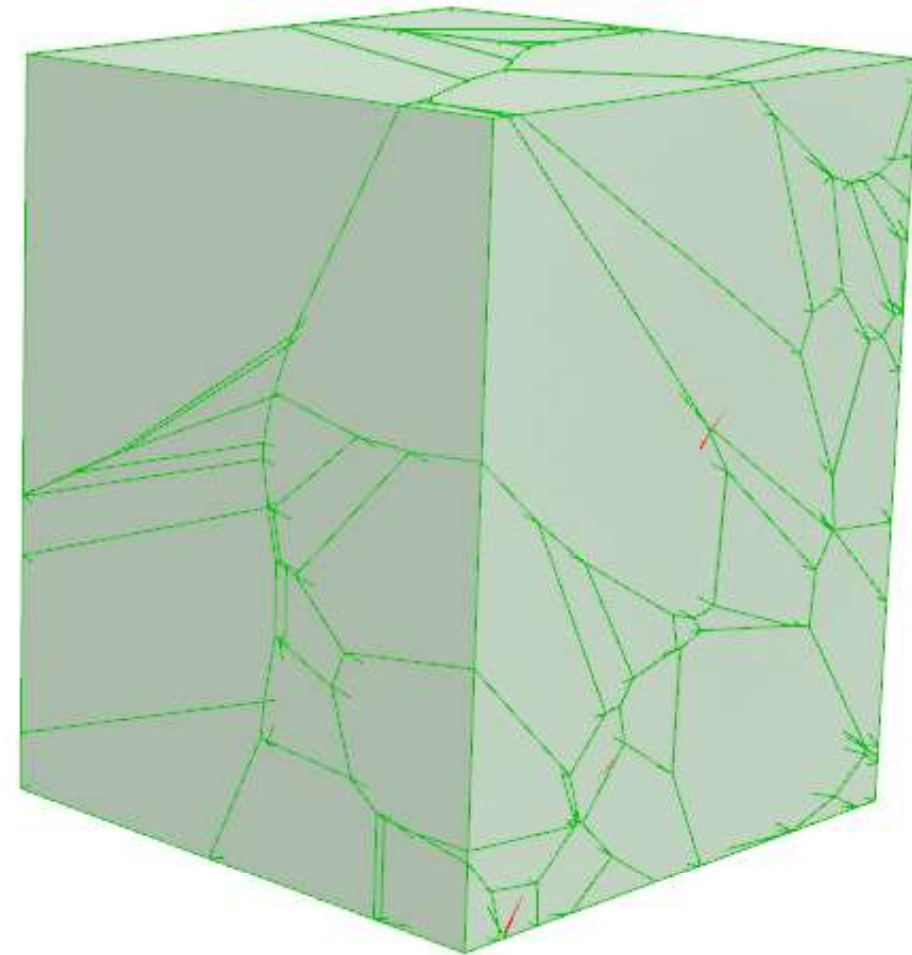
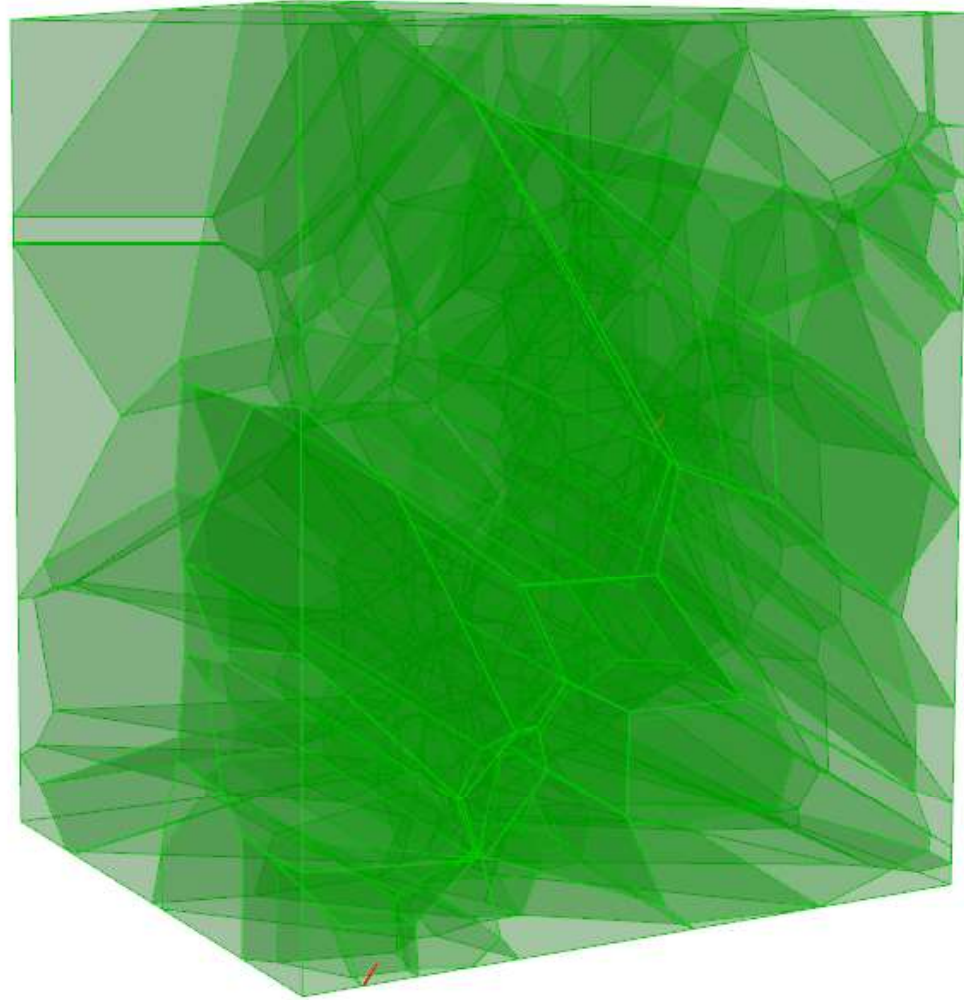


Selection of points close/away from the stress lines
Point randomness with 500 mm offset

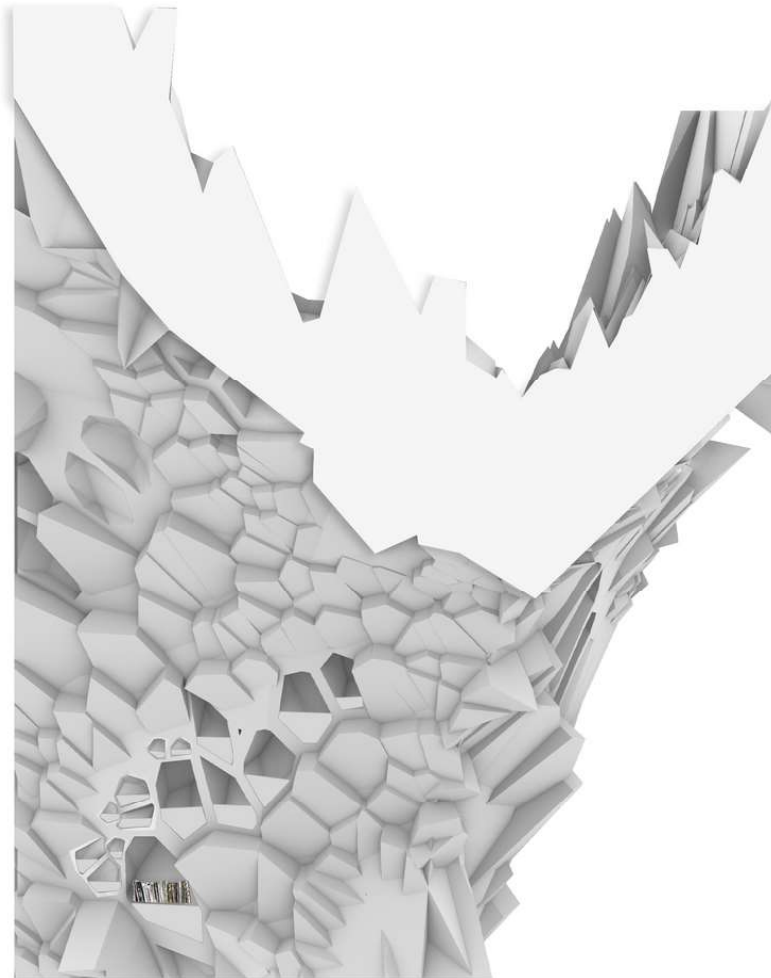
POINTS ZONING



VORONOI SIZE VARIETY

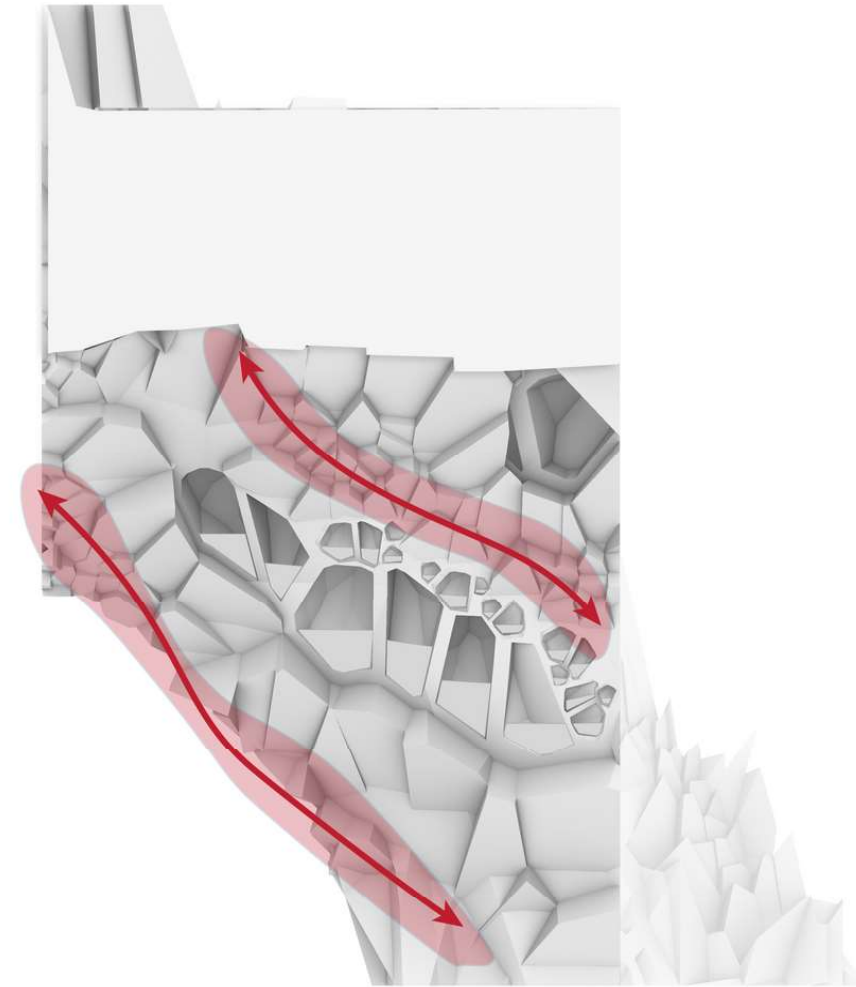
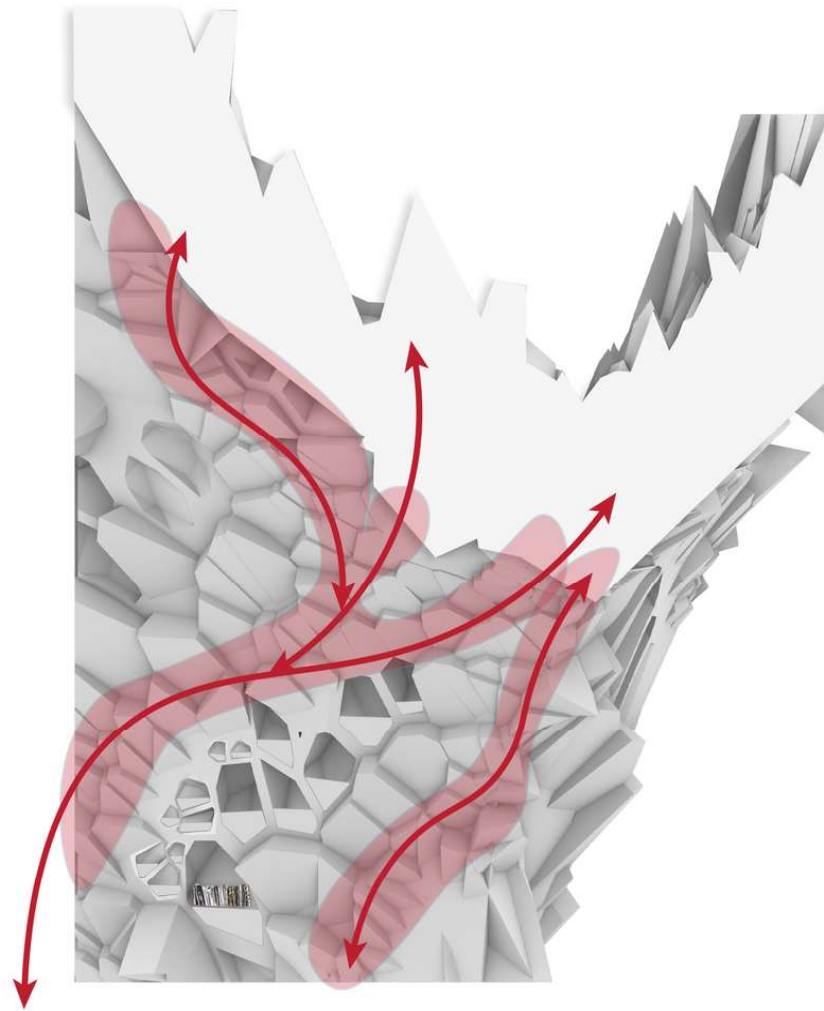


FINAL FRAGMENT DESIGN





STRUCTURE

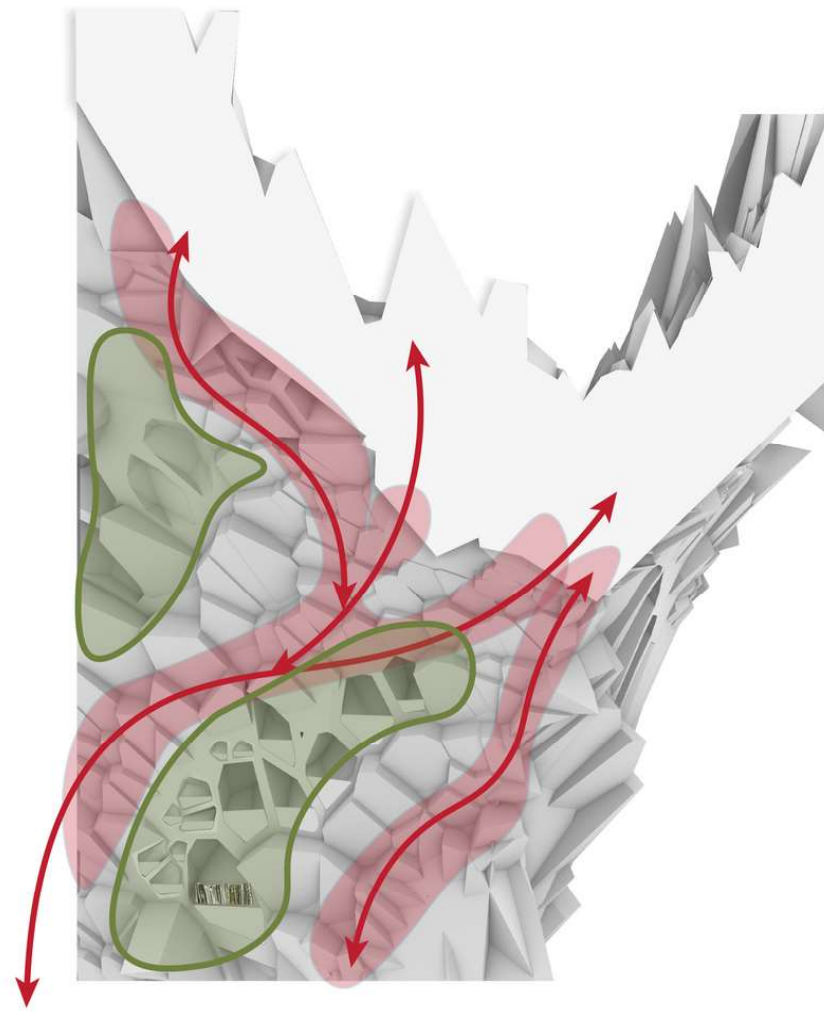




STRUCTURE



AQUAPONICS





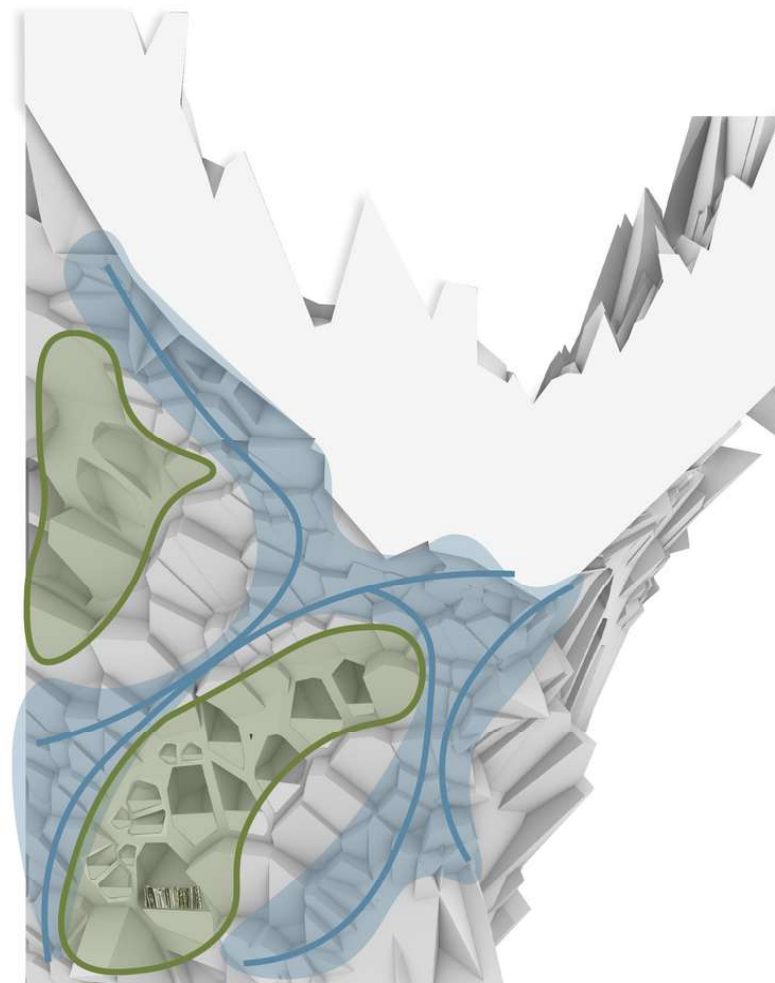
STRUCTURE



AQUAPONICS

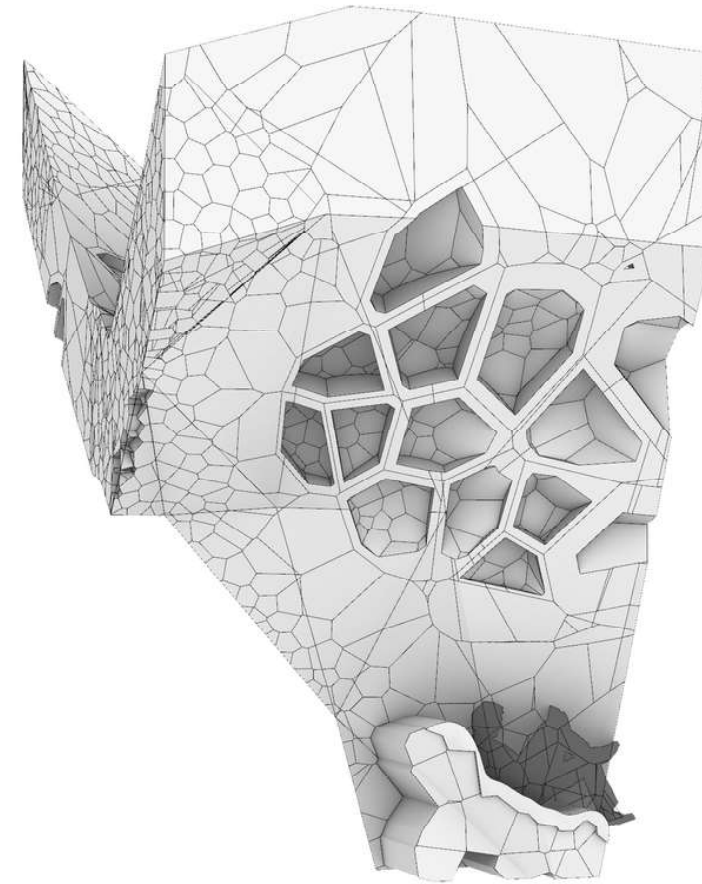
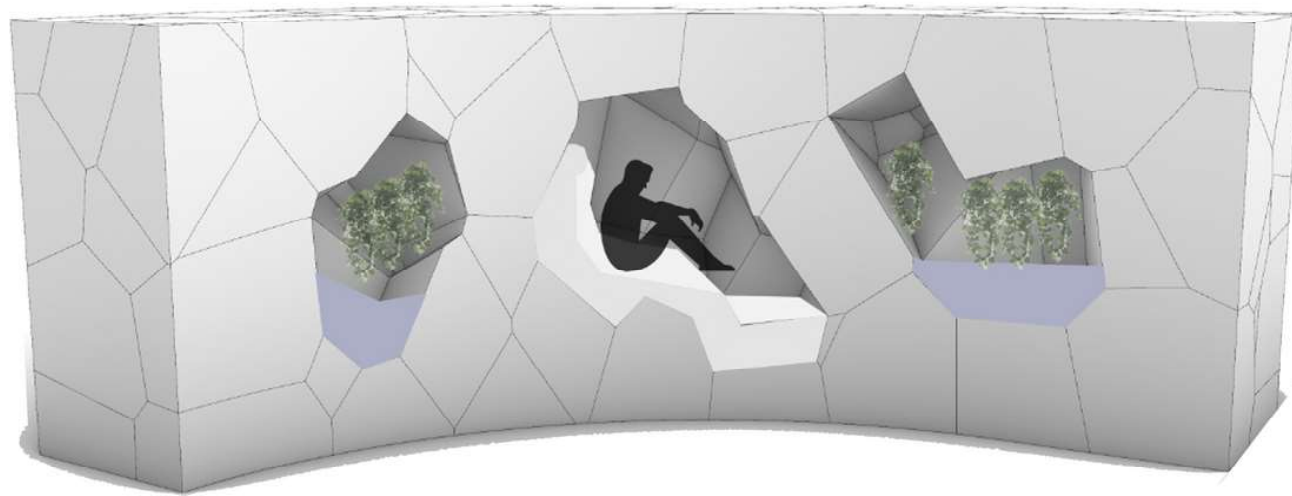


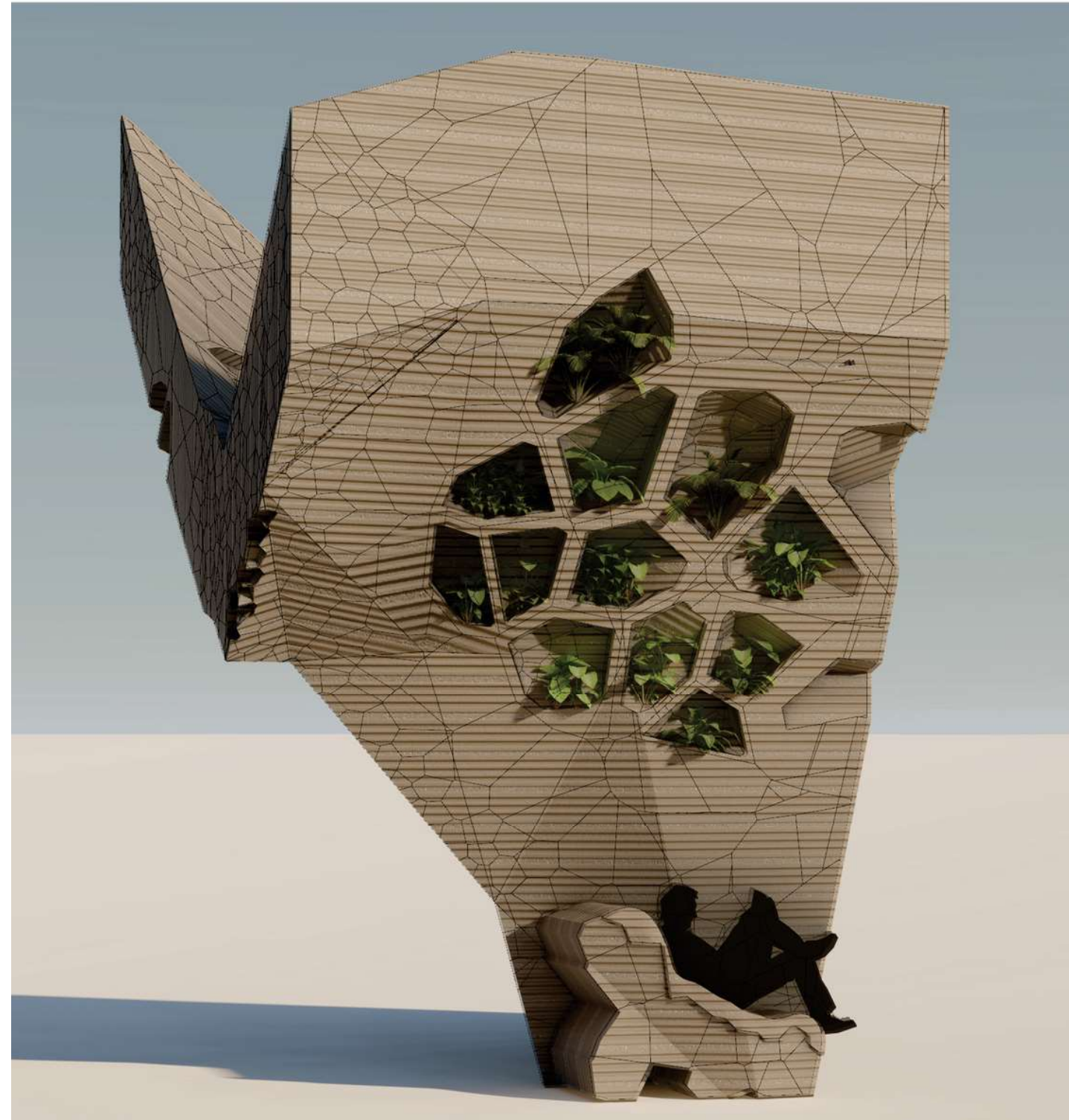
ACOUSTICS





FURNITURE





Seating Option



LIGHTING

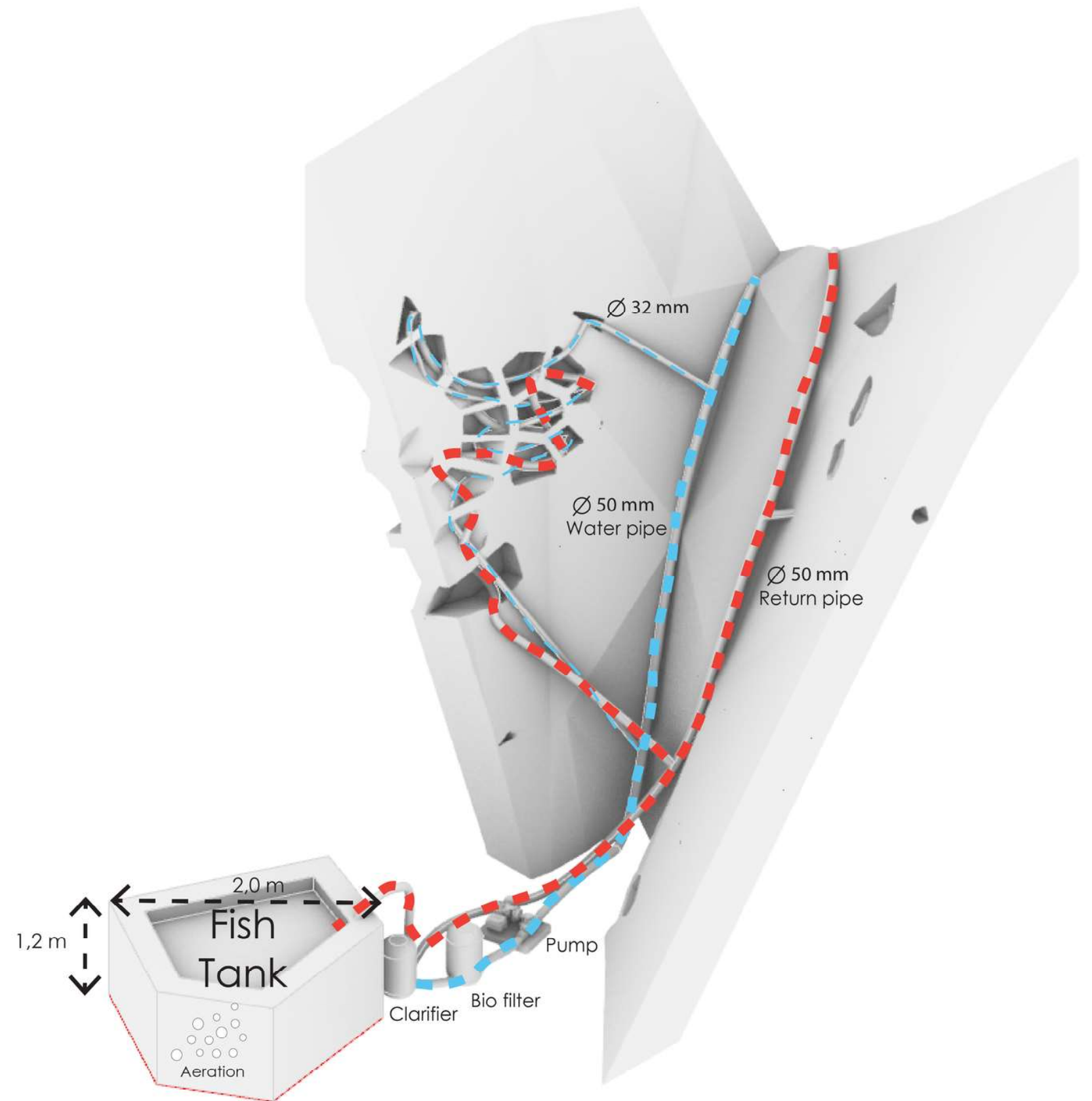


Lighting integration & shelves



**INCORPORATING
LIFE SUPPORT
SYSTEMS**

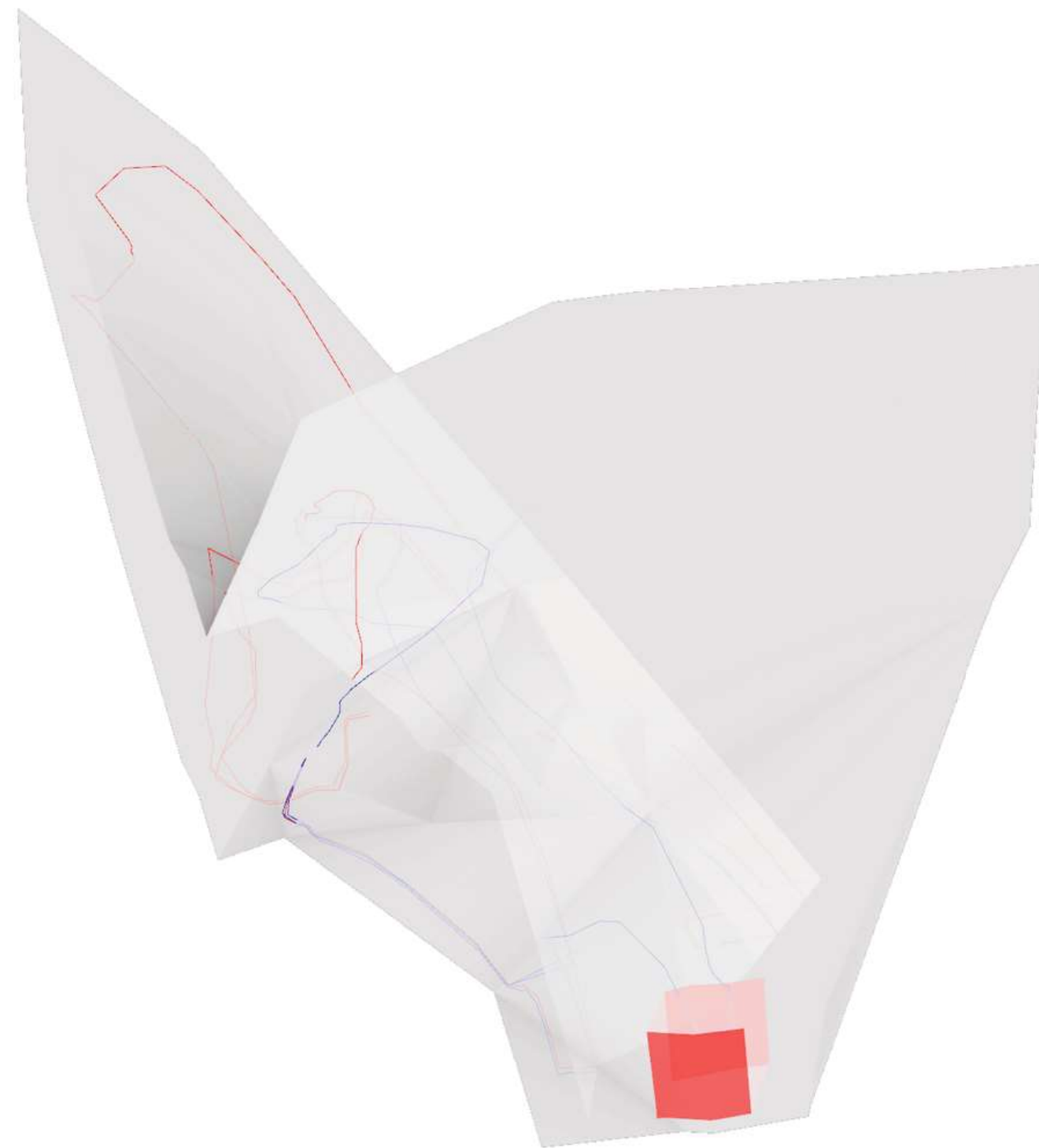
AQUAPONICS PRINCIPLE



1:1 FRAGMENT

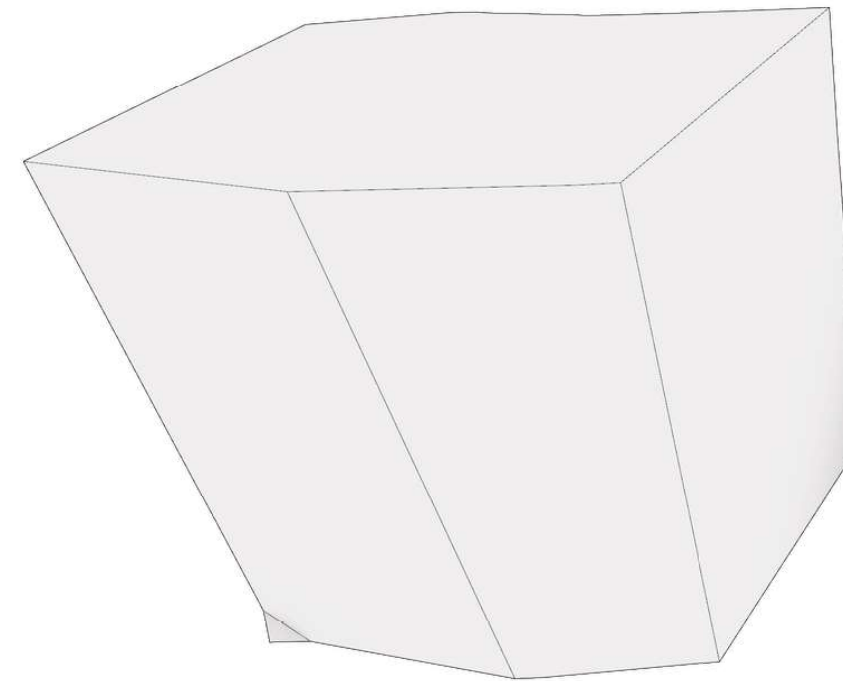
SELECTION

We choose a part of the fragment for the production of 1:1 prototype.



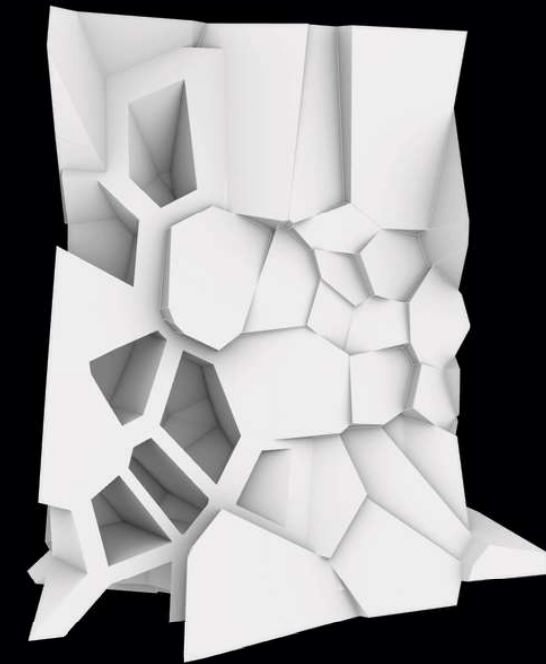
SELECTION

The following fragment has interesting lean to it which allows to test the principle of leaning walls.



PROTOTYPE 1

First prototype create geometry hard to 3D print. The amount of aquaponics was also problematic as they overtook the geometry and hence eliminated acoustics in some parts. Our aim of the exercise was to create a script that allows for a lot of control over the geometry - therefore we set goals and continued working. It performs well compared to initial geometry.

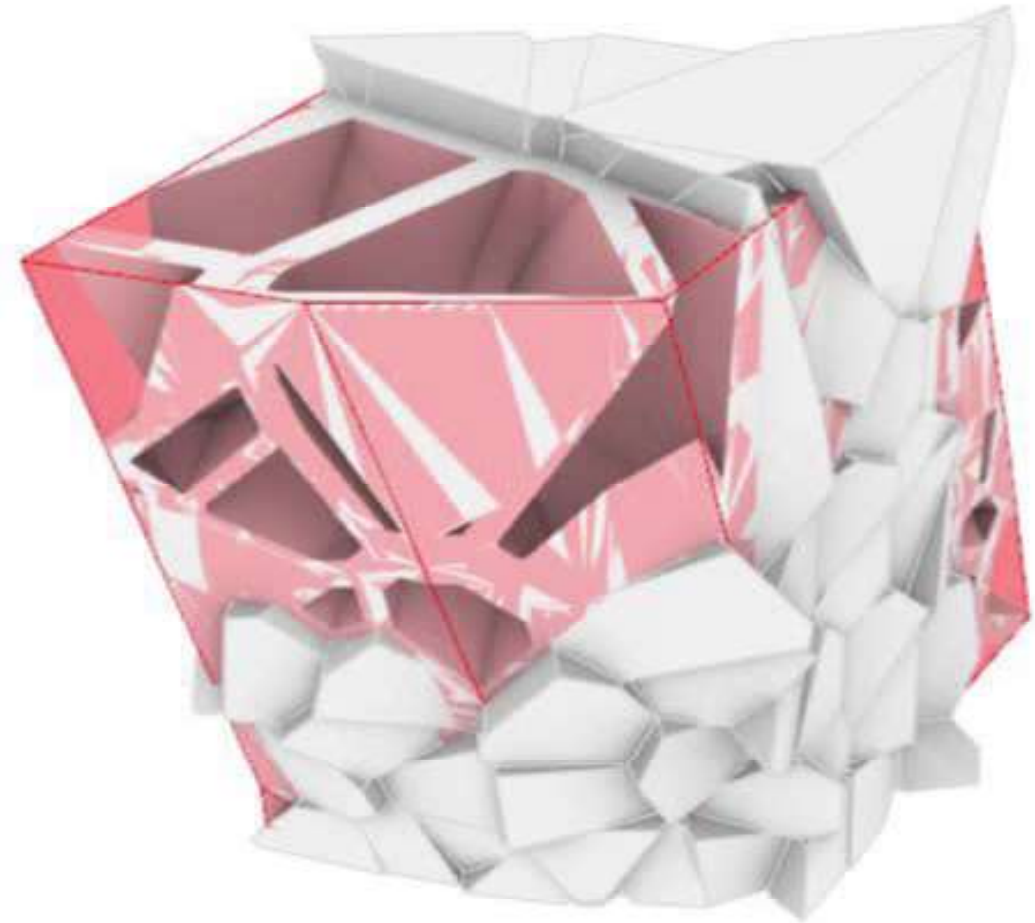


FRONT



BACK

PROTOTYPE 1



PROTOTYPE 2

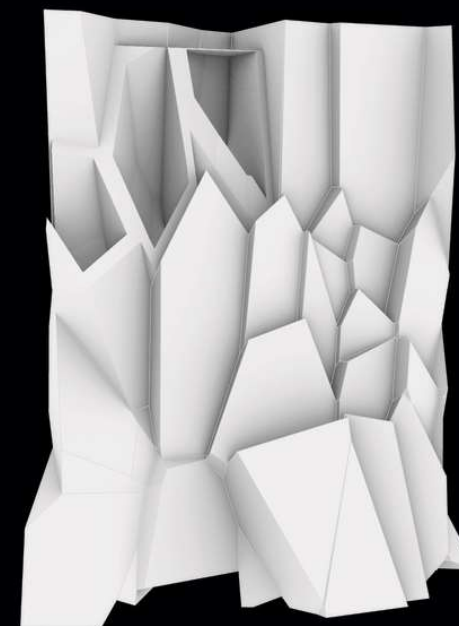
The prototype creates the best aquaponics. However the front faces are unprintable as they hang in the air.

This we had to change it. Also the smallest voronoi here are approximately 16cm where we aim for 4 cm.

Thus we further improve.

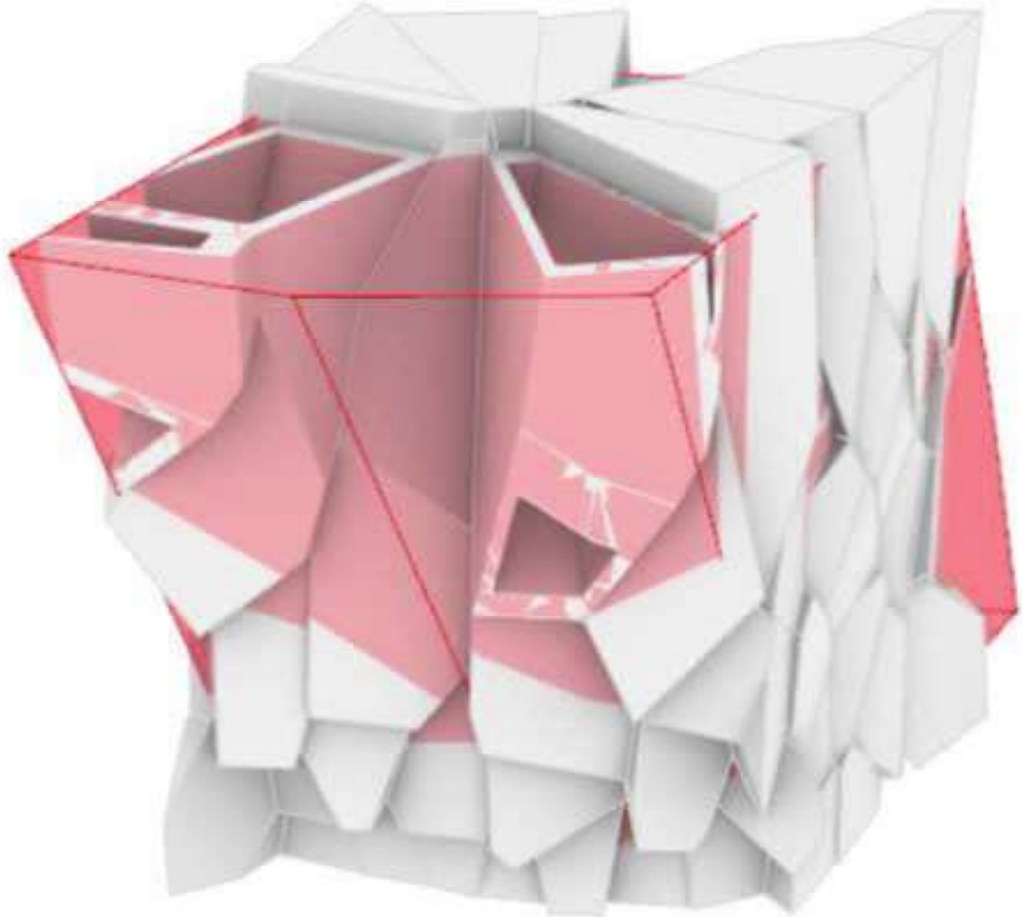


FRONT



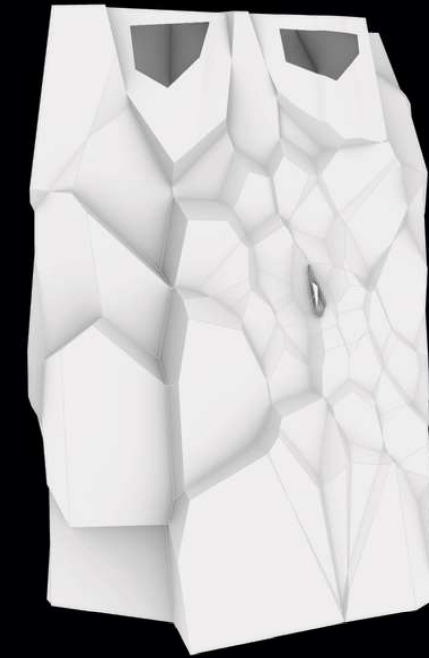
BACK

PROTOTYPE 2



PROTOTYPE 3

We managed to get a selection of smaller voronoi. The geometry is also more printable. Yet, too much deviation from the initial shape happens therefore it has to be fixed.

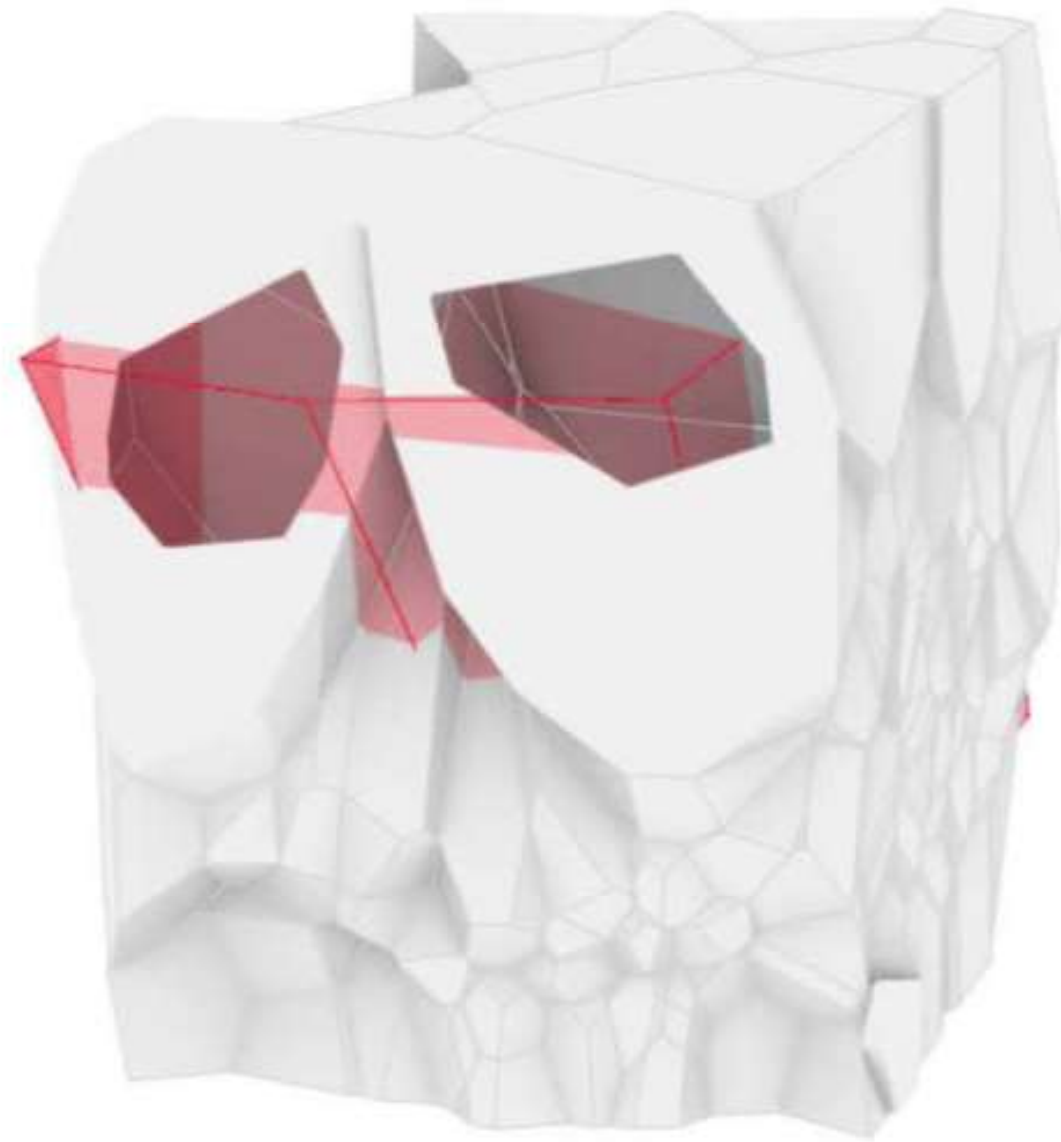


FRONT



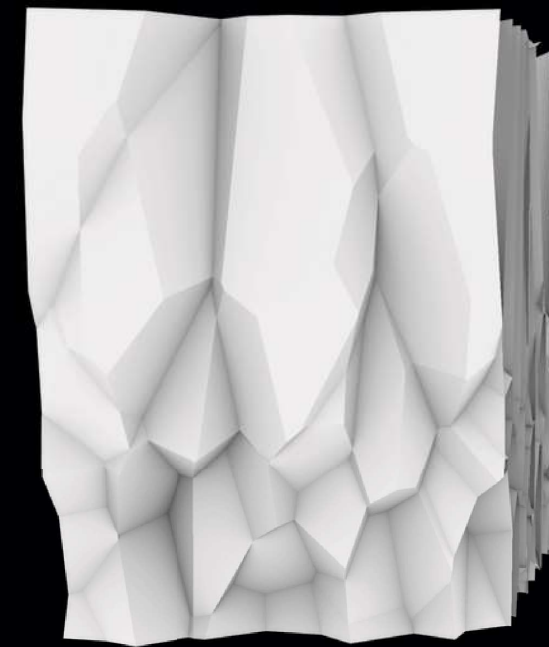
BACK

PROTOTYPE 3

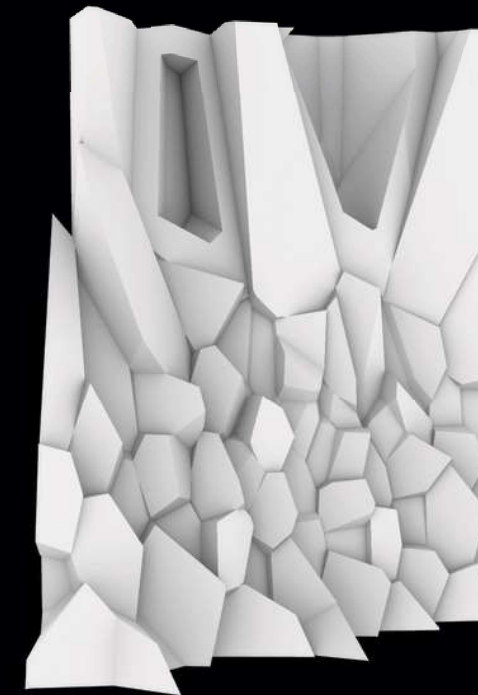


PROTOTYPE 4

Almost perfect geometry but there is significant deviation from the initial shape at the front top part. The back is precisely what we wanted. That is a lot of small voronoi, two aquaponics and gradient from the stress lines.

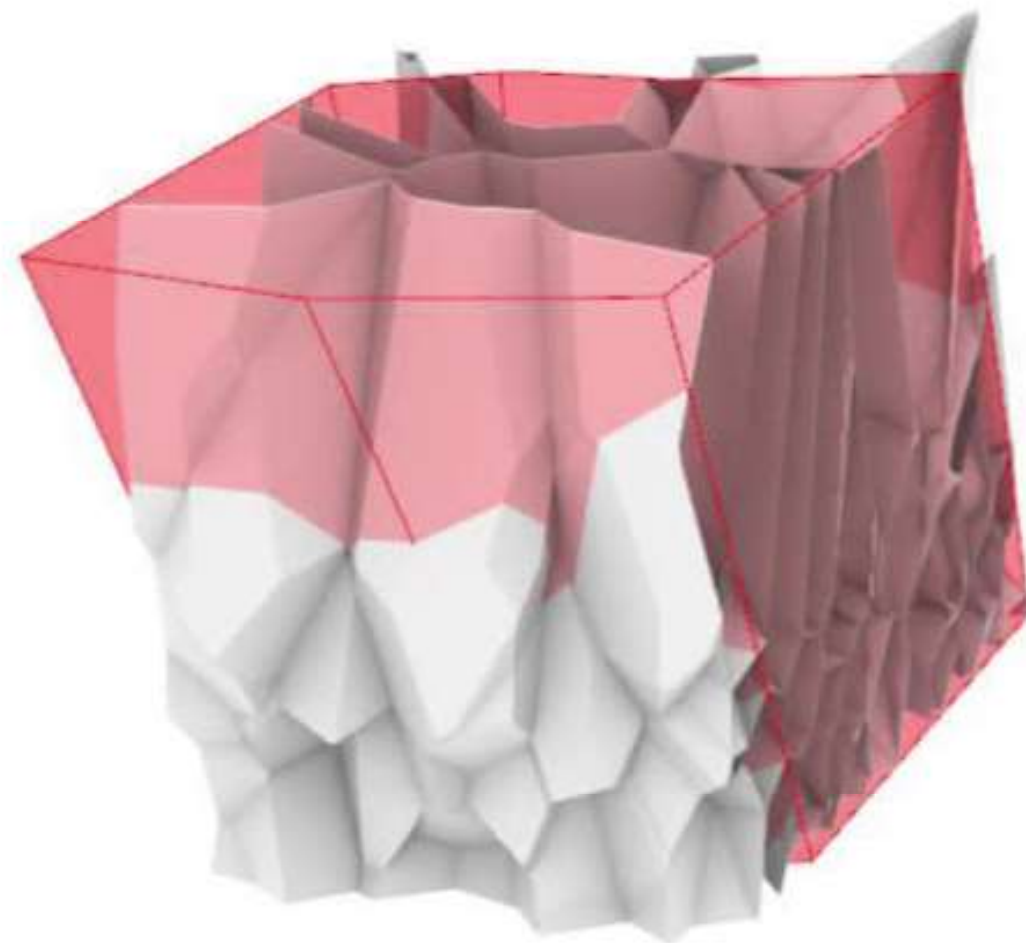


FRONT



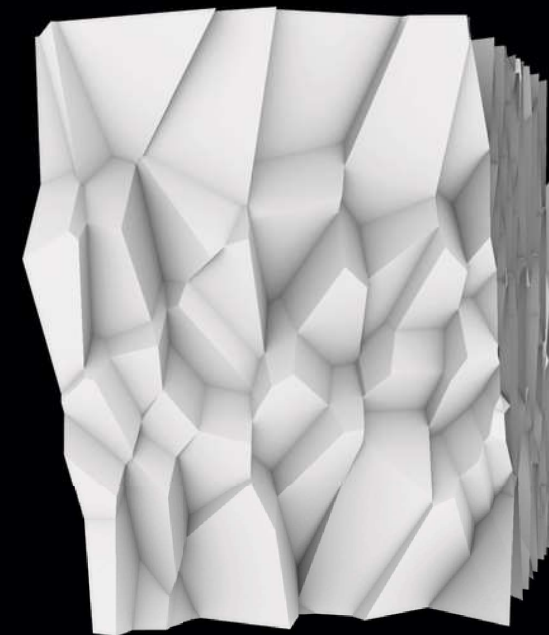
BACK

PROTOTYPE 4

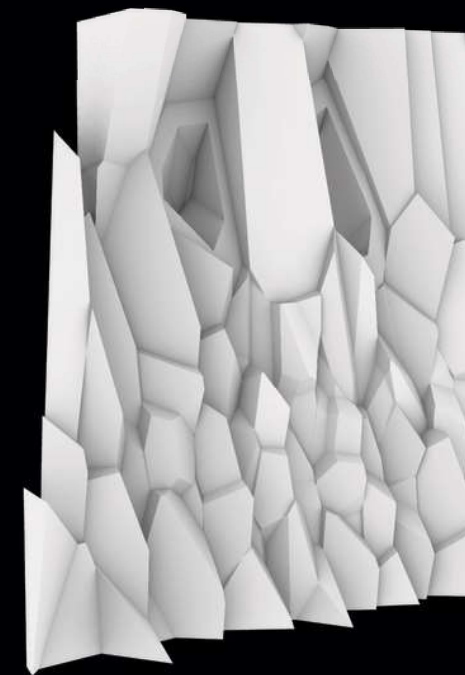


PROTOTYPE 5

We are very proud of this fragment. We were able to achieve it with little manual adjustments. The front texture is also printable and it has been fixed at the script level which means that the principle is scalable and can be applied on other fragments too.

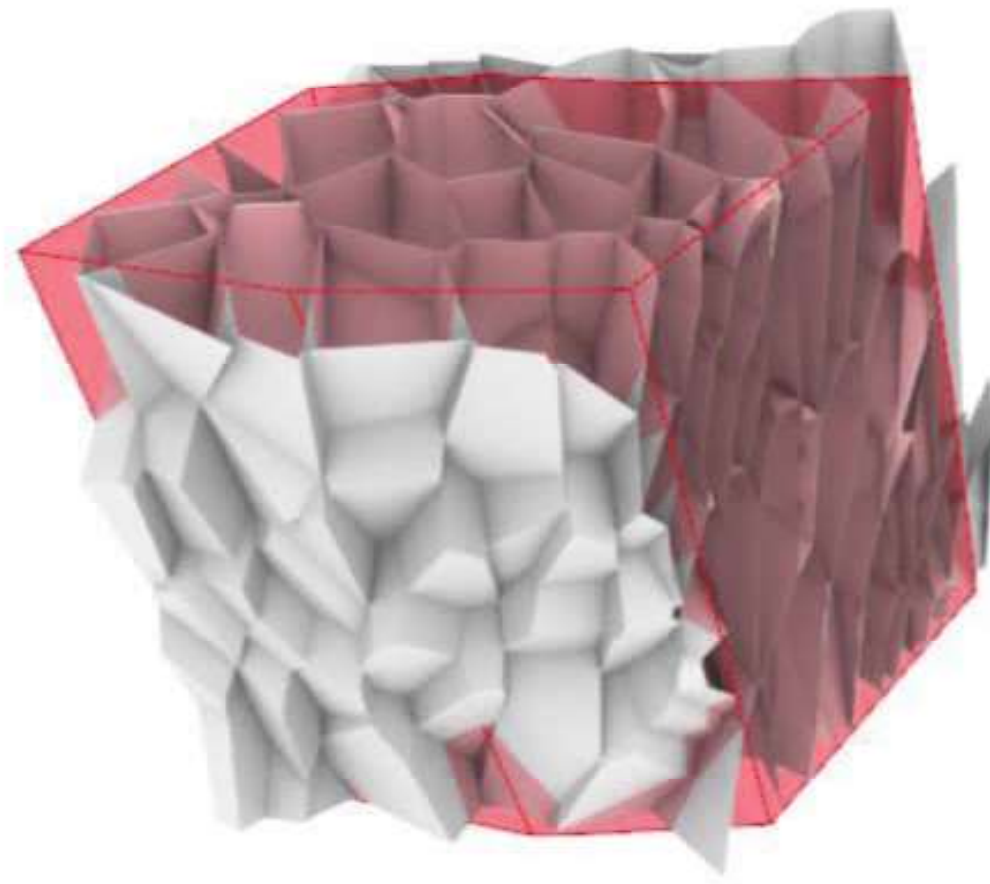


FRONT



BACK

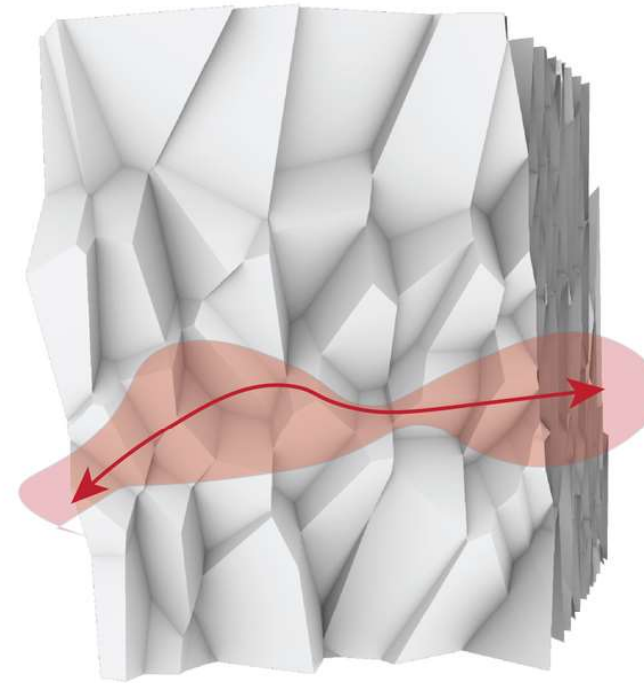
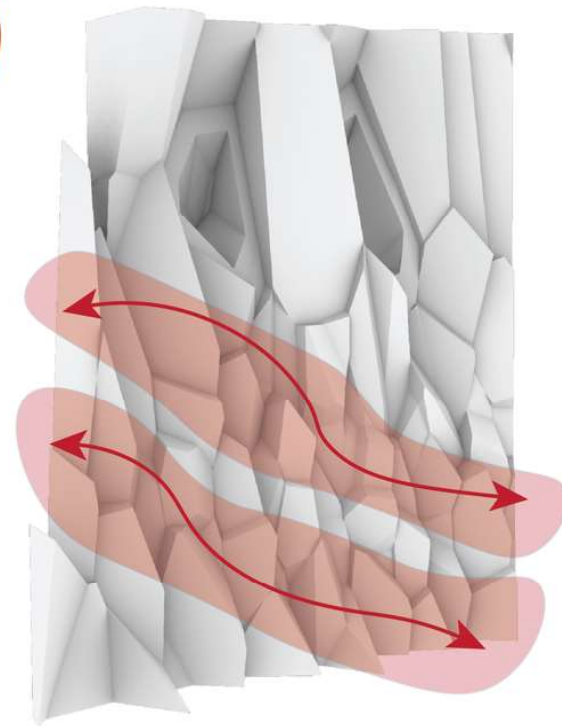
PROTOTYPE 5



STRUCTURE



STRUCTURE



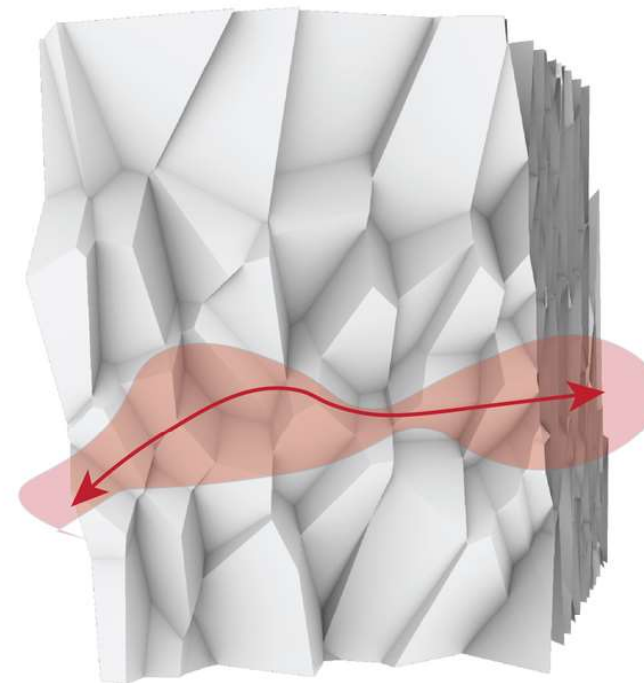
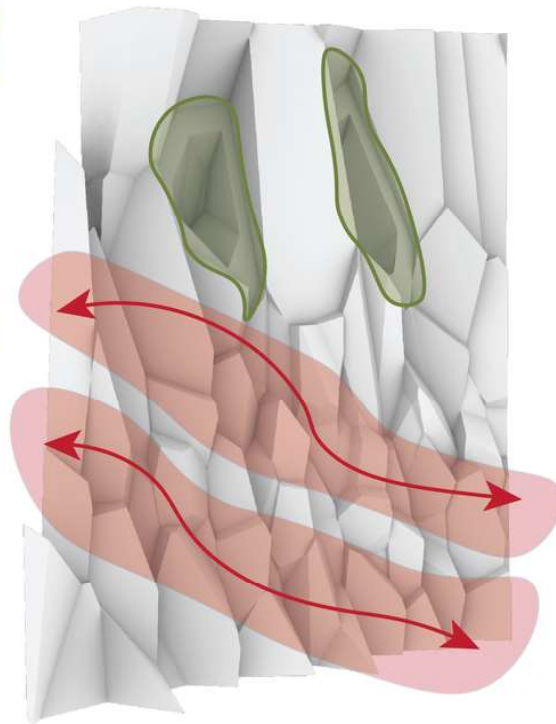
AQUAPONICS



STRUCTURE



AQUAPONICS



ACUSTICS



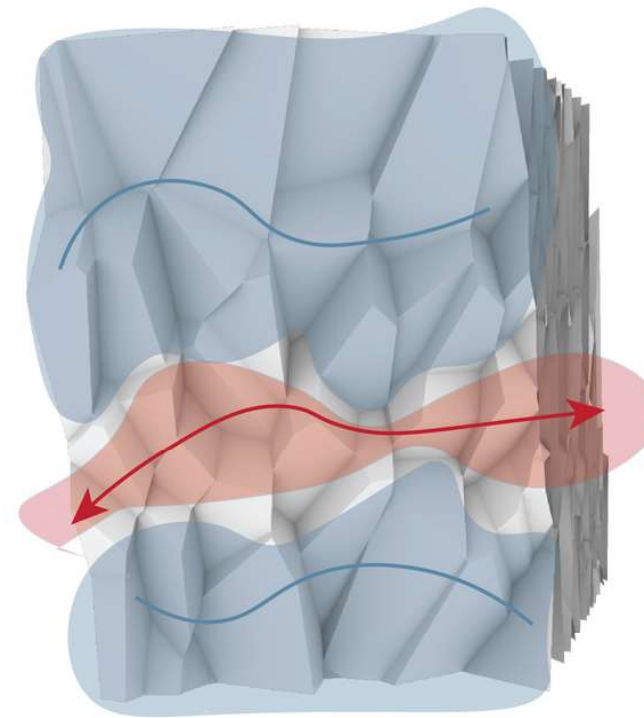
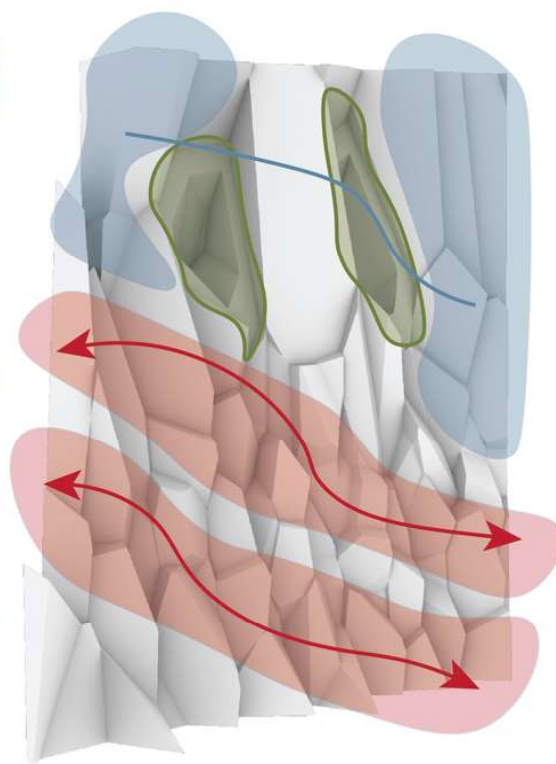
STRUCTURE

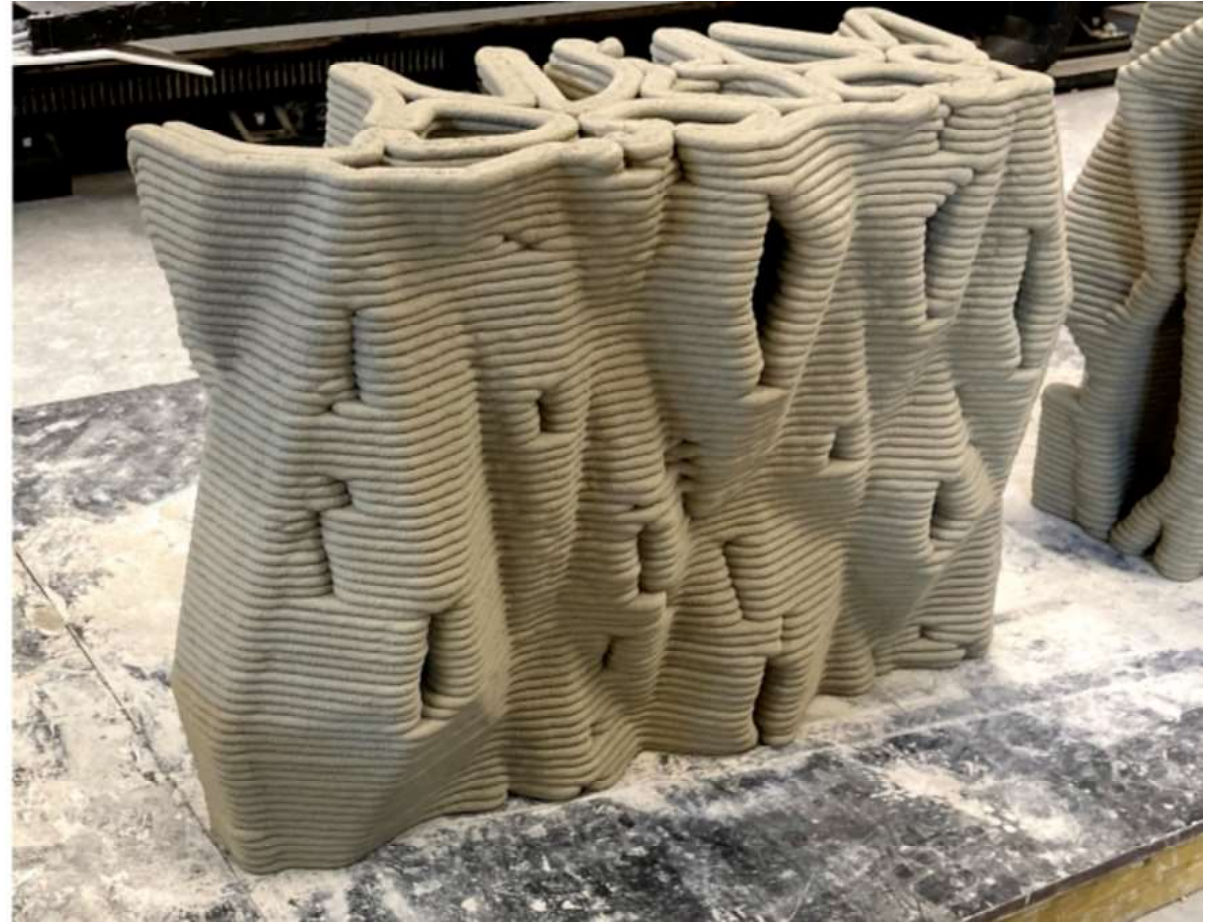
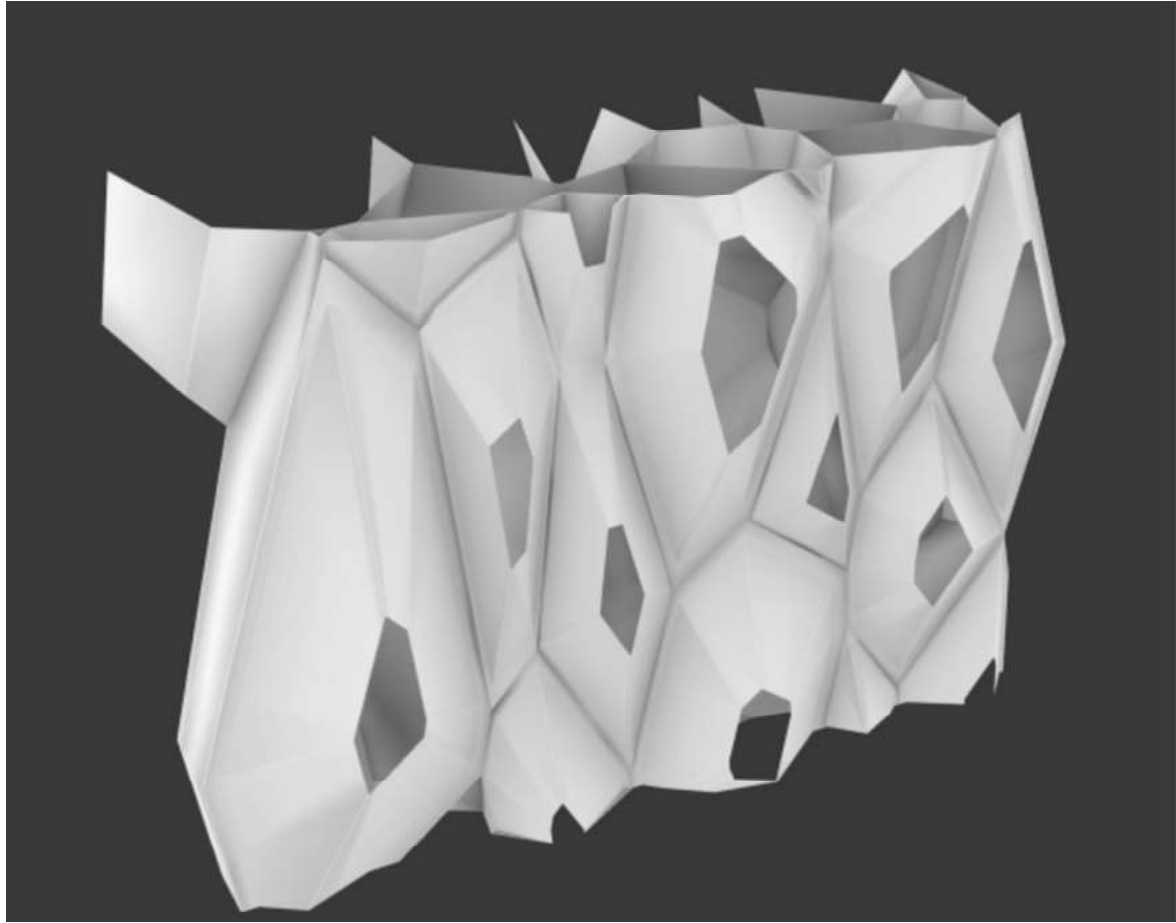


AQUAPONICS



ACOUSTICS

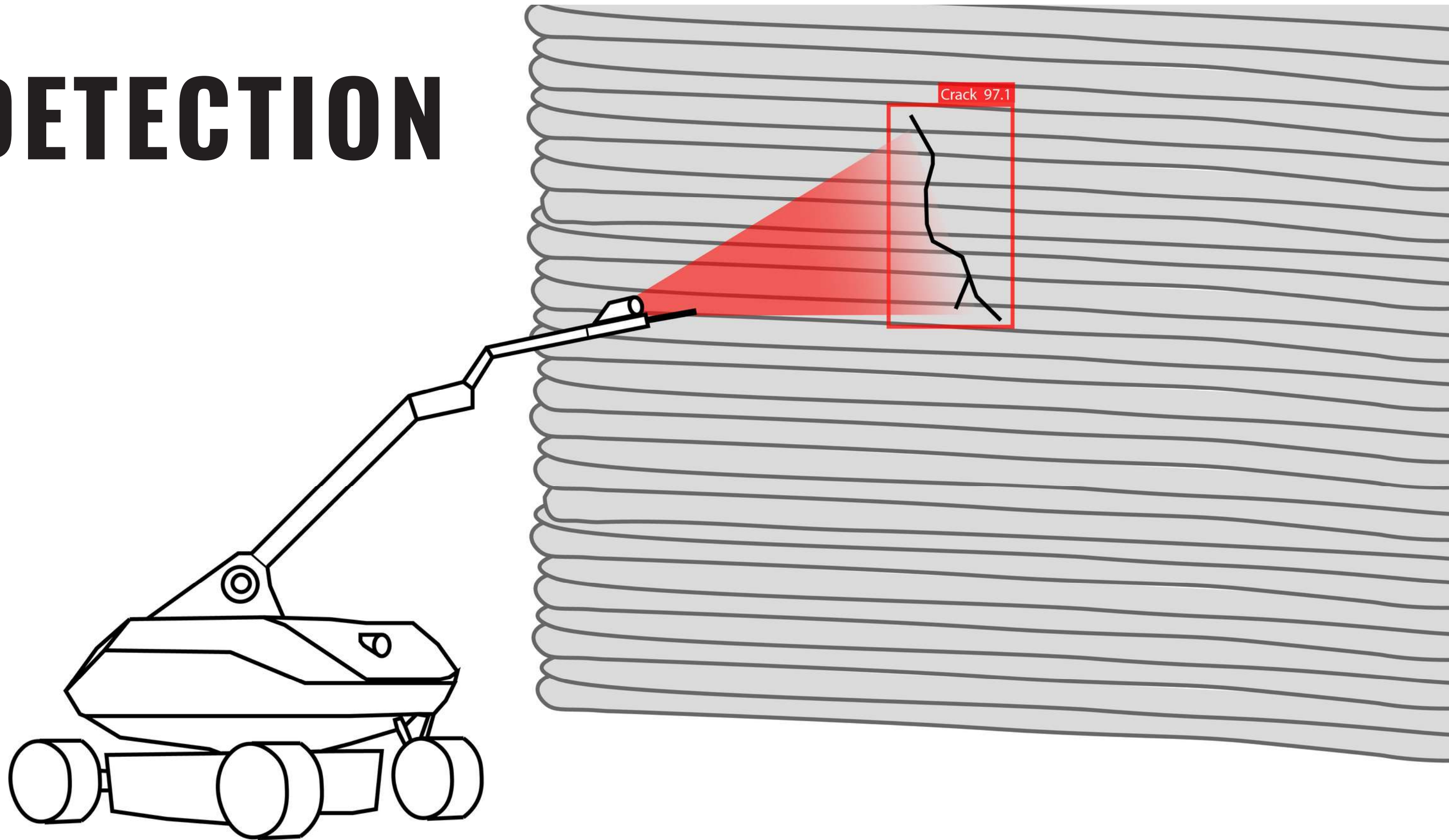




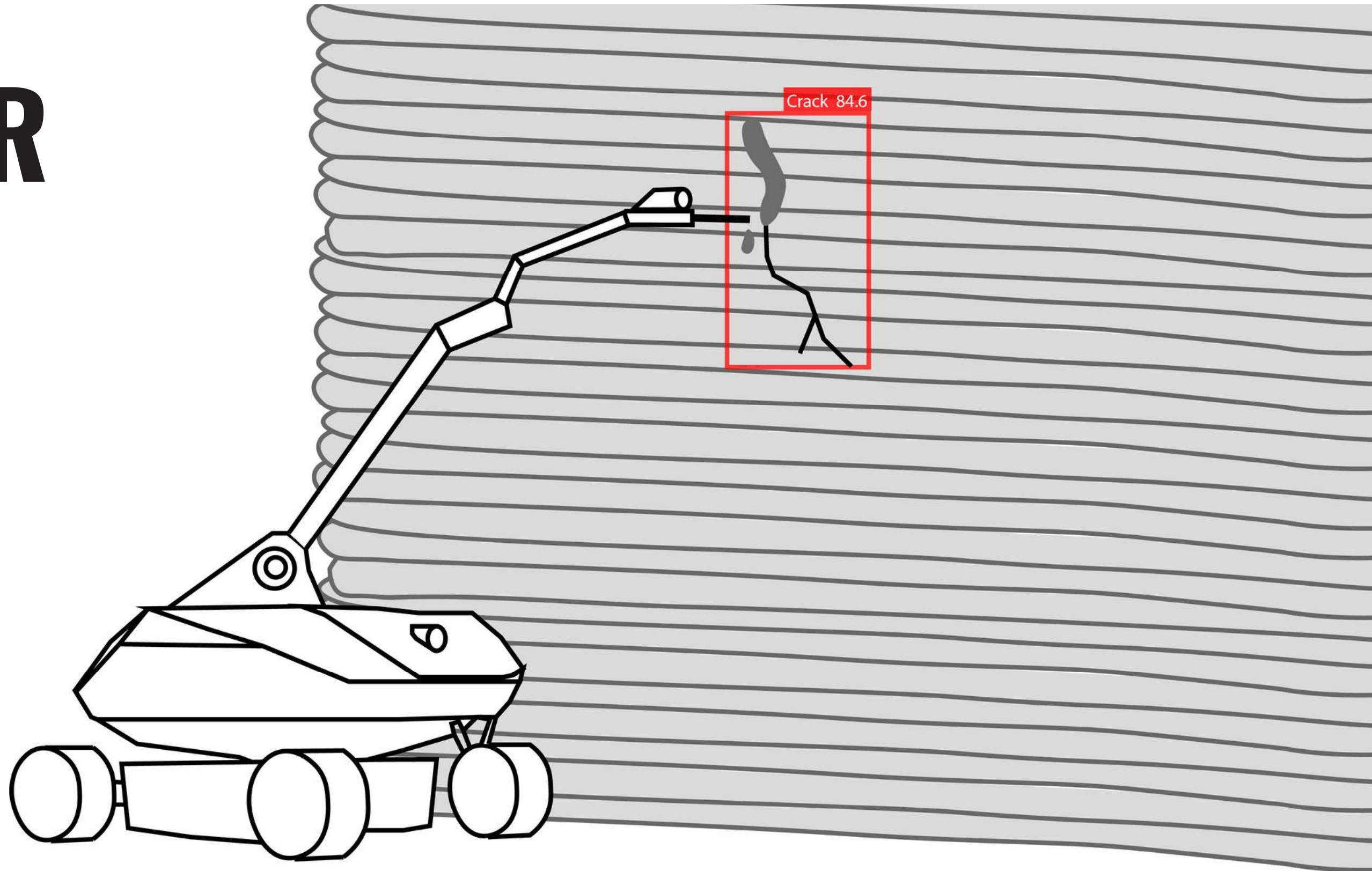
MAINTANANCE & PROTECTION

CV + HRI

DETECTION



REPAIR



yolov7 Public

Watch 98

Fork 3.4k

Star 10.5k

main 9 branches 1 tag

Go to file

Add file

Code

WongKinYiu Update README.md	84932d7 last month	🕒 132 commits
📁 cfg	main code	last year
📁 data	Added param loss_ota for hyp.yaml, to disable OTA for faster training	10 months ago
📁 deploy/triton-inference-server	Update README.md (#850)	8 months ago
📁 figure	Add files via upload	last month
📁 inference/images	Support dynamic batch for TensorRT and onnxruntime (#329)	last year
📁 models	main code	9 months ago
📁 paper	main code	last year
📁 scripts	main code	last year
📁 tools	updated Reparameterization weight path & added steps for doing Repa...	9 months ago
📁 utils	utils/loss.py minor bug fix (#1344)	5 months ago
📄 .gitignore	update gitignore (#461)	10 months ago

About

Implementation of paper - YOLOv7: Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors

- pytorch
- darknet
- yolov3
- yolov4
- scaled-yolov4
- yolor
- yolov7

- 📖 Readme
- 📄 GPL-3.0 license
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- 🌟 10.5k stars
- 👁 98 watching
- 🍴 3.4k forks
- 📧 Report repository

Releases

🏷 1 tags

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crack
Instance Segmentation

Overview

Images 1551

Dataset 2

Model

API Docs

2022-09-29 2:14pm

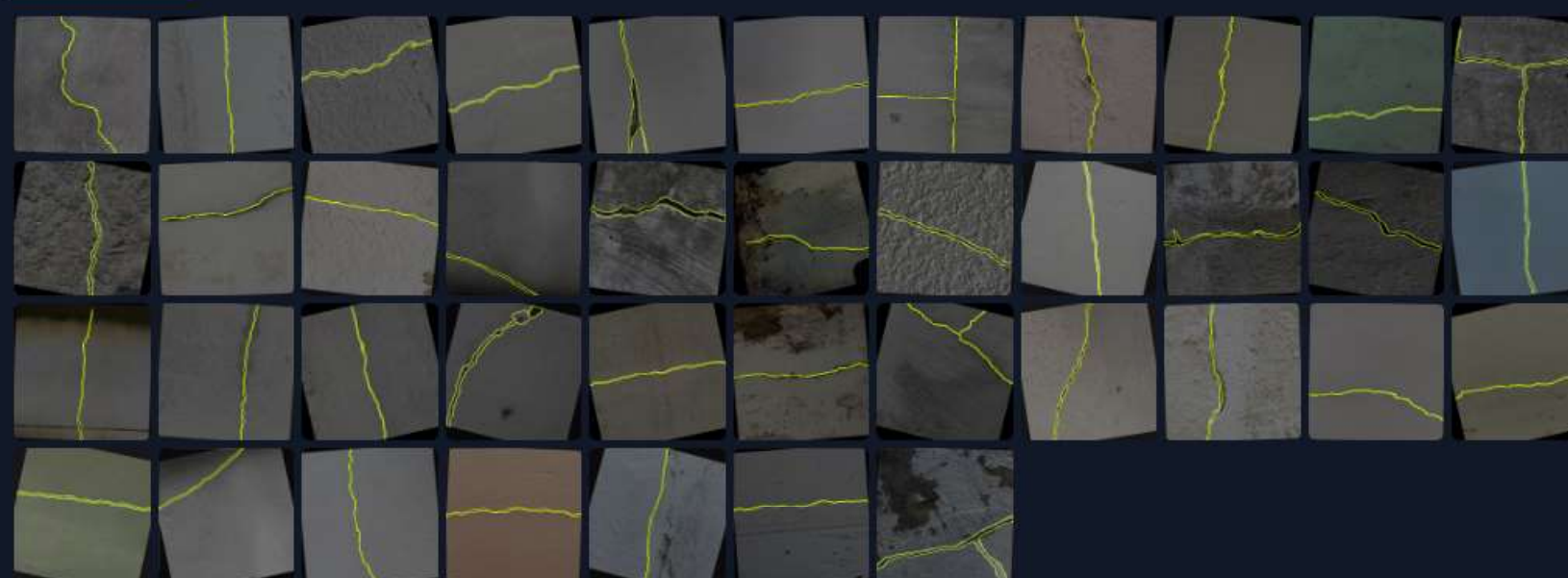
v2 Sep 29, 2022

2022-09-02 11:09pm

v1 Sep 2, 2022

Download Dataset

Train 3717 Valid 200 Test 112



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Crack Detector

yolo7-seg

Concrete damag...

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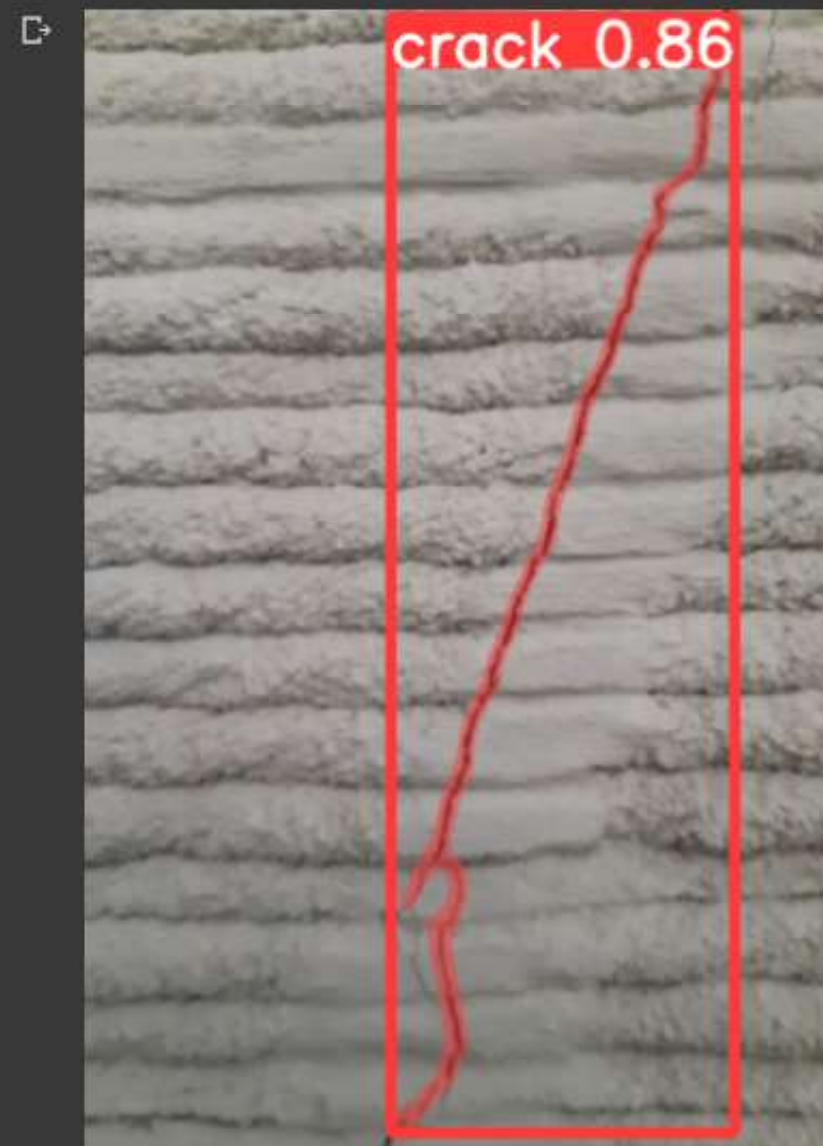
400 img

```
[ ] Using 2 dataloader workers
Logging results to runs/train-seg/custom
Starting training for 10 epochs...
```

Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
0/9	12.8G	0.07064	0.04226	0.02526	0	15	640: 100% 233/233	[03:43<00:00,	1.04it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.18it/s]
	all	200	249	0.733	0.551	0.575	0.226	0.572	0.478	0.365	0.0924		
Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
1/9	12.3G	0.04831	0.02473	0.01802	0	13	640: 100% 233/233	[03:36<00:00,	1.08it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.33it/s]
	all	200	249	0.627	0.586	0.6	0.297	0.48	0.474	0.377	0.103		
Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
2/9	12.3G	0.04559	0.02405	0.01454	0	7	640: 100% 233/233	[03:34<00:00,	1.09it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.38it/s]
	all	200	249	0.444	0.506	0.417	0.189	0.34	0.393	0.228	0.0616		
Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
3/9	12.3G	0.04035	0.02363	0.01382	0	15	640: 100% 233/233	[03:34<00:00,	1.09it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.39it/s]
	all	200	249	0.549	0.636	0.618	0.343	0.445	0.558	0.429	0.127		
Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
4/9	12.3G	0.03702	0.02345	0.01295	0	15	640: 100% 233/233	[03:34<00:00,	1.09it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.39it/s]
	all	200	249	0.778	0.667	0.667	0.361	0.663	0.558	0.467	0.14		
Epoch	GPU_mem	box_loss	seg_loss	obj_loss	cls_loss	Instances	Size						
5/9	12.3G	0.03239	0.02297	0.01272	0	10	640: 100% 233/233	[03:34<00:00,	1.09it/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95)	Mask(P	R	mAP50	mAP50-95): 100% 7/7	[00:05<00:00,	1.40it/s]
	all	200	249	0.753	0.622	0.686	0.406	0.58	0.546	0.441	0.136		

```
import glob
from IPython.display import Image, display

for imageName in glob.glob('runs/predict-seg/exp32/print_test11.png')[:2]:
    display(Image(filename=imageName))
    print("\n")
```



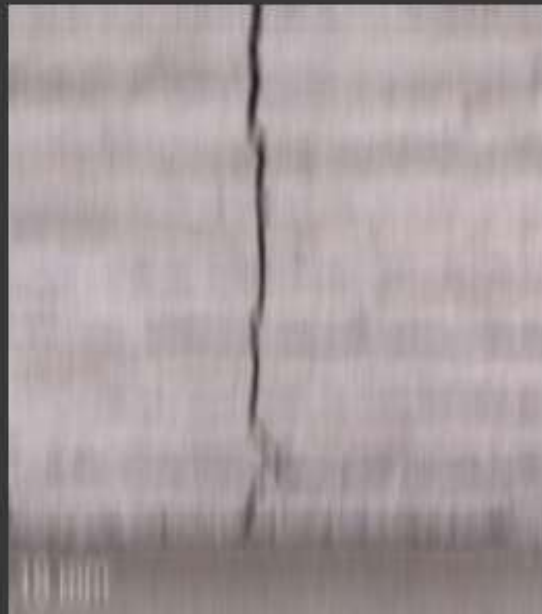
```
▶ # Fill the middle row with ones
kernel_v[:, int((kernel_size - 1)/2)] = np.ones(kernel_size)

# Normalize
kernel_v /= kernel_size

# Apply the vertical kernel
vertical_mb = cv2.filter2D(img, -1, kernel_v)

# Save the outputs
cv2.imwrite('test_vertical.jpg', vertical_mb)

# Print out
img2 = cv2.imread('test_vertical.jpg', cv2.IMREAD_UNCHANGED)
cv2_imshow(img2)
```



✓
0s



```
import glob
from IPython.display import Image, display

for imageName in glob.glob('runs/predict-seg/exp9/print_test1.jpg')[:2]:
    display(Image(filename=imageName))
    print("\n")
```



✓
0s



```
import glob
from IPython.display import Image, display

for imageName in glob.glob('runs/predict-seg/exp25/print_test7.png')[:2]:
    display(Image(filename=imageName))
    print("\n")
```





Photo by [Martijn Baudoin](#) on [Unsplash](#)

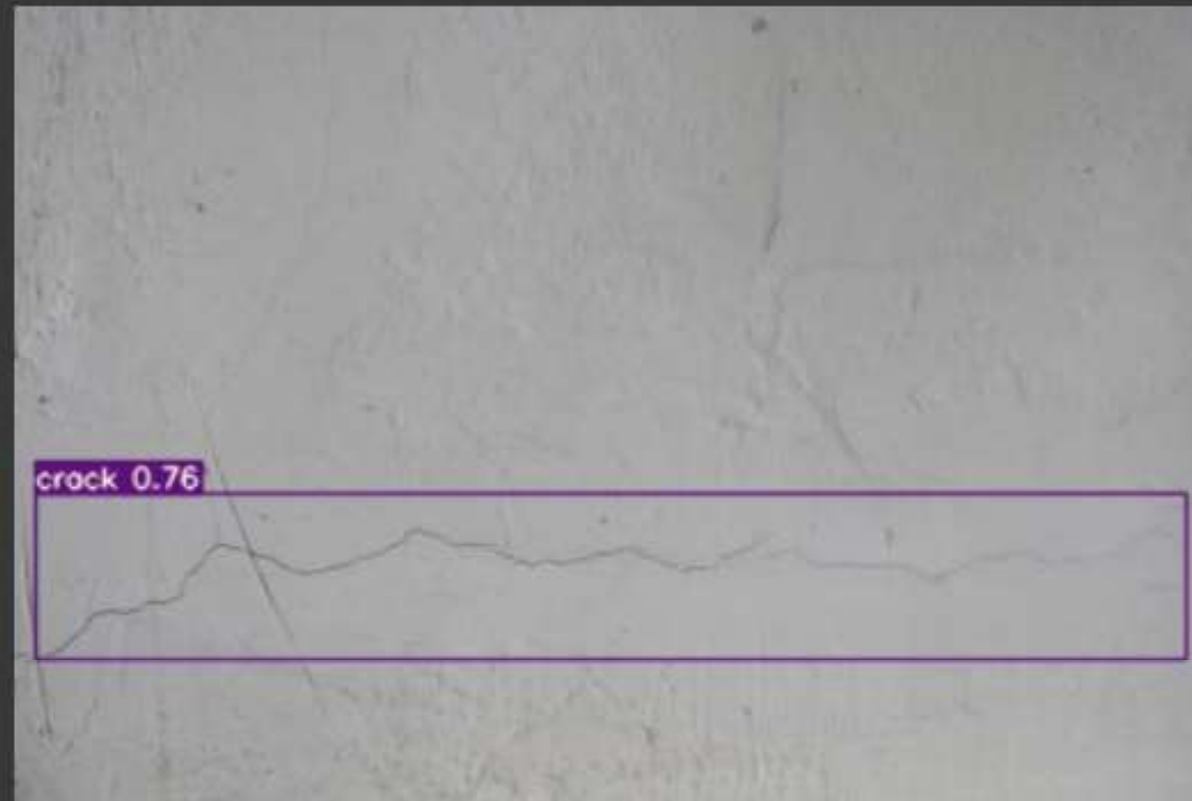
Custom YOLOv7 Object Detection with TensorFlow.js

Training a custom YOLOv7 model in PyTorch and converting it to TensorFlow.js for real-time offline detection on the browser

```
[ ] %cd /content/yolov7
!python train.py --epochs 64 --workers 4 --device 0 --batch-size 32 \
--data /content/yolov7/data.yaml --img 640 640 --cfg /content/yolov7/cfg/training/yolov7_crack-tiny.yaml \
--weights 'yolov7-tiny.pt' --name yolov7_tiny_crack --hyp data/hyp.scratch.tiny.yaml
```

Image sizes 640 train, 640 test
Using 2 dataloader workers
Logging results to runs/train/yolov7_tiny_crack
Starting training for 64 epochs...

Epoch	gpu_mem	box	obj	cls	total	labels	img_size
0/63	0.47G	0.07802	0.01106	0	0.08908	14	640: 100% 117/117 [03:14<00:00, 1.66s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 4/4 [00:04<00:00, 1.16s/it]
	all	200	249		0.0289	0.1	0.0174 0.00411
1/63	5.11G	0.05707	0.01035	0	0.06742	10	640: 100% 117/117 [02:59<00:00, 1.53s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 4/4 [00:03<00:00, 1.15it/s]
	all	200	249		0.28	0.341	0.185 0.0428
2/63	6.26G	0.04459	0.009907	0	0.0545	28	640: 100% 117/117 [02:56<00:00, 1.51s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 4/4 [00:04<00:00, 1.05s/it]
	all	200	249		0.604	0.465	0.497 0.18
3/63	6.26G	0.03898	0.009411	0	0.04839	11	640: 100% 117/117 [02:56<00:00, 1.51s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 4/4 [00:03<00:00, 1.17it/s]
	all	200	249		0.604	0.588	0.559 0.227
4/63	6.27G	0.03961	0.009165	0	0.04878	16	640: 100% 117/117 [03:06<00:00, 1.59s/it]



▼ Converting to TFJS

Pytorch to ONNX with NMS (and inference)

```
[ ] !python export.py --weights runs/train/yolov7_tiny_crack/weights/yolov7-tiny.pt \
    --grid \
    --topk-all 100 --iou-thres 0.65 --conf-thres 0.35 \
    --img-size 640 640 --max-wh 640 # For onnxruntime, you need to specify this value as an integer, when it is 0 it means agnostic NMS,
    # otherwise it is non-agnostic NMS
```

```
Namespace(weights='runs/train/yolov7_tiny_crack/weights/yolov7-tiny.pt', img_size=[640, 640], batch_size=1, dynamic=False, dynamic_batch=False, grid=True, end2end=False, max_wh=640, tc
YOLOR 🚀 v0.1-126-g84932d7 torch 2.0.1+cu118 CPU
```

Fusing layers...

```
/usr/local/lib/python3.10/dist-packages/torch/functional.py:504: UserWarning: torch.meshgrid: in an upcoming release, it will be required to pass the indexing argument. (Triggered inte
return _VF.meshgrid(tensors, **kwargs) # type: ignore[attr-defined]
```

Model Summary: 200 layers, 6006646 parameters, 6006646 gradients, 13.0 GFLOPS

Starting TorchScript export with torch 2.0.1+cu118...

```
/content/yolov7/models/yolo.py:52: TracerWarning: Converting a tensor to a Python boolean might cause the trace to be incorrect. We can't record the data flow of Python values, so this
if self.grid[i].shape[2:4] != x[i].shape[2:4]:
```

TorchScript export success, saved as runs/train/yolov7_tiny_crack/weights/yolov7-tiny.torchscript.pt

CoreML export failure: No module named 'coremltools'

Starting TorchScript-Lite export with torch 2.0.1+cu118...

TorchScript-Lite export success, saved as runs/train/yolov7_tiny_crack/weights/yolov7-tiny.torchscript.ptl

Starting ONNX export with onnx 1.14.0...

```
/content/yolov7/models/yolo.py:582: TracerWarning: Converting a tensor to a Python boolean might cause the trace to be incorrect. We can't record the data flow of Python values, so thi
if arguments:
```


crack-detection-deploy

Git Repository

Domains

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The deployment that is available to your visitors.

Build Logs

Runtime Logs

↻ Instant Rollback



DEPLOYMENT

crack-detection-deploy-dr2flg11i-julialewandowska.vercel.app

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crack-detection-deploy.vercel.app ↗ +2

STATUS

● **Ready** 10d ago by julialewandowska 🌸

CREATED ⓘ

SOURCE

🔗 main

🔗 227bbd6 Update README.md











