

A photograph of a modern building interior, likely a staircase or atrium. The walls are made of large, light-colored concrete panels. A series of glass railings with dark metal frames runs along the edge of a walkway. Sunlight streams in from a large window or glass wall on the right, creating strong shadows and highlights on the concrete and glass surfaces. The overall atmosphere is clean, minimalist, and architectural.

PORTFOLIO

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Cover Page

Photography work at 21_21 Design Sight

Tokyo, Japan

2017

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01 Machine Learning Generated Urban Blocks

Individual work: May 2021 - May 2022

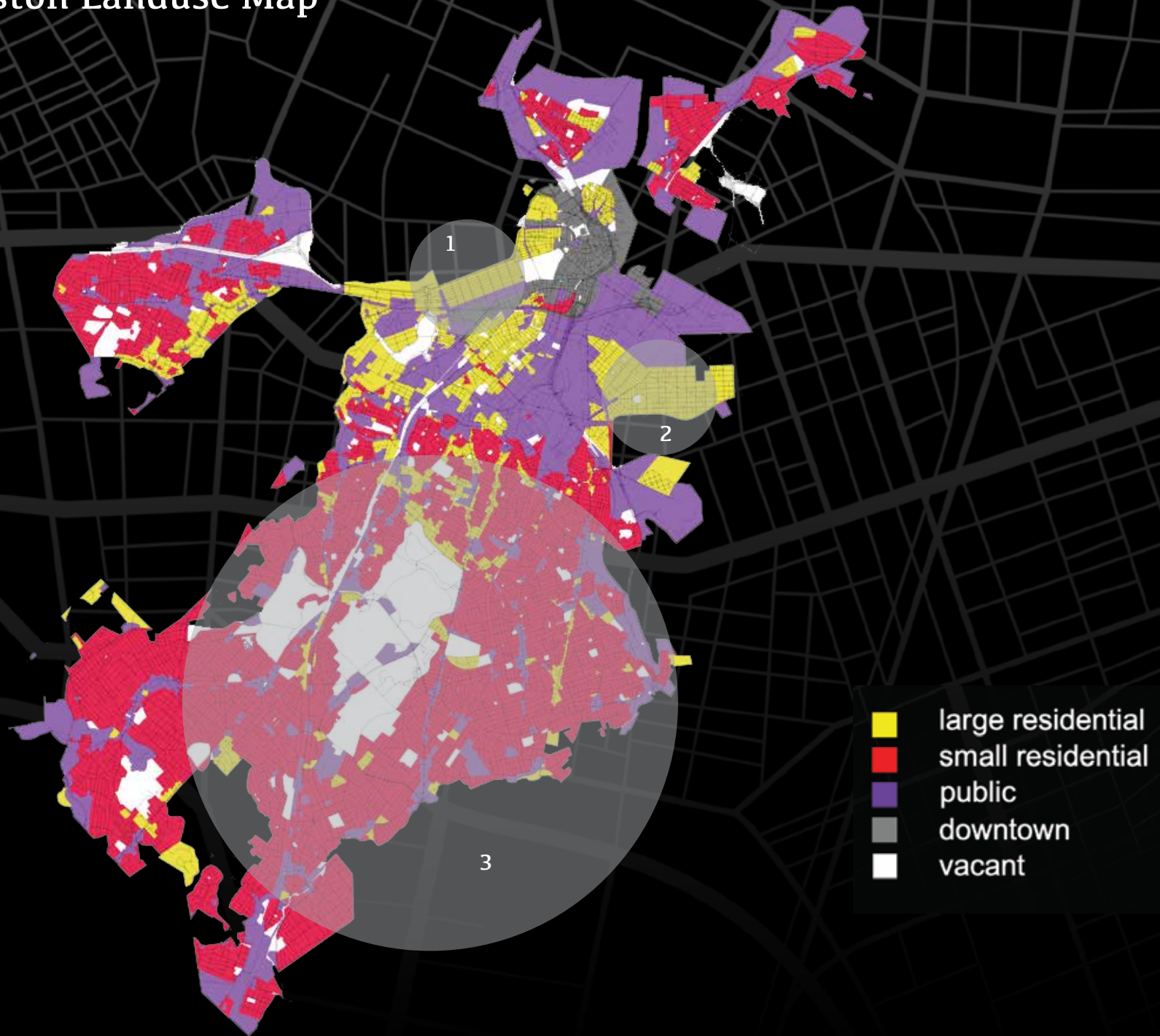
Supervisor: Prof Anand Bhojan

Layout generated using deep learning model & details procedually generated using blender.

This is my first attempt to examine the potential of deep learning model in understanding urban patterns. A modified version of CycleGAN was used to learn the building footprints in different zones of Boston.



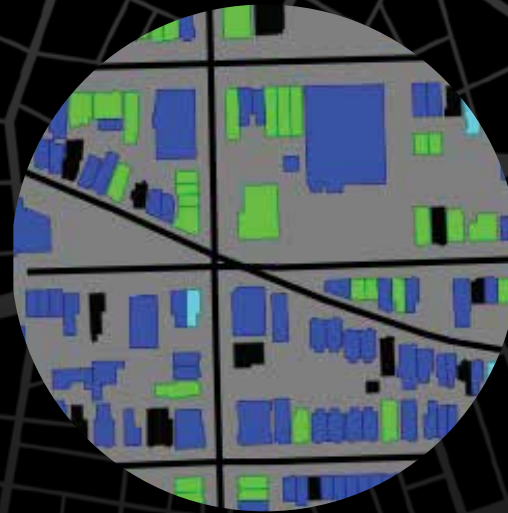
Case Study: Boston Landuse Map



In many city, buildings exhibit unique style that is intereting to model after.
The goal is to focus on regions where the houses exhibit clear pattern in terms of footprints and height.
I picked three categories as shown below. Varying heights are represented using different colors.



Region 1: Back Bay



Region 2: South Boston

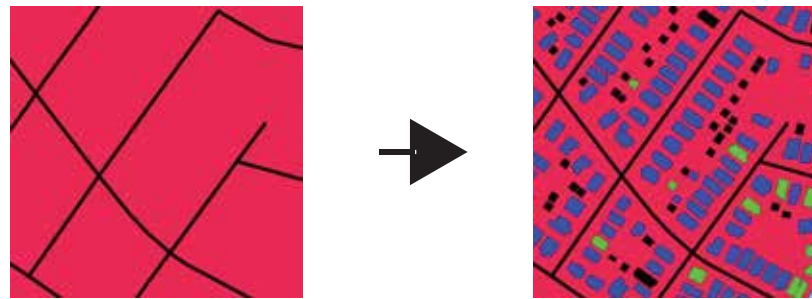


Region 3: A large areas of residential zones predominated with landed houses

Generative Adversarial Networks for Architects

A generative adversarial network consists of two parts: a generator and a discriminator.

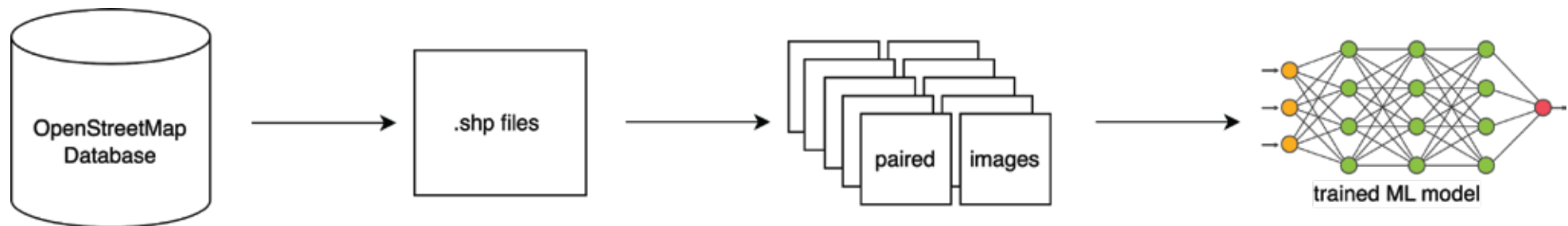
The generator learns the distribution in the training data. In this case, the generator learns the mapping from the roads & the land use information to the footprints and heights of the buildings.



The discriminator evaluates how likely the generated result is from the original distribution. The two are trained together in a competitive setting. Randomly initialised generator and discriminator will converge eventually.

Imagine the generator as an art forger who wishes to copy the style of certain painter, and the discriminator as an expert in identifying art forger. Although initially they may not be experts in their field, as they compete with each other, both grow in their expertise.

The Workflow of Machine Learning Pipeline - Model Preparation

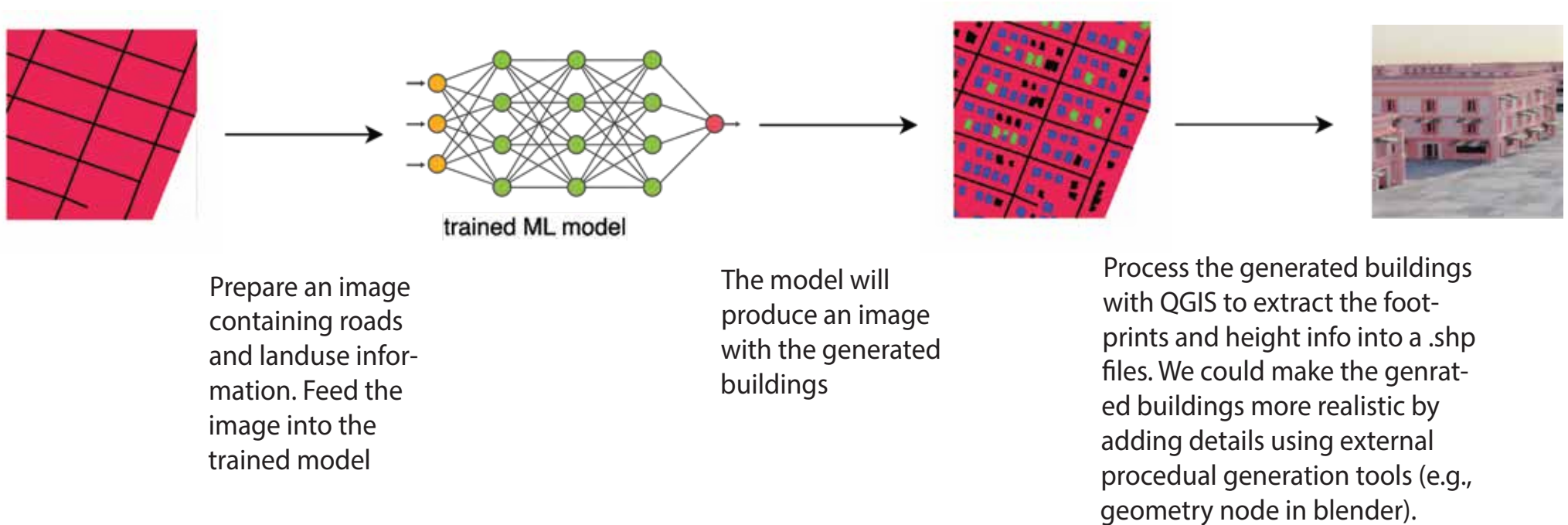


Retrieve the buildings, roads and landuse data from open source database such as OSM

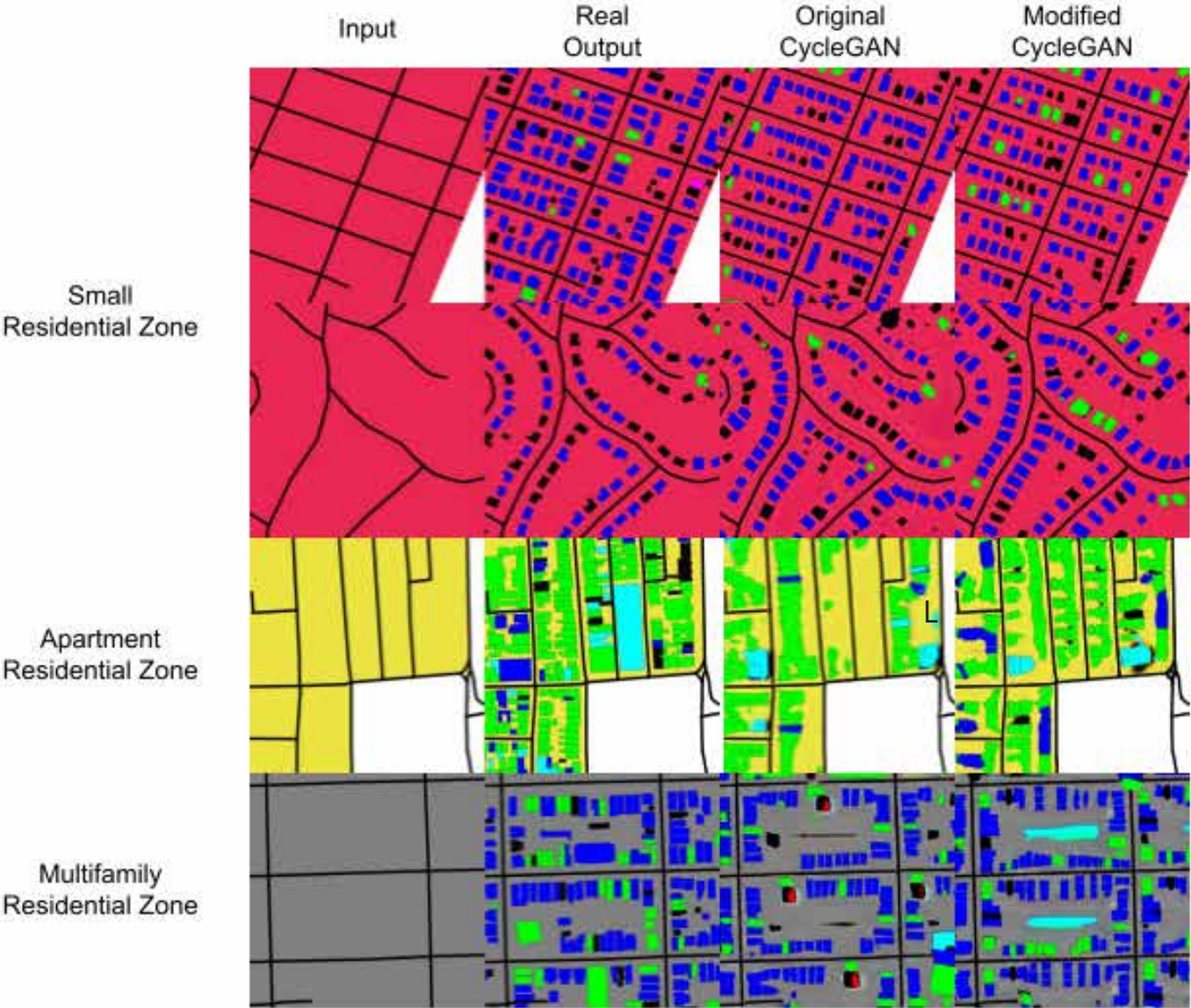
Process the .shp files into images using GIS tools (e.g., QGIS). Use different colors to present information so that the computer can understand.

Generate image pairs and feed it into a modified version of CycleGAN for training.

The Workflow of Machine Learning Pipeline - Model Usage



Results



Evaluation Metrics

Table 6.1: Evaluation on the distance from buildings to the nearest road (unit: km)

zoning category	Training Data				Generated Data				T-test	
	size	mean	median	s.d.	size	mean	median	sd	t-statistics	p-value
apartment-res	5501	0.008772	0.008762	0.004382	257	0.009936	0.009279	0.005188	3.536819	0.000238
multifamily-res	4109	0.008571	0.007536	0.005625	251	0.009244	0.007030	0.006594	1.581692	0.057441
small-res	12595	0.014295	0.011380	0.008394	266	0.015916	0.012303	0.013813	1.907250	0.028777

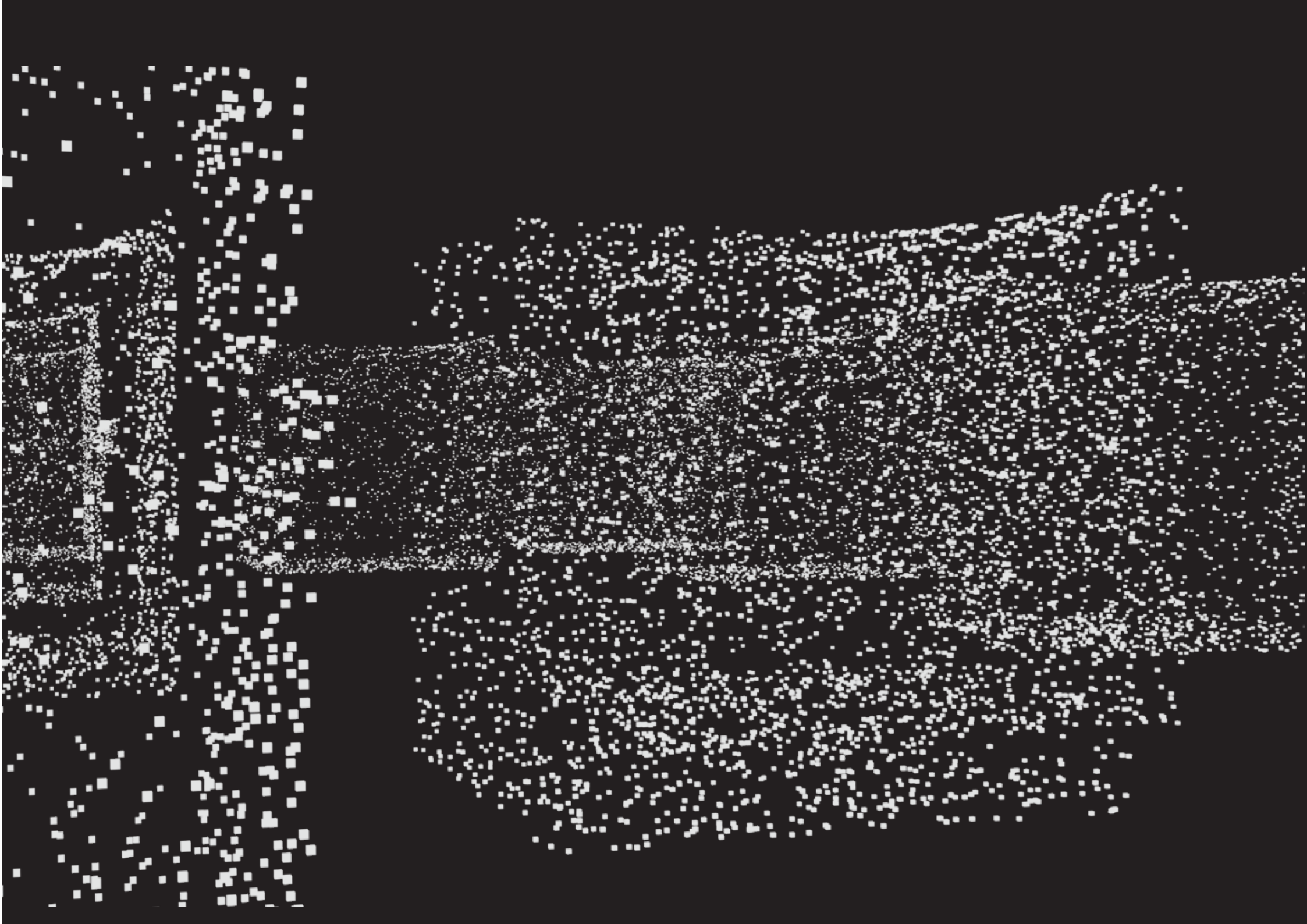
Table 6.2: Evaluation on the size of the buildings (unit: meter square)

zoning category	Training Data				Generated Data				T-test	
	size	mean	median	s.d.	size	mean	median	sd	t-statistics	p-value
apartment-res	5501	132.53739	95.55550	139.49698	257	274.90661	115.00000	409.14779	5.56322	0.00000
multifamily-resi	4109	117.06215	90.11780	156.99926	251	183.59212	133.41803	184.60762	5.58749	0.00000
small-res	12595	116.69473	109.88330	102.70262	266	104.61224	94.56700	58.71599	-3.25270	0.00064

Table 6.3: Evaluation on the height of the buildings (unit: floor)

zoning category	Training Data				Generated Data				T-test	
	size	mean	median	s.d.	size	mean	median	sd	t-statistics	p-value
apartment-res	5501	3.420651	3	1.095728	257	3.525292	3	1.992286	0.836122	0.201923
multifamily-resi	4109	2.199562	2	0.878388	251	2.450199	2	1.849009	2.132899	0.016940
small-res	12595	1.776340	2	0.687953	266	1.799228	2	0.656303	-0.271877	0.607041

Calculated using PostGIS

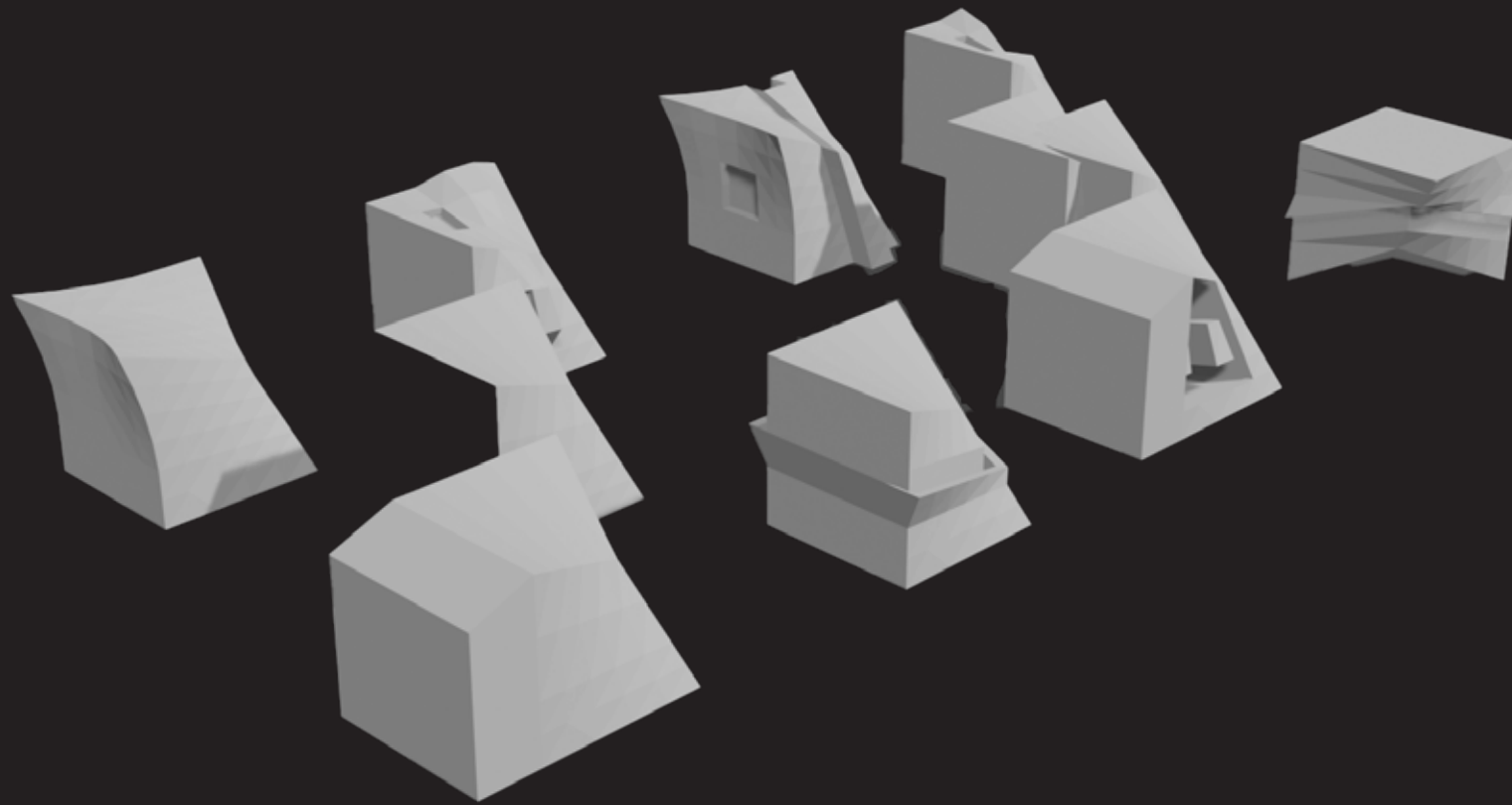


02 PointNet and Typology

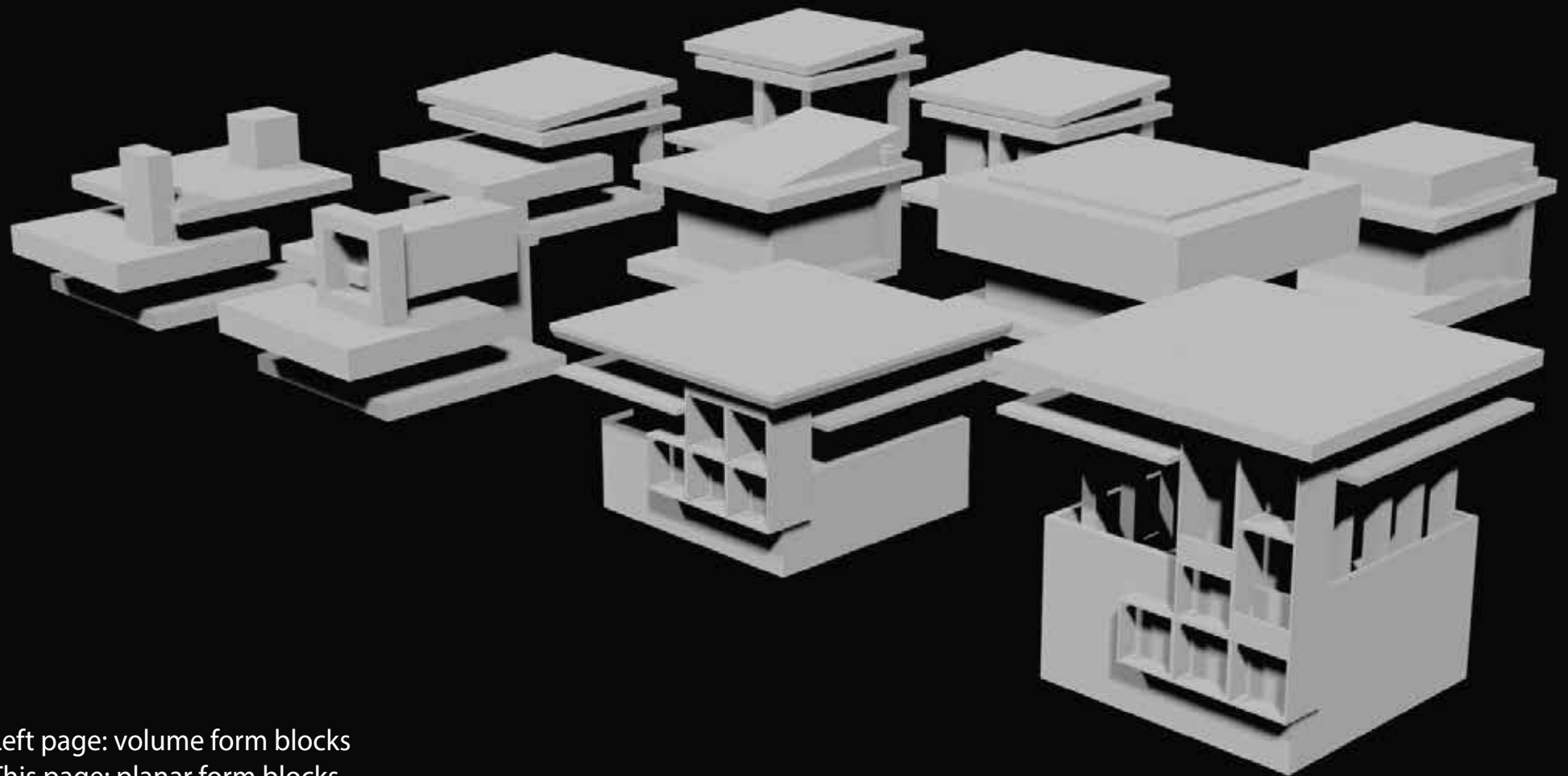
Individual work: Oct 2022- Nov 2022

Data modelled using blender; classification done using PointNet

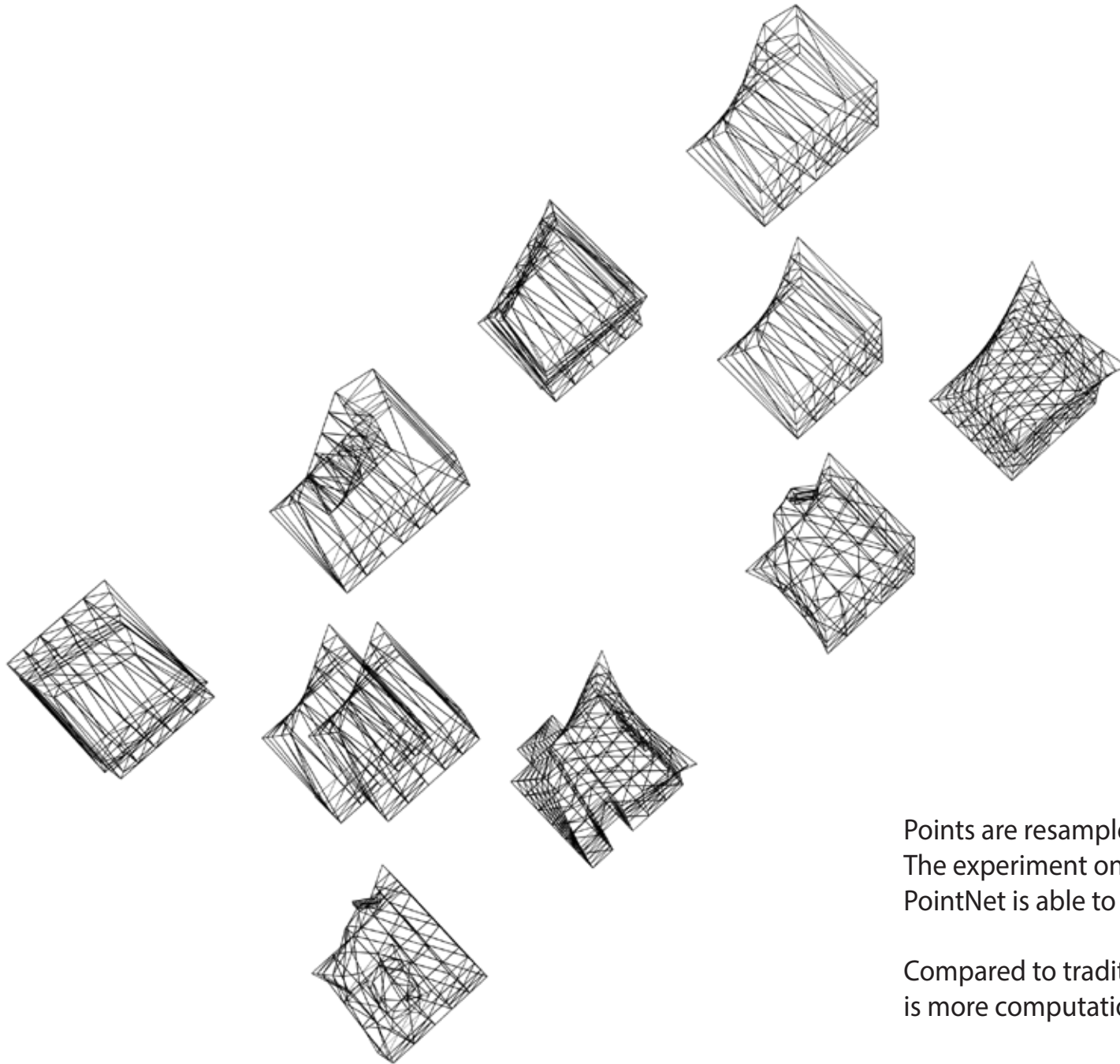
Typology is an crucial concept for both architecture and urban design. Therefore, in this research project, I propose an experiment to examine the potential of the point cloud deep learning model, PointNet in parsing buildings. Specifically, I propose an exploratory study: classify buildings based on their structural forms.



In this experienment, I generated two types of building blocks - volume-form buildings which morphed from a cube and planar-form buildings consisted of vertical and horizontal slabs. I use PointNet, an deep learning model that operates on point clouds to see if it is able to differenciate the two types.



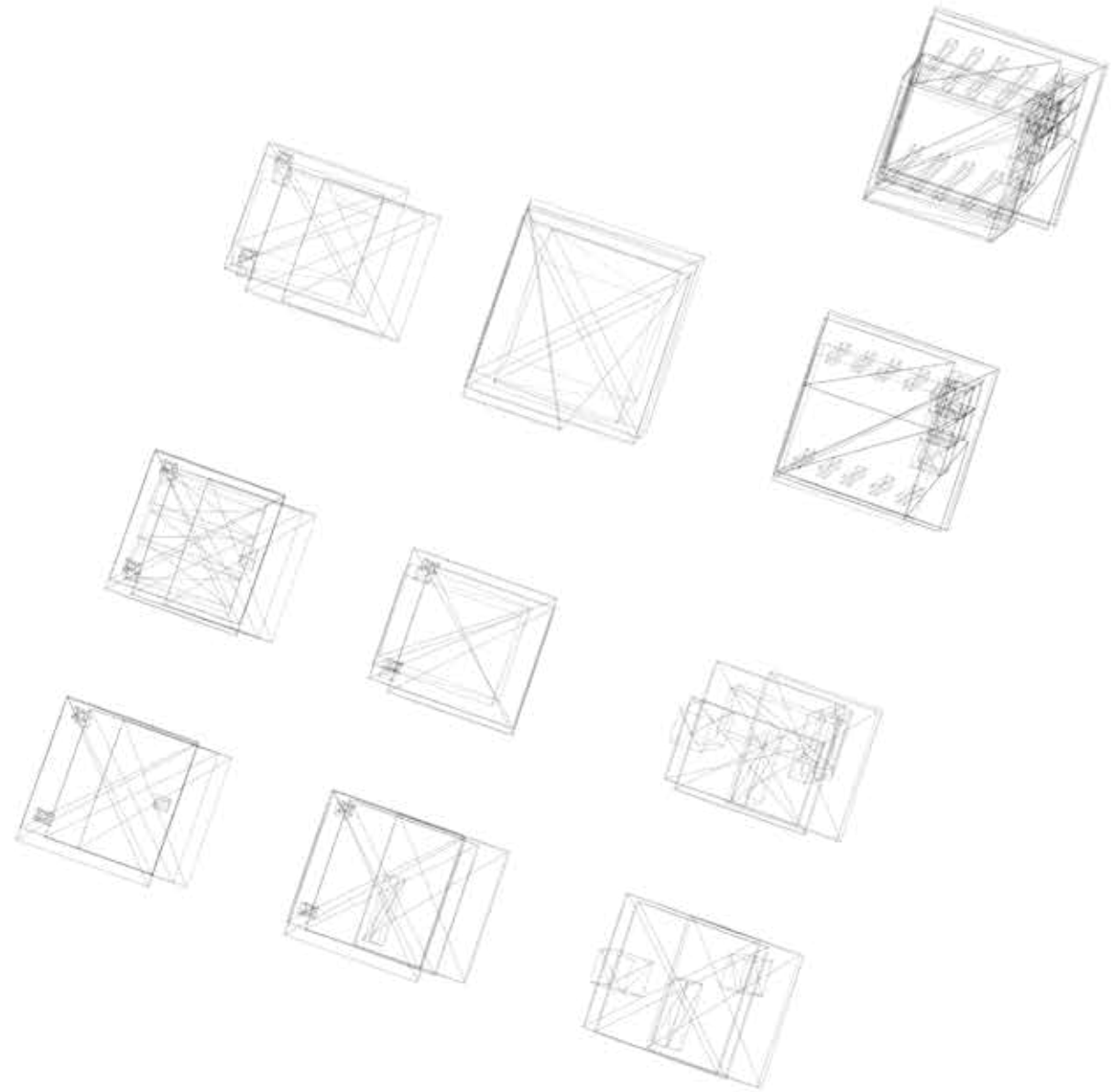
Left page: volume form blocks
This page: planar form blocks



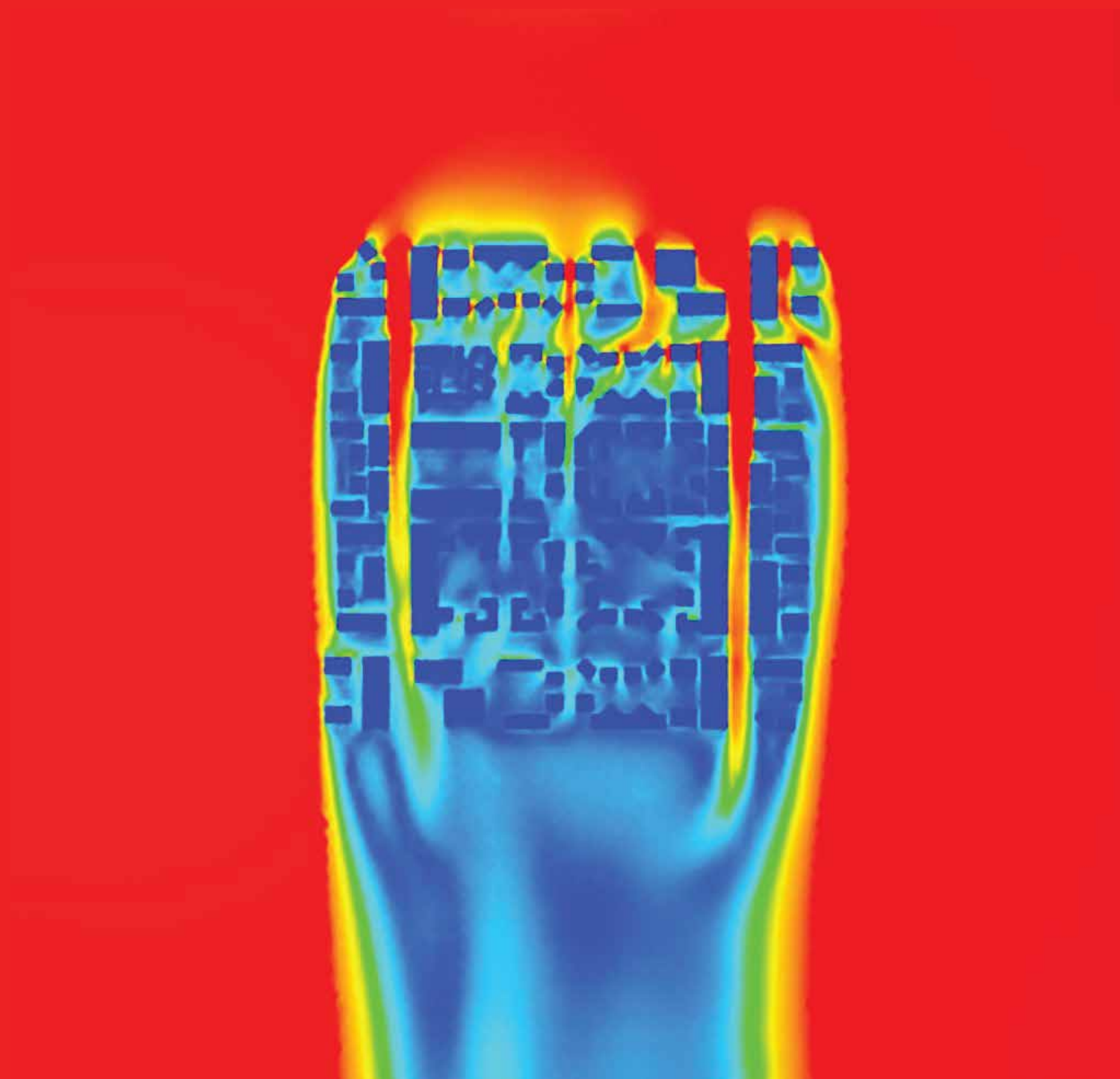
Points are resampled on block surfaces to produce the dataset. The experiment on a small dataset demonstrates that PointNet is able to learn the differences of the two groups.

Compared to traditional neural networks like CNN, PointNet is more computationally efficient.

Besides classification, PointNet can also be used for segmentation tasks too. It could potentially be used for purposes like the building restoration and the identification and analysis of historical architectures.



Wireframe visualization of the generated models.



03 Fast Wind Prediction with deep learning

Final Year Project

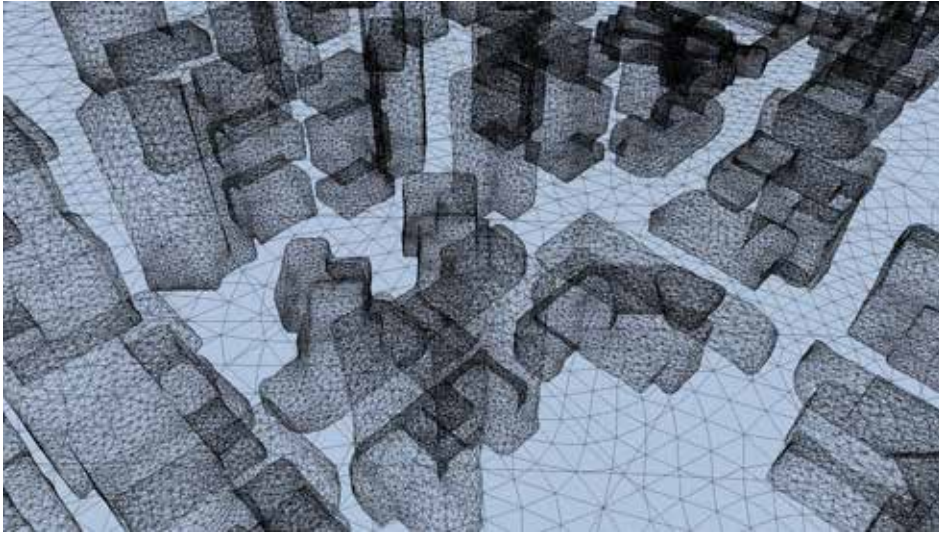
Sep 2022- Now (Ongoing)

Supervisors: Prof Tung Kum Hoe Anthony & Prof Filip Biljecki

All images credit to NUS Centre for Sustainable Asian Cities

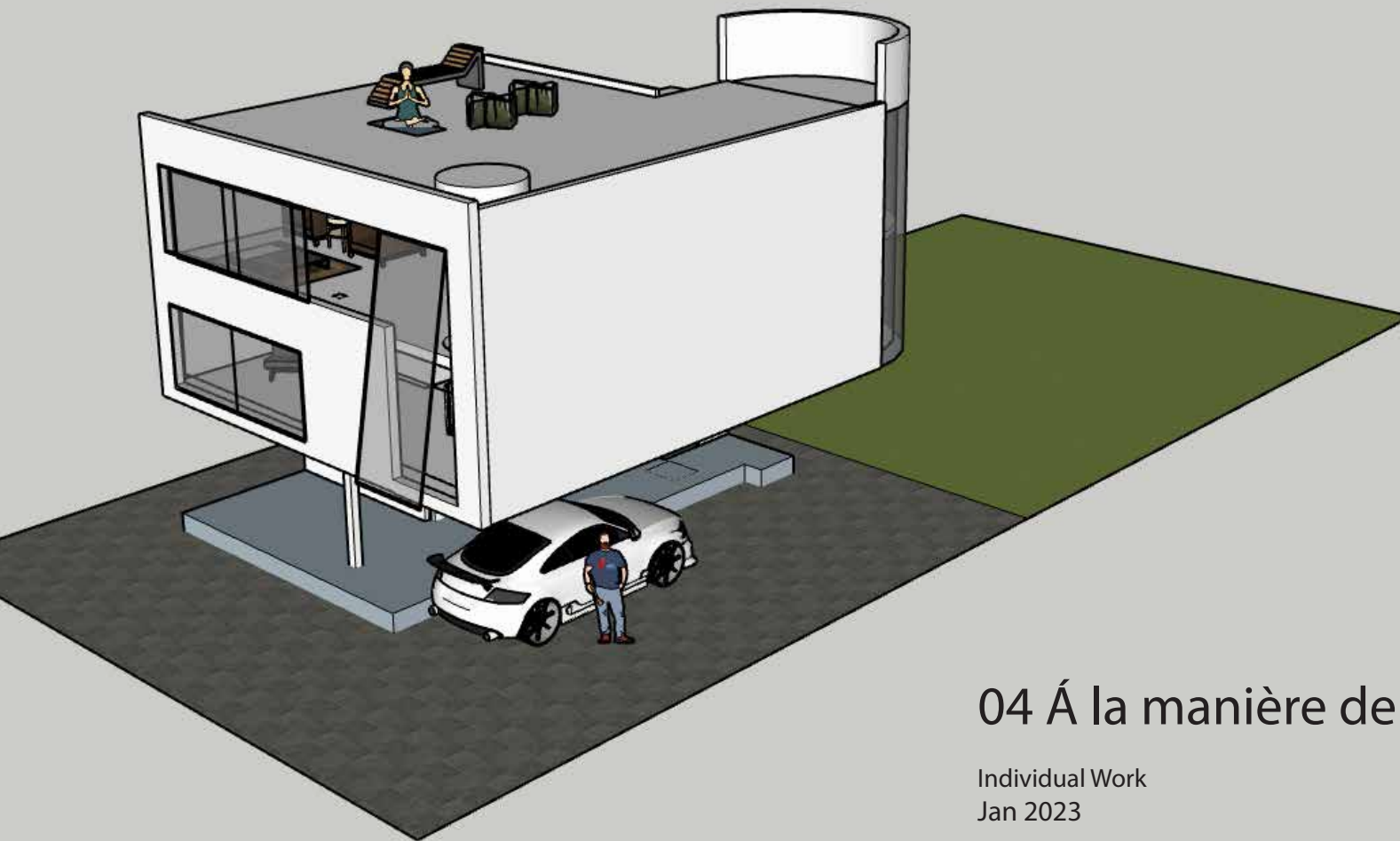
While wind speed performance is a crucial factor when design building morphology, traditional methods of wind speed estimation is frustratingly slow. In this project, I aim to improve existing machine learning methods on fast wind speed prediction.

My contribution: Literature Review, ML model building and training.



Situated in a hot and humid island, increasing natural ventilization using architectural design has been a popular research topic in NUS.

As part of the ongoing project, the aim is to use machine learning techniques such as UNet and GANs to speed up the wind speed prediction in order to aid the design process.



04 À la manière de Corbu

Individual Work
Jan 2023

When I was doing the Edx module, the Architectural Imagination, I was tasked to design a small row house resembling the style of Le Corbusier.

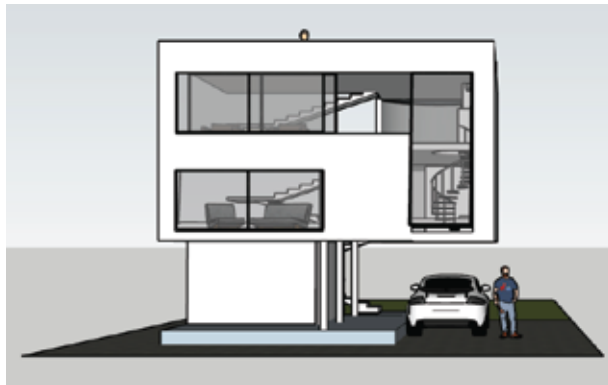
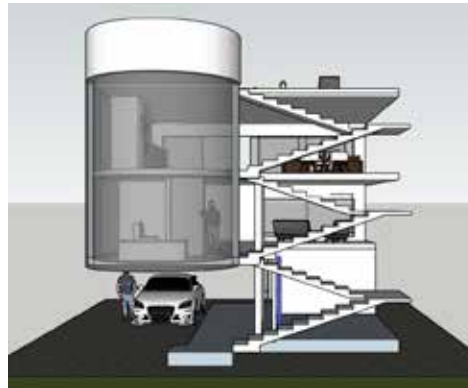
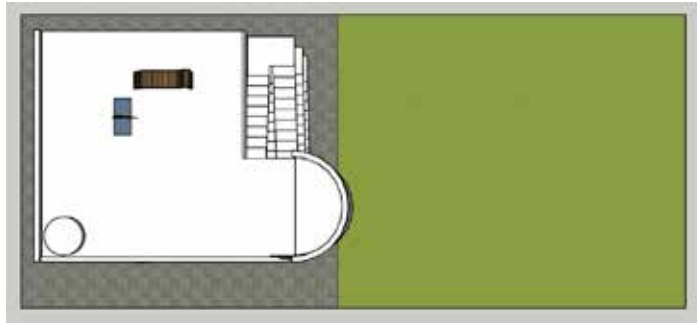


Do-mino, Le Corbusier

In his 1923 work *Toward an Architecture*, Le Corbusier proposed "the Five Points of the New Architecture" as a universally applicable language of the new architecture.

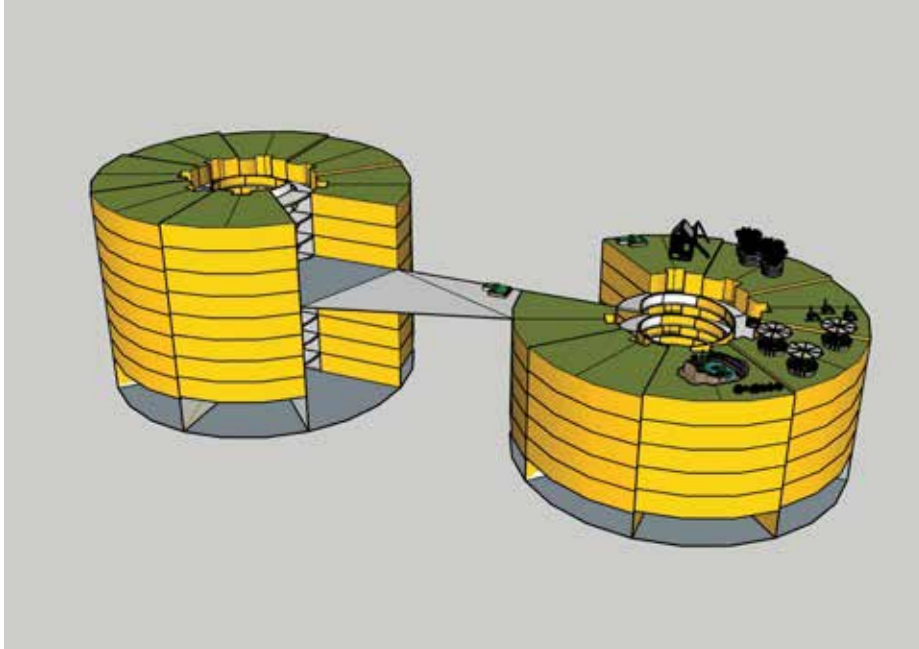
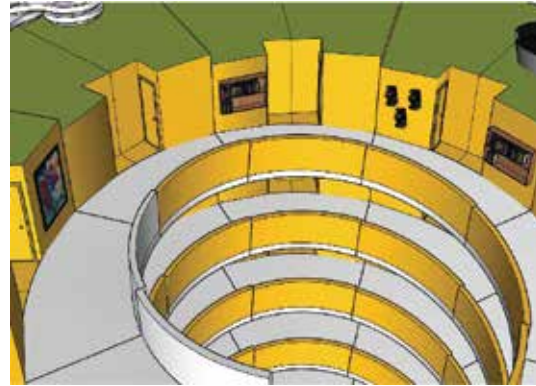
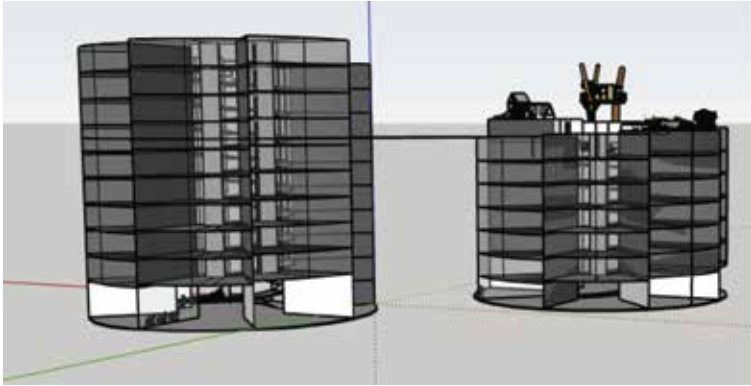
They are:

- the pilotis,
- the roof garden,
- the free plan,
- the ribbon window, and
- the free facade.



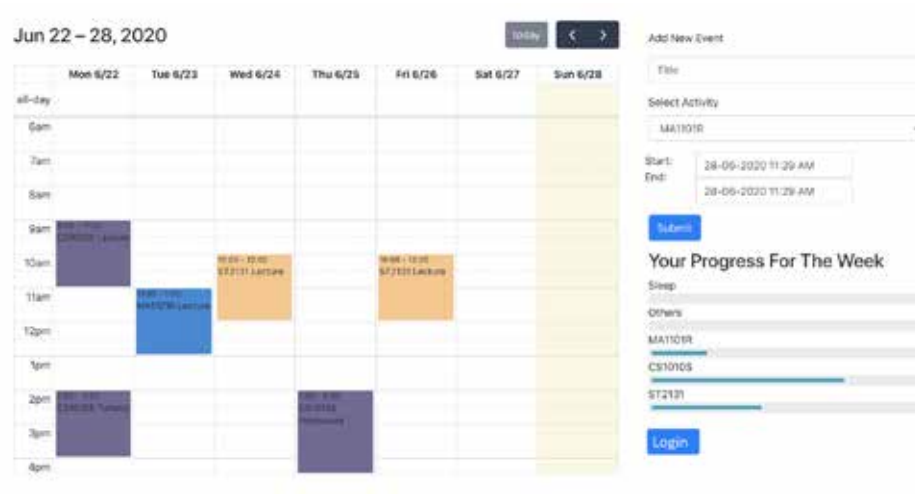
I designed this row house based on the do-mino model and the five points of new architecture. I was also trying to create “the architectural promenade” where the dwellers can experience different sceneries as they explore the space. I created two modes of the circulation. A stairway for more public usage and a spiral staircase which allows the resident to access the living room from the personal bedroom directly.

05 Other Works



Jan, 2018 China Think Big Competition
Team Project with Wu Weiying, Zhang Huiwen, Long Zeling

CTB is an competition what encourage students to explore social issues. During this competition, we focused one common sentiments shared by youths living in big cities - loneliness. I created this design of a communal living space for the youth in the hope that with more public spaces - roof garden, void deck and a long shared circular ramp - there would be more interaction among the residents.



Web App Design - A time management app for Uni students

May - Aug, 2020

Team Project with Zou Runzhong

I coded this web app using React JS and rails. It is an website that monitor and tracks the time the users spent on various activities and the assignment deadlines.

My contribution is the software engineering and the UI/UX design.

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Module Code

Upload your calendar (.ics) file from HuaMods.com

You can download .ics files from HuaMods.com (Lecture and Workshop)

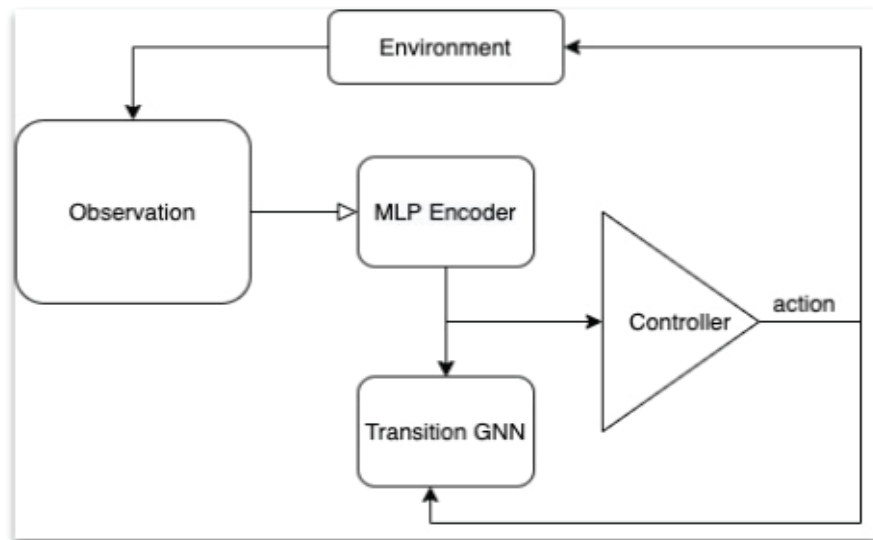
Manage your module commitments!

Do you feel your module actually require less or more than the recommended 10 hours?

Edit the expected module commitment hours to keep track of the time you realistically need for the module.

Delete the module you may have dropped. Delete the activity will delete all related events/deadlines.

ST2131 10



SMAWM - Structured Multi-Agent World Models

Aug- Dec, 2021

Team Project with Jet New

As part of the module AI Planning and Decision Making, my team created this project where we proposed SMAWM for reinforcement learning. By encompassing other agents in a compositional structure, SMAWM better models each individual agent's properties, interactions and relationships between agents, and Interactions between agents and the environment.

Contribution: Literature review and experiment environment set up