

DT-PosFormer : A Digital Twin Enabled Transformer Network for Precise Pose Estimation and Trajectory Prediction of MiC Modules.

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5/31 2024



The development of MiC and digital twin

Offsite Modular Prefabrication

Industry 4.0

Industrialized Construction

USA and Canada

Europe

Hong Kong MiC

Singapore ICPHs
bca.gov.sg hdb.gov.sg

Australia



The Chief Executive's 2022 Policy Address

To foster the development of MiC

The Chief Executive delivered the 2022 Policy Address on 19 October. To enhance Quantity, Speed, Efficiency and Quality, the Government will increase public housing and land supply by different measures, including the building of Light Public Housing by MiC to shorten the waiting time of public housing. The Housing Authority and Housing Society will adopt the MiC in more public housing projects. The DEVB will establish a cross-departmental steering committee, to streamline approval processes, and also making available land in the Northern Metropolis for manufacturing and storage of modules.



Uplift the Productivity of the Construction Industry



Public Housing Supply (I)

Every year in past
5 and 10 years



Construction Digitalisation Roadmap for Hong Kong

A Pathway Towards Smart Construction

SURVEY ON THE POTENTIAL UTILISATION OF PREFABRICATION YARDS IN HONG KONG - SURVEY REPORT

The special condition in Hong Kong

Modular Integrated Construction (MiC) is progressively replacing traditional construction.



Traditional construction site.

Complicated construction sites,
Labor-intensive industries.

**Goes against the Smart Site Safety System (4S)
Construction Innovation and Technology Fund (CITF)**



MiC construction site.

Improved construction quality,
simplified management,
reduced construction duration

Hong Kong

High density

High rise

High labor costs

High safety requirement

Limited site



Substantial Efforts: ITFs + RIF + SPPR + Consulting

Smart Manufacturing

Smart Logistics

ITP/045/13LPI
ITF 2014-16

R7027-18
RIF 2019-23

CIC MiC
CIC 2019

S2019.A8.013.19S
SPPR 2020

ITT/003/18LP
ITF 2018-19

Construction Blockchain Consulting
Industry 2019

PRP/068/20LI
ITF 2020-22

- Substantial Funding
- Close Industrial Collaboration
- Extensive Publications
- Academic insights

現場安裝服務 (當前版本 Beta/Build 150915)
On-site Assembly Service
RBIMP平台中的一個數字化管理工具包

首頁 | 項目進展 | 預製組件 | 定位及追蹤 | 事件提醒 | 設置及管理

現場安裝服務 > 項目進展 > 3D BIM 模型

3D BIM 模型：目標建築的物理和功能的現狀的立體展示。
第一步：瀏覽模型
第二步：點選施工中的預製構件查看詳情，或前往 [2D 平面圖](#) 瀏覽非當前樓層

Block: G/F 35/F Move Zoom Rotate Select Zoom All Orientation

Facade: Block5 9F; No15-TX8r

1. Produced
3. Arrived at site

RBIMP -- 現場安裝服務 OAS，由香港理工大學團隊研發，香港大學 Pi 實驗室提供部分技術支持。版權所有 © 2014-2016 香港理工大學

Prefab Production

Cross-border Logistics

Construction Site





Research Gaps and Further Requirements

“The faster the stacking process, the less the project duration and costs”

For the construction process

- Rely on the worker observation and the expertise of crane drivers
- Limited observation perspective and delayed stacking control

Safety accidents Installation errors Module collision



For the stakeholders

- Slow manual stacking process
- Limited management and information sharing

Module accumulation Delays in project delivery
Affecting customer satisfaction and the company's reputation



Further requirements

Comprehensive stacking perception
Dynamic and rapid response to emergent incidents
Predictive decision-making

Efficient information transmission and sharing among stakeholders

Predictive information and spatial and temporal traceability are the most important.



Analysis of feasible methods for realizing spatio-temporal tracking

Real-time spatio-temporal information collection, analysis, and visualization for module pose estimation and trajectory prediction may be the only viable and revolutionary solution

Mainstream method can be divided into visual-based and nonvisual based method.

Computer Vision

● Environment

Lighting conditions, cluttered background, and surface reflections of modules

Vibrations and shacking caused by construction activities

The camera perspectives are constrained by the deployment positions and angles

Image blurring and noise interference

● Cost

Full coverage camera on building sites is costly

Camera is easily vandalized and costly to maintain.

The cost of calculating full-scene data is huge.

Non-visual sensor-based method

● Environment

Ease of deployment

Long-term operational stability

Adaptability to variable environments

● Cost

Lower cost

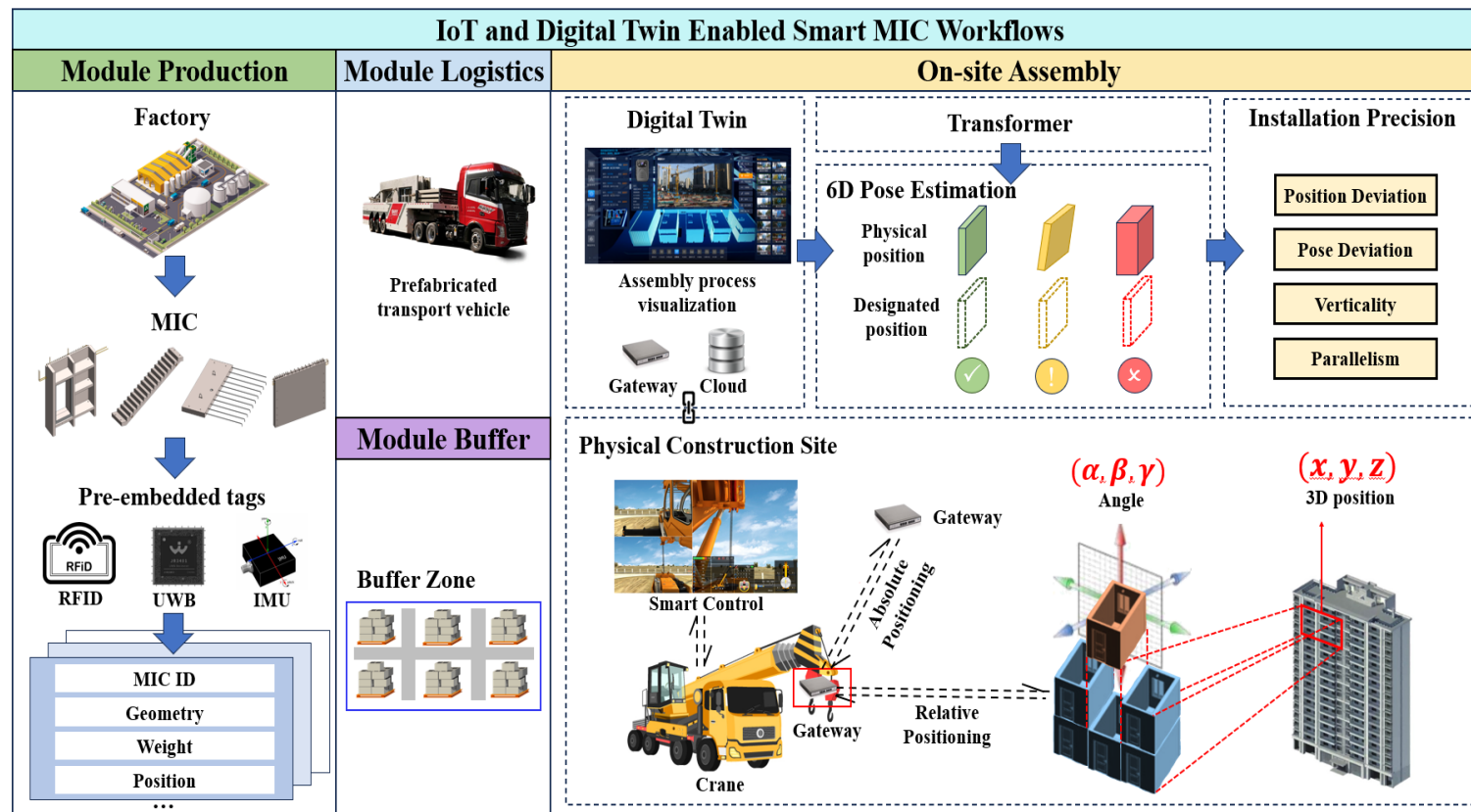
**Real-time 3D position information –UWB
(Ultra-Wide Band)**

**Real-time 3D Attitude information –IMU
Inertial Measurement Unit**

The workflows of smart MiC



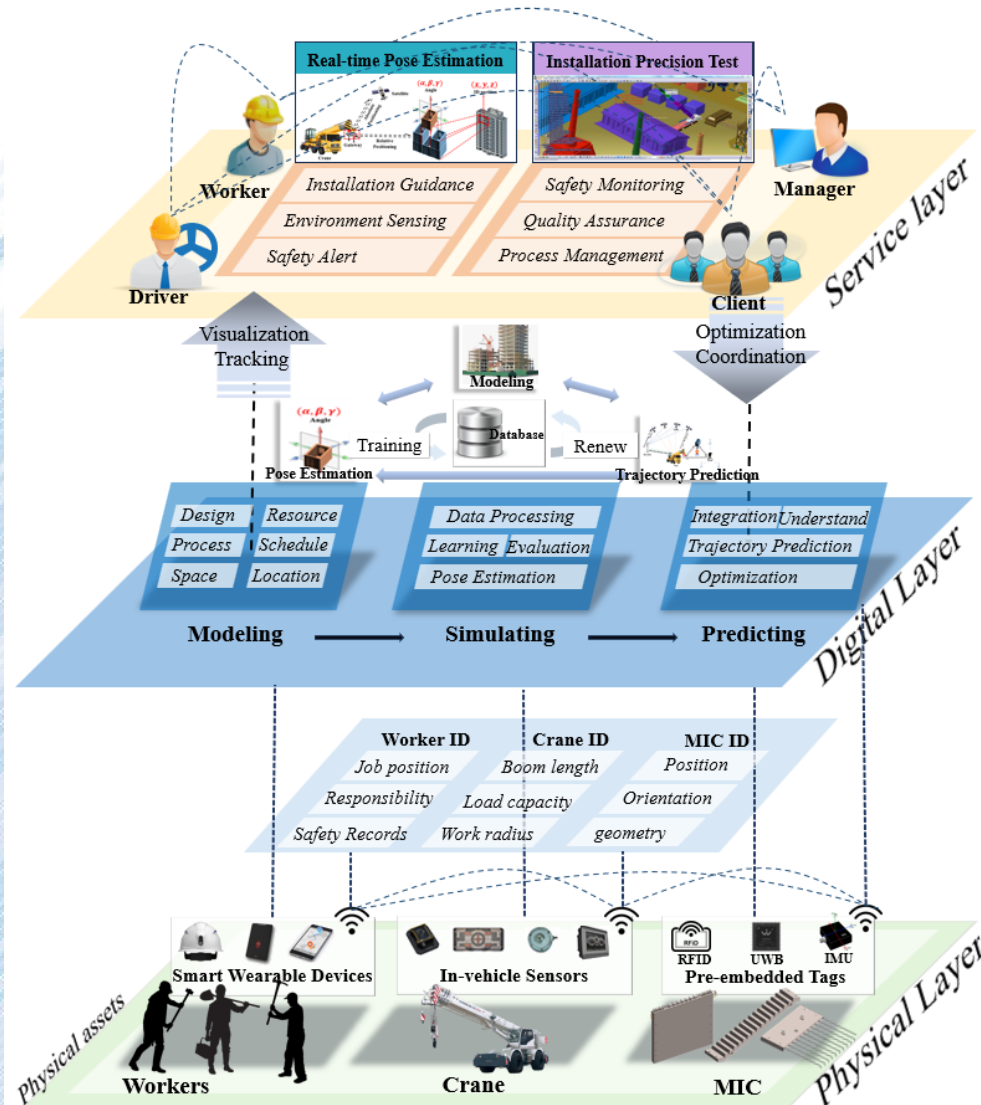
A DT-PosFormer-enabled pose estimation and trajectory prediction system (DTPTS) for MiC on-site assembly.



- A digital twin platform based on IoT is proposed for real-time perception and bidirectional interoperability
- A novel network based on Transformer is embedded into the digital twin for spatio-temporal module trajectory prediction
- Comparative studies with the current state-of-the-art DL models are conducted to validate the feasibility of the system and the effectiveness of the proposed method



The framework of DTPT system for MiC



Service layer

The service layer creates a real-time visualization and information sharing platform for all stakeholders together. It primarily encompasses two types of services: 1) Guidance service. 2) Management service.

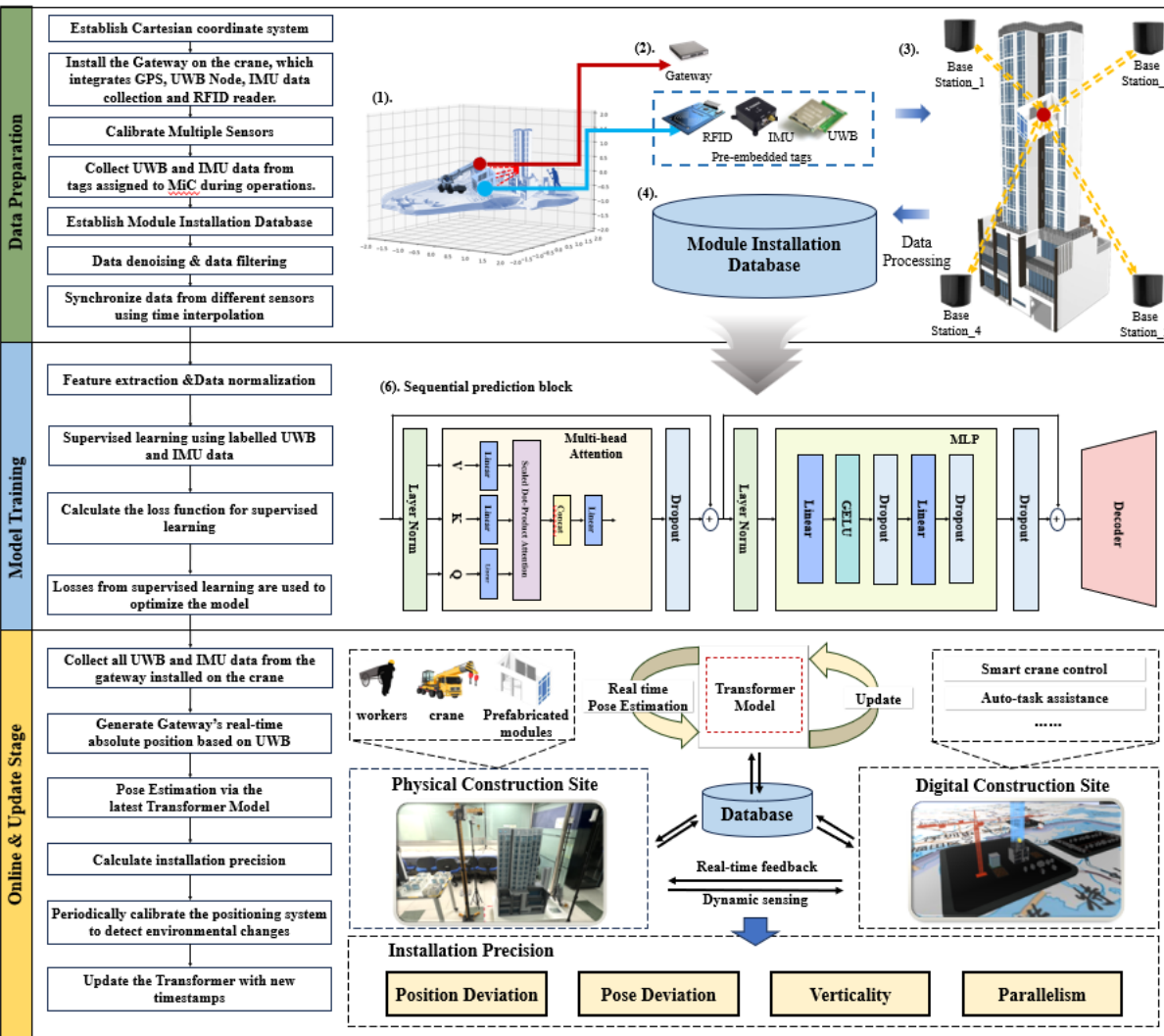
Virtual layer

The virtual layer mainly realizes the unified management of the data uploaded by the physical layer and establishes a high-fidelity digital twin of the stacking process.

Physical layer

The on-site assets are embedded in smart IoT devices for intelligent sensing. Multi-dimensional and multi-scale data of the internal characteristics and dynamic activities are captured by the IoT.

Pose estimation and trajectory prediction



The past trajectory of module i:

$$X_i = (x_i^t, y_i^t, z_i^t, \varphi_i^t, \theta_i^t, \psi_i^t)$$

The true trajectory of module i:

$$Y_i = (x_i^{t+t_{\text{pred}}}, y_i^{t+t_{\text{pred}}}, z_i^{t+t_{\text{pred}}}, \varphi_i^{t+t_{\text{pred}}}, \theta_i^{t+t_{\text{pred}}}, \psi_i^{t+t_{\text{pred}}})$$

The predicted trajectory of module i:

$$\hat{Y}_i = (\hat{x}_i^{t+t_{\text{pred}}}, \hat{y}_i^{t+t_{\text{pred}}}, \hat{z}_i^{t+t_{\text{pred}}}, \hat{\varphi}_i^{t+t_{\text{pred}}}, \hat{\theta}_i^{t+t_{\text{pred}}}, \hat{\psi}_i^{t+t_{\text{pred}}})$$

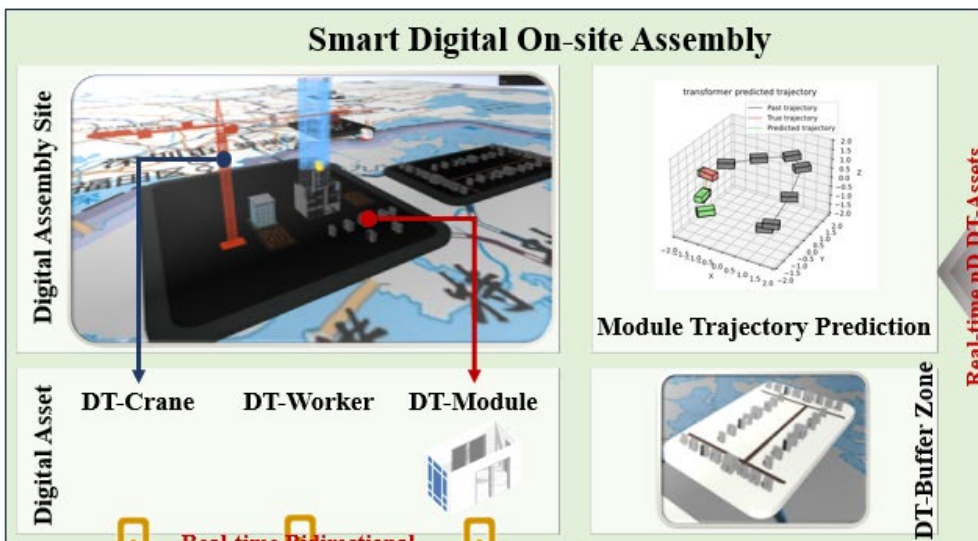
The loss function:

$$\text{SmoothL1} = \frac{1}{n} \sum_{i=1}^n \begin{cases} 0.5 * (p_{i_p}^t - p_{i_{GT}}^t)^2, & \text{if } |p_{i_p}^t - p_{i_{GT}}^t| < 1 \\ |p_{i_p}^t - p_{i_{GT}}^t| - 0.5, & \text{otherwise} \end{cases}$$

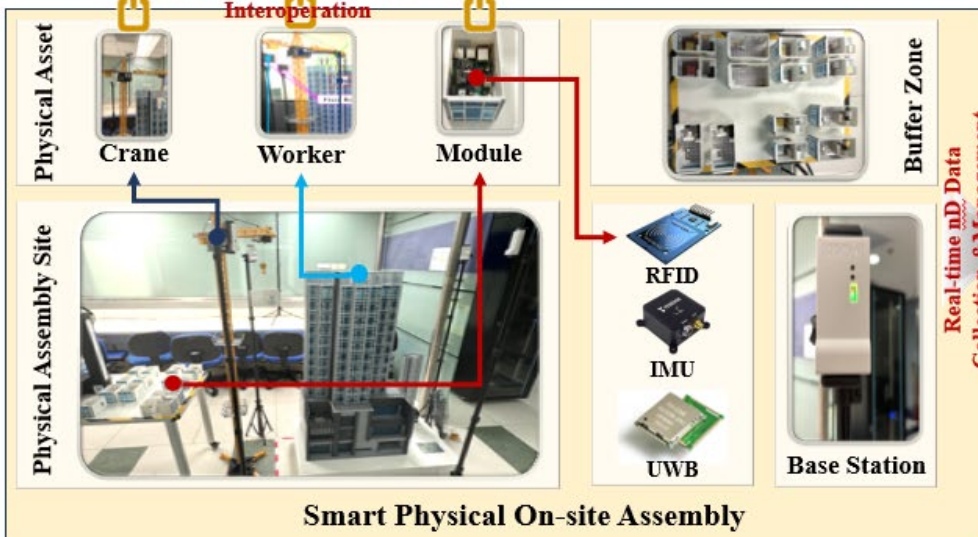
Experiments



Smart Digital On-site Assembly



Real-time nD DT-Assets
Visualization & Tracking



Real-time nD Data
Collection & Management

Smart Data Management



Evaluation metrics

1) Final Displacement Error (FDE) :

This is the distance between the last point of the predicted trajectory and the last point of the real trajectory.

$$FDE = \frac{1}{n} \sum_{i=1}^n \left| p_{i_p}^{tf} - p_{i_{GT}}^{tf} \right|$$

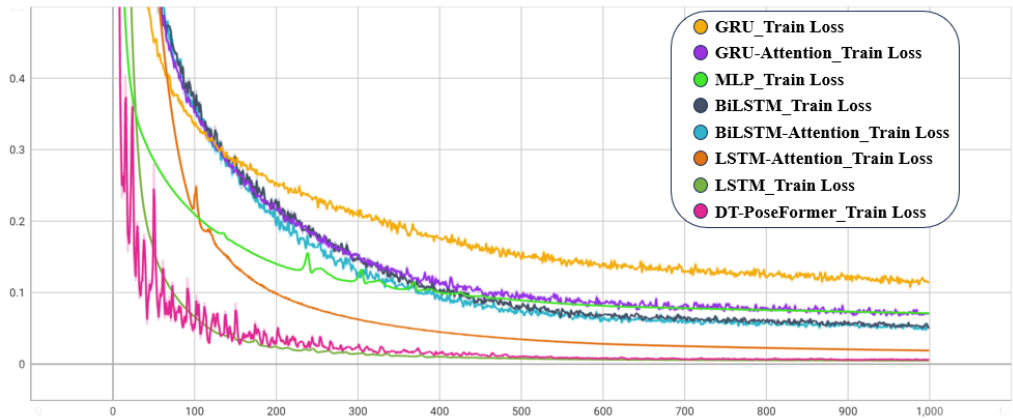
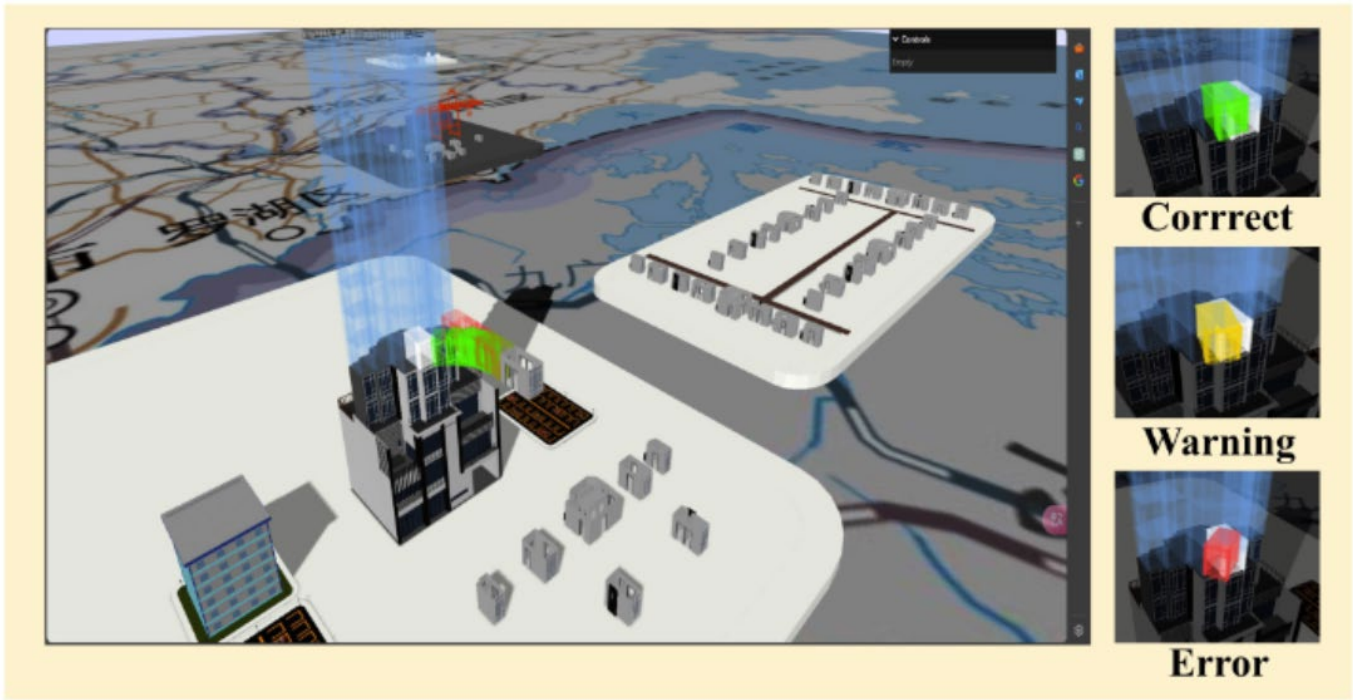
2) Average Displacement Error (ADE) :

This is the average distance between all points of the predicted trajectory and the corresponding points of the real trajectory.

$$ADE = \frac{1}{n} \sum_{i=1}^n \frac{1}{t_f} \sum_{t=1}^{t_f} \left| p_{i_p}^t - p_{i_{GT}}^t \right|$$



Results



The loss of different methods.

TABLE I. COMPARISONS ON EVALUATION METRICS AND LOSS

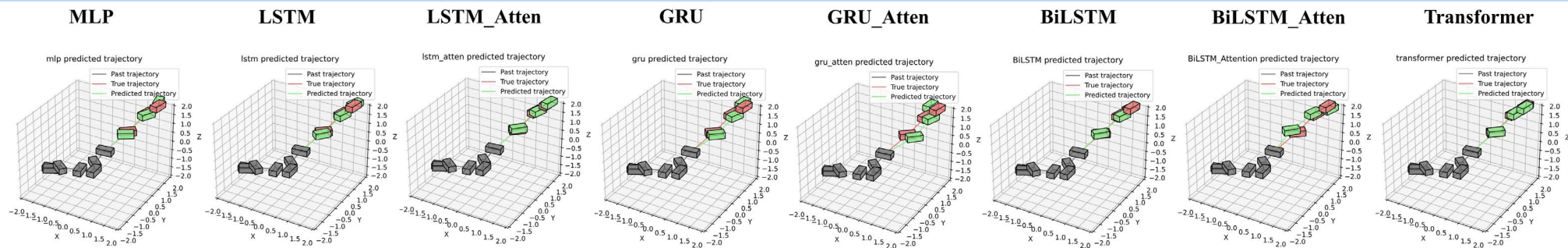
Model	Smooth L1 loss	ADE	FDE
MLP	0.07129	57.32236	316.55729
GRU	0.1141	26.19518	137.12731
GRU-Attention	0.07162	26.17367	138.24061
LSTM	0.004935	26.30367	139.94350
BiLSTM	0.052	26.01963	139.14750
BiLSTM-Attention	0.05083	26.02726	139.12281
LSTM-Attention	0.01908	26.01323	138.46538
Ours(PoseFormer)	0.00661	23.58859	136.27763

The evaluation matrices of different methods.

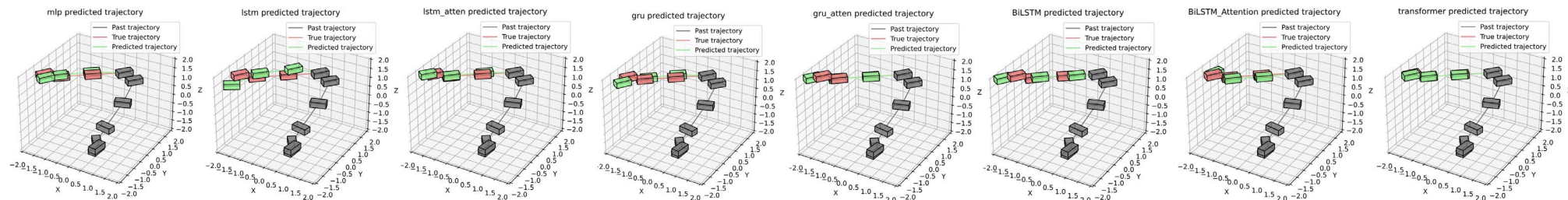


Results

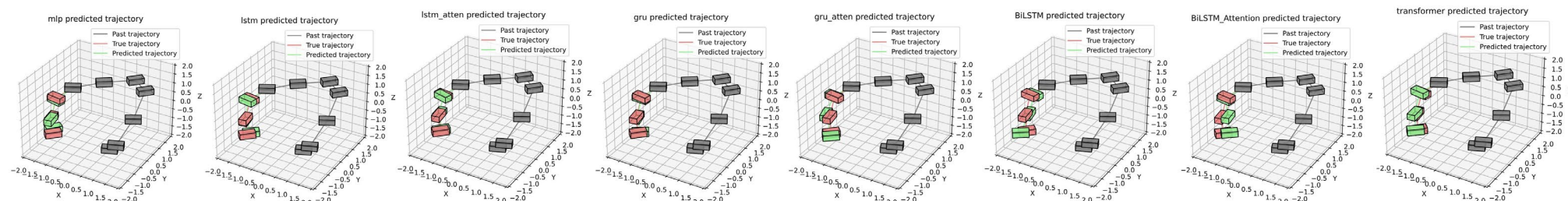
Scenario 1



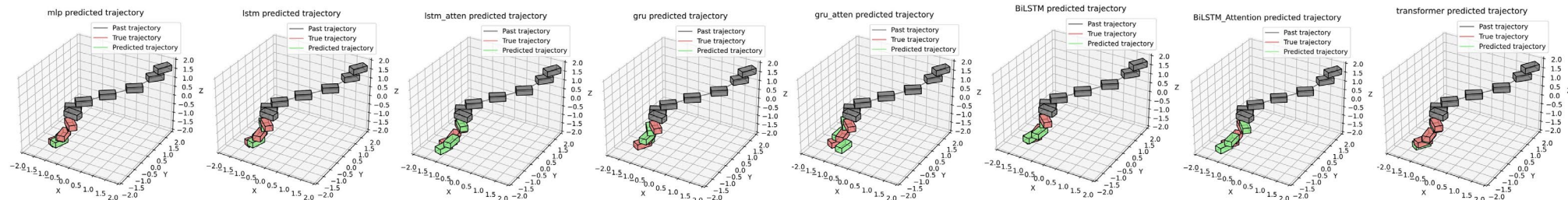
Scenario 2



Scenario 3



Scenario 4



The visualization of pose estimation and trajectory prediction in different scenarios.



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