



CS566
Fall 2025

Vision-Based Dynamic Objects Path Predictions for Safer Robot Navigation on Construction Sites.

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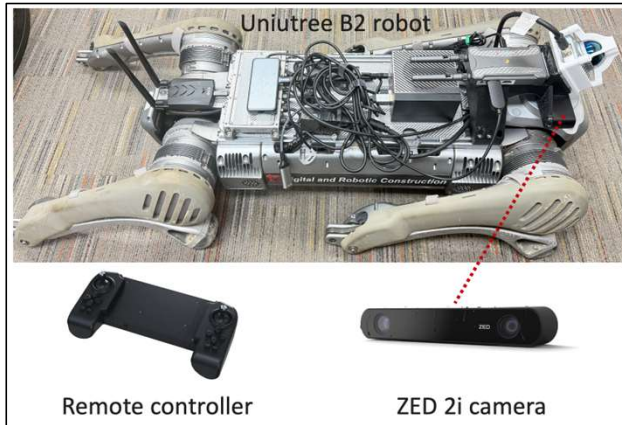
Project inspiration

UW-Madison Kellner Family Athletic Center.

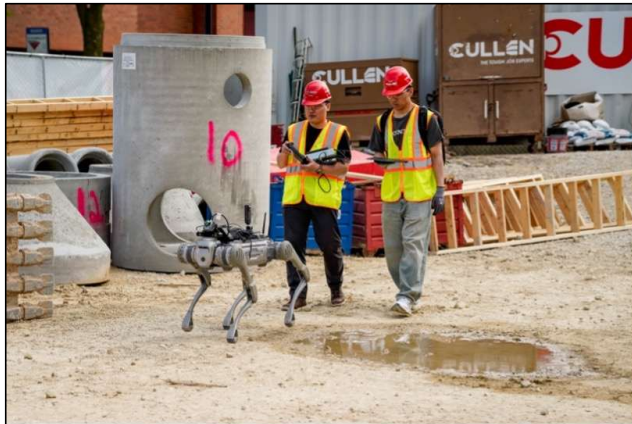


Layout of the construction site, photographed on May 21, 2025

Pipeline – Data Collection



Unitree B2 used for field data collection

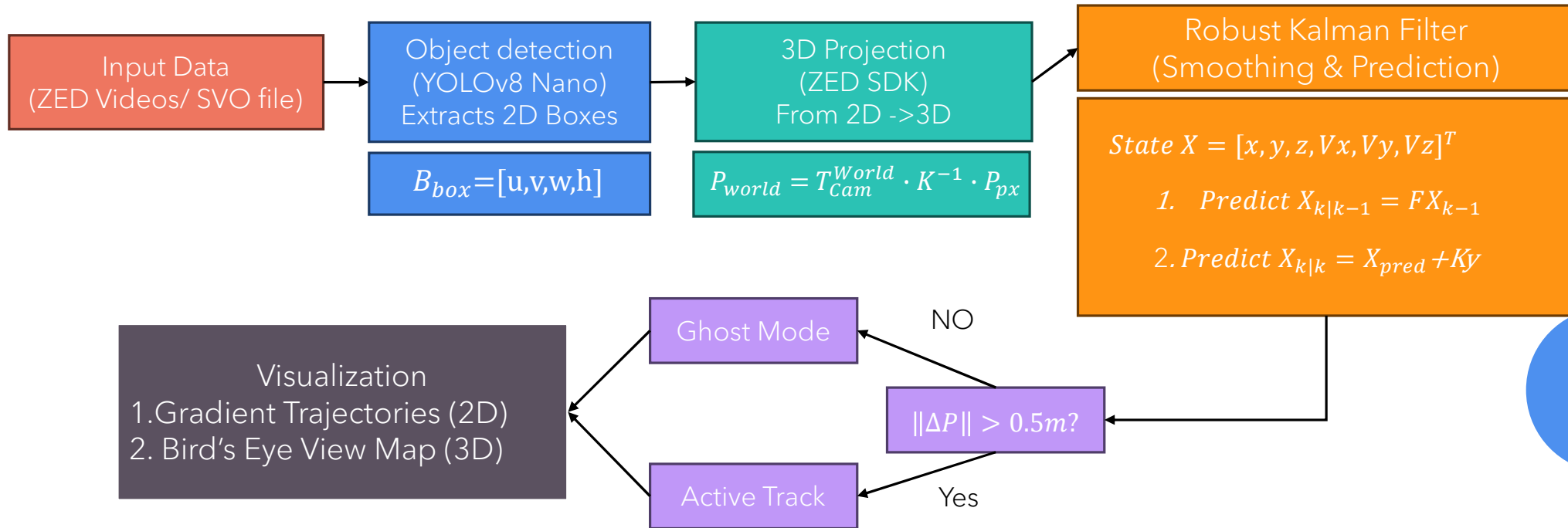


Field Data Collection



Example of the recorded Video

Pipeline – Our Approach



Demo-1 Worker



Robot View

Bird's Eye View Map

Demo-2 Worker Standing



Robot View

Demo-3 Worker Stop & Go



Bird's Eye View Map

Robot View

Demo-4 Load and Truck



Bird's Eye View Map

Robot View

Discussion and Future Work

1. Comparison: Physics-Based vs. Data-Driven

Our Approach (Robust Kalman Filter):

- **Pros:** Real-time performance (low latency), highly interpretable ("Ghost Mode"), and works immediately without massive training datasets.
- **Cons:** Assumes linear velocity; struggles with sudden sharp turns.

Deep Learning Pipelines (e.g., LSTM / Transformers):

- **Pros:** Can model complex, non-linear paths (e.g., walking in a curve).
- **Cons:** "Black box" unpredictability, higher computational cost, and requires huge annotated datasets to avoid overfitting.

2. Future Improvements

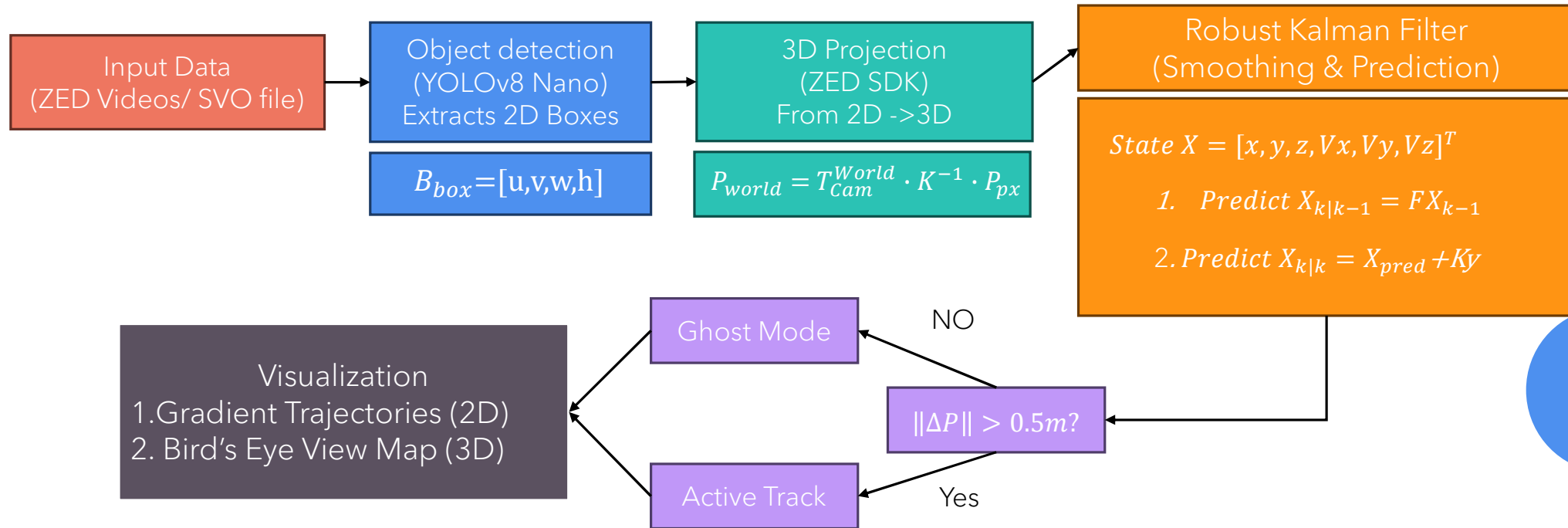
Interaction-Aware Modeling: Implement **Social-GAN** to predict how workers avoid each other (currently, we track objects independently).

Sensor Fusion: Integrate **LiDAR data** with the ZED camera to solve the low-light/occlusion failures we observed in the field.

Long-Term Forecasting: Train an **LSTM** specifically for >5-second predictions where the linear Kalman assumption fails.



Pipeline – Our Approach





Thank you!

Q&A!