

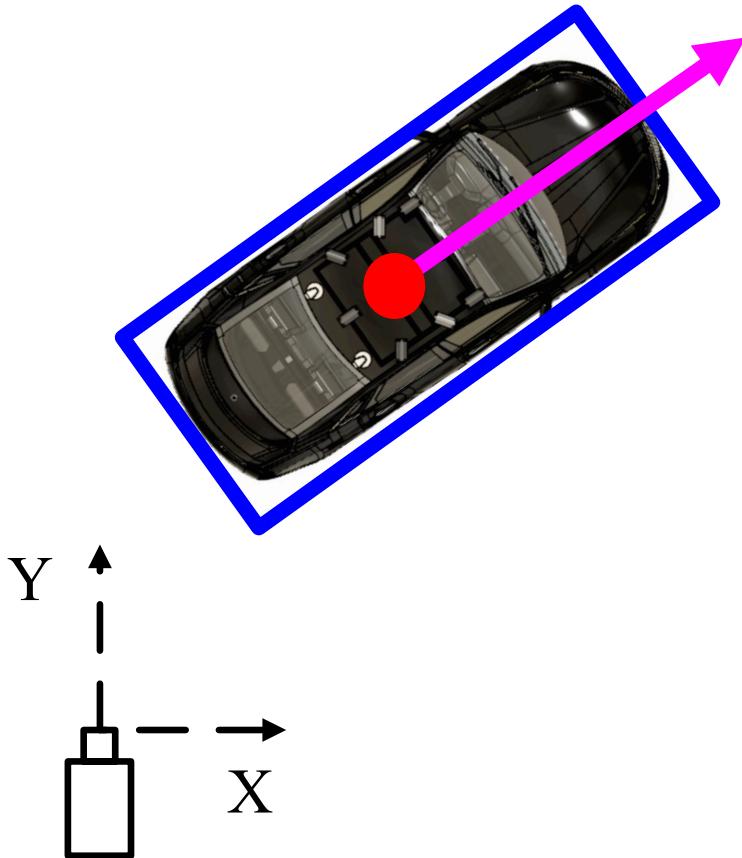
Using 2D Object Detectors for Self-Driving Perception

Course 3, Module 4, Lesson 4



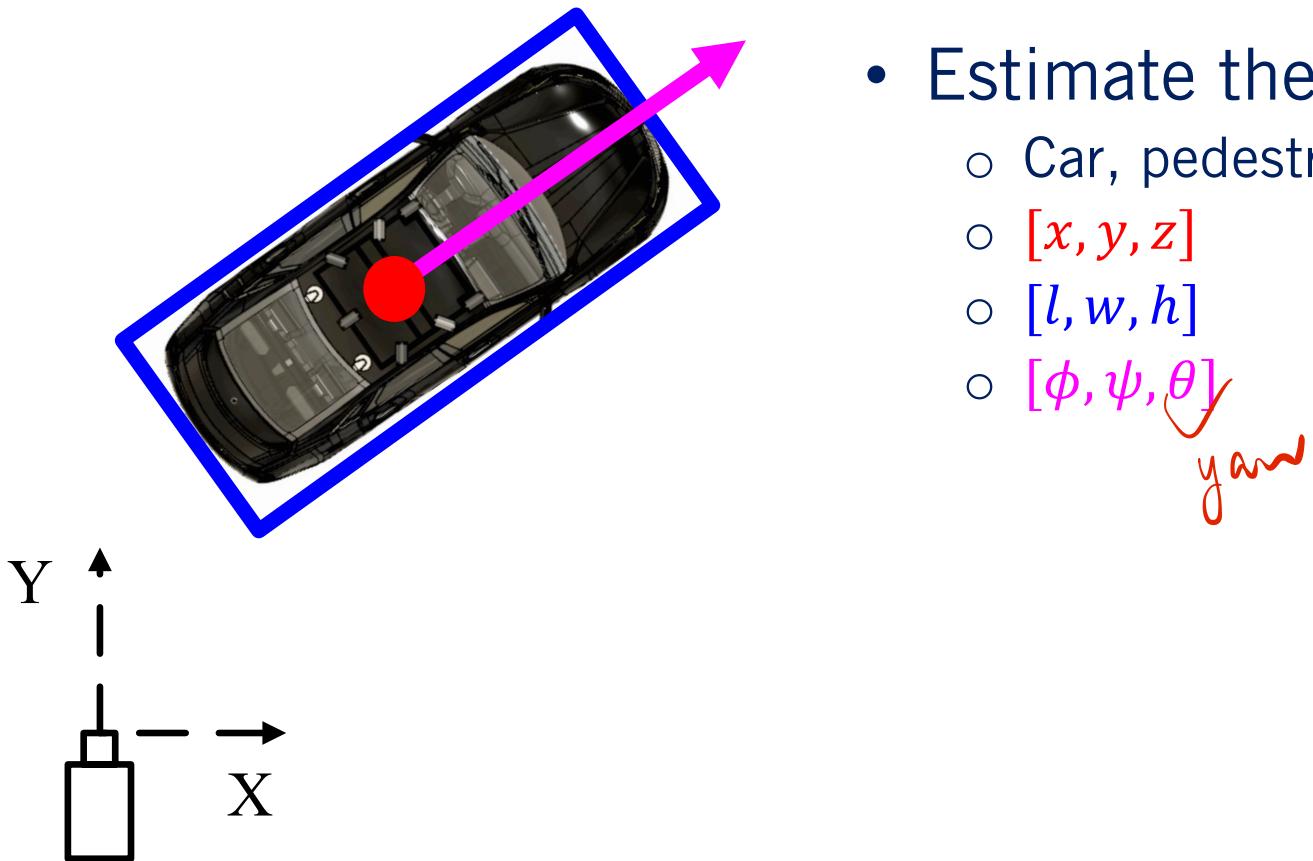
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3D Object Detection



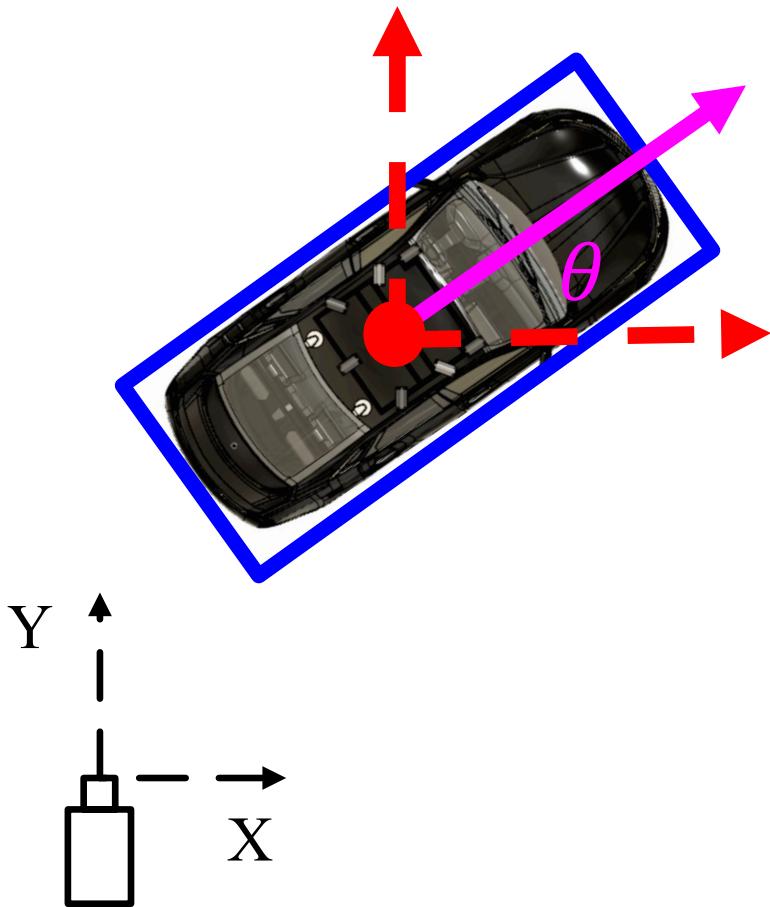
- Estimating the:
 - Category Classification
 - Position of the centroid in 3D
 - Extent in 3D
 - Orientation in 3D

3D Object Detection



- Estimate the:
 - Car, pedestrian, cyclist
 - $[x, y, z]$
 - $[l, w, h]$
 - $[\phi, \psi, \theta]$
yawn

3D Object Detection



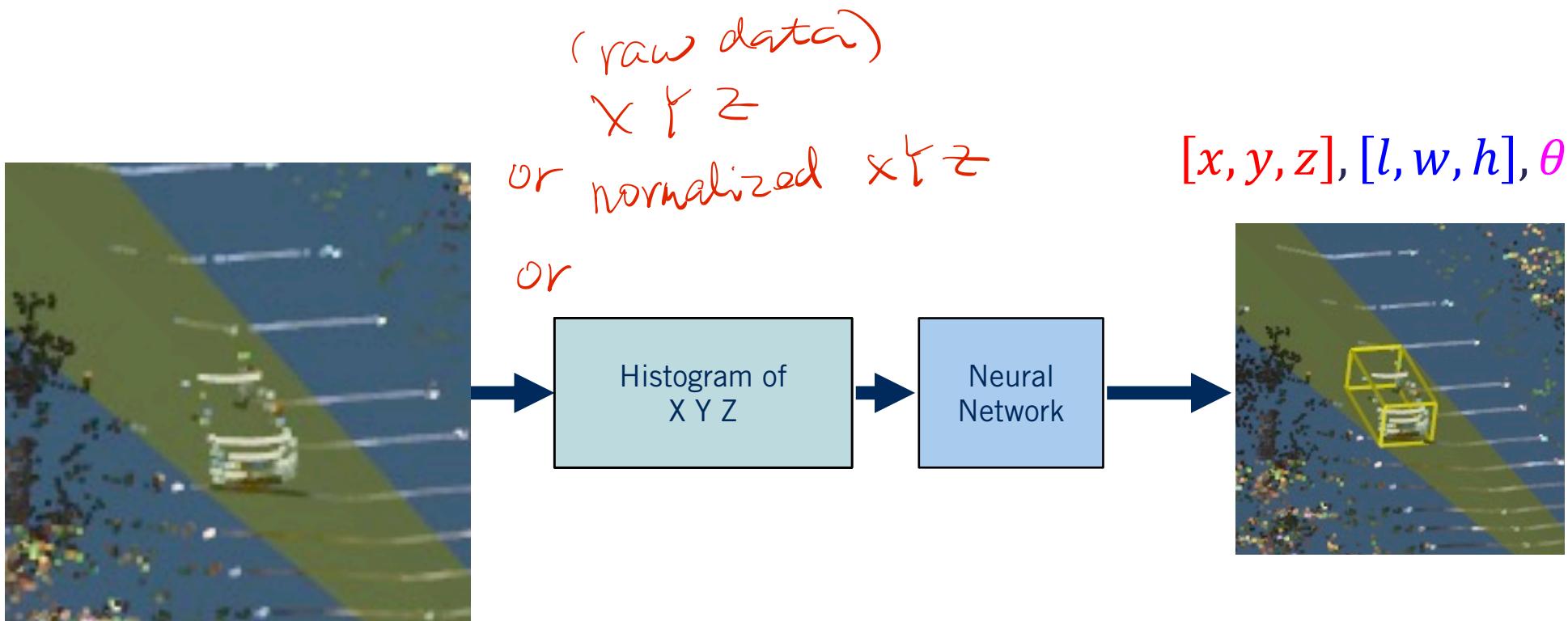
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From 2D → 3D Object Detection

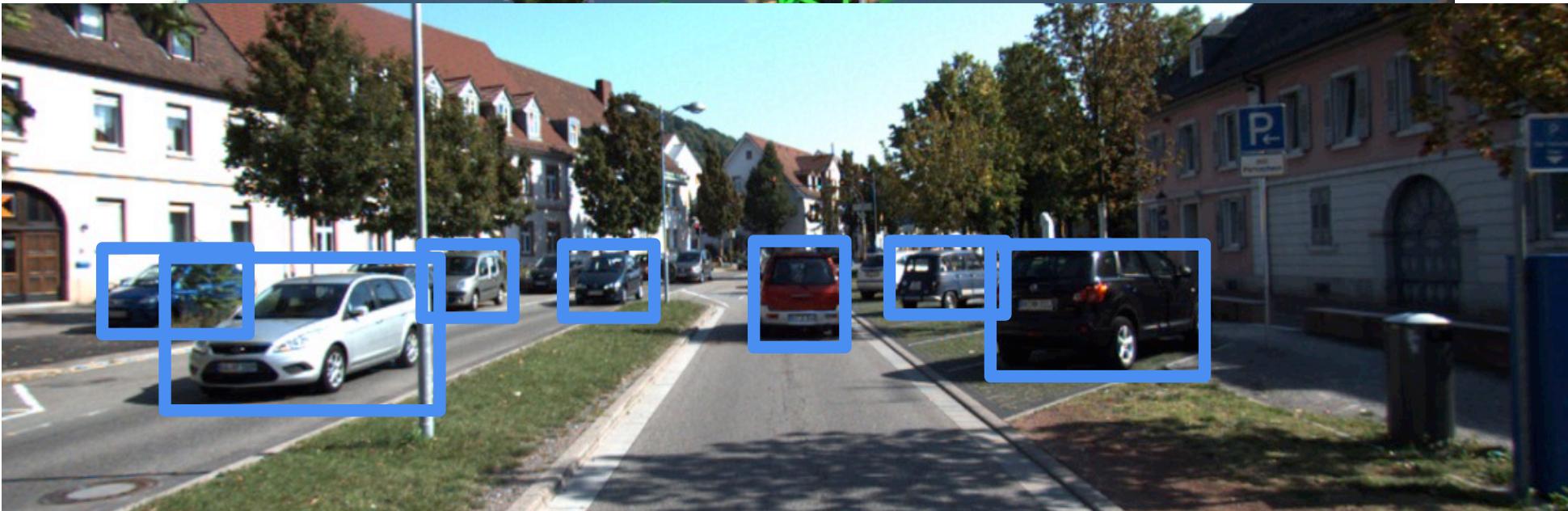
Data From: Geiger, Andreas, Philip Lenz, and Raquel Urtasun. "Are we ready for autonomous driving? the KITTI vision benchmark suite." *Computer Vision and Pattern Recognition (CVPR), 2012 IEEE Conference on*. IEEE, 2012.



From 2D → 3D Object Detection



Typical 3D object detection results



Code at: <https://github.com/kujason/avod>

From 2D → 3D Object Detection

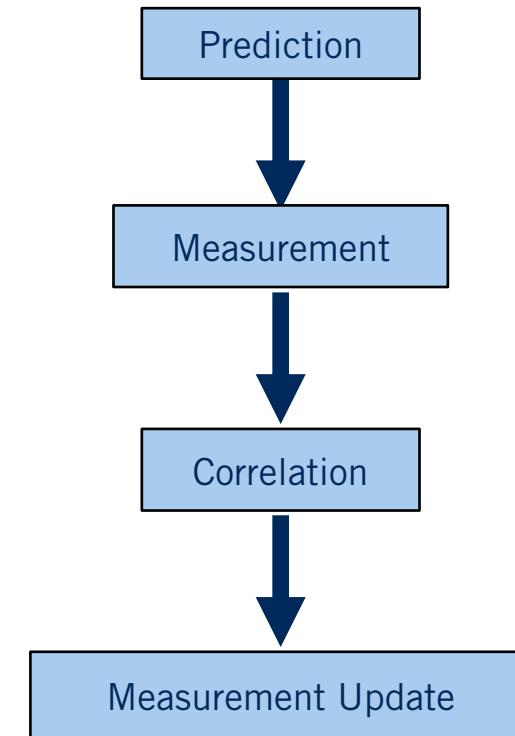
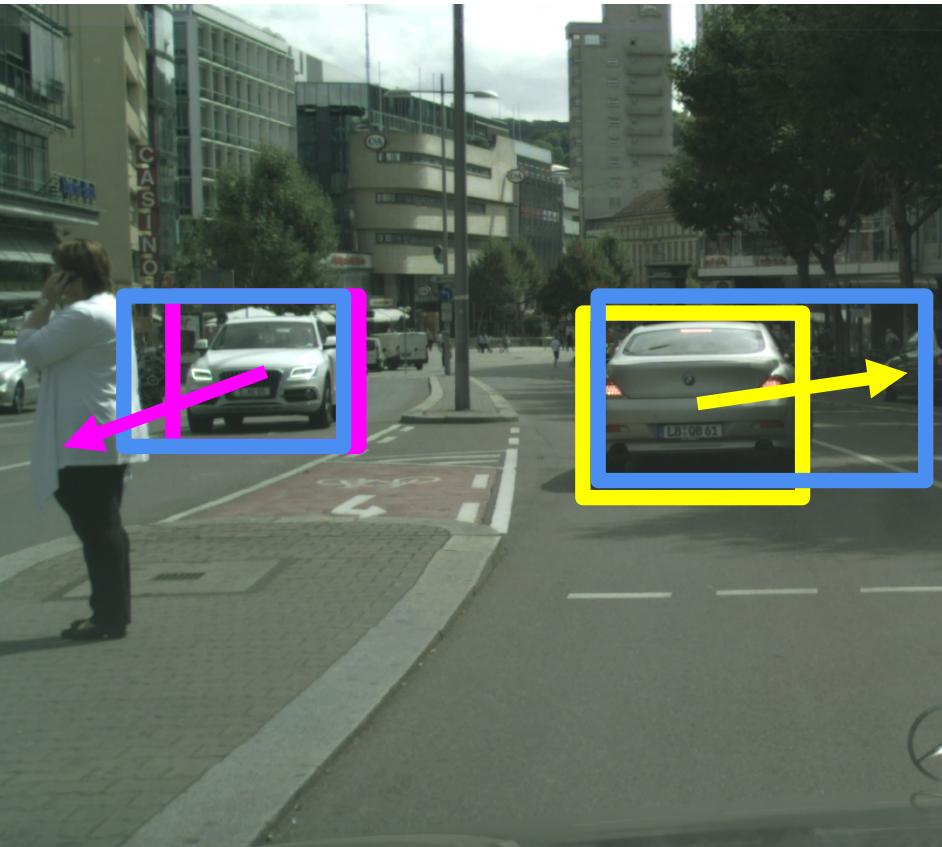
- Advantages:
 - Allows exploitation of mature 2D object detectors, with high precision and recall
 - Class already determined from 2D detection
 - Does not require prior scene knowledge, such as ground plane location
- Disadvantages:
 - The performance of the 3D estimator is bounded by the performance of the 2D detector
 - Occlusion and truncation are hard to handle from 2D only
 - 3D estimator needs to wait for 2D detector, inducing latency in our system

delayed perception

2D Object Tracking

- **Detection:** We detect the object independently in each frame and can record its position over time
- **Tracking:** We use image measurements to estimate position of object, but also incorporate position predicted by dynamics, i.e., our expectation of object's motion pattern
- **Tracking Assumptions:**
 - Camera is not moving instantly to new viewpoint
 - Objects do not disappear and reappear in different places in the scene
 - If the camera is moving, there is a gradual change in pose between camera and scene

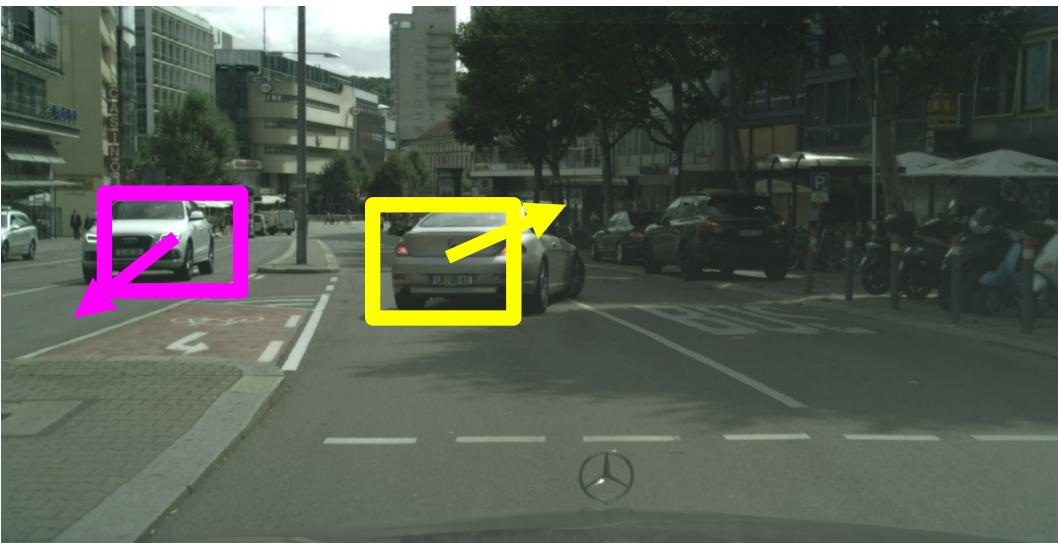
2D Object Tracking



Object Tracking: Prediction

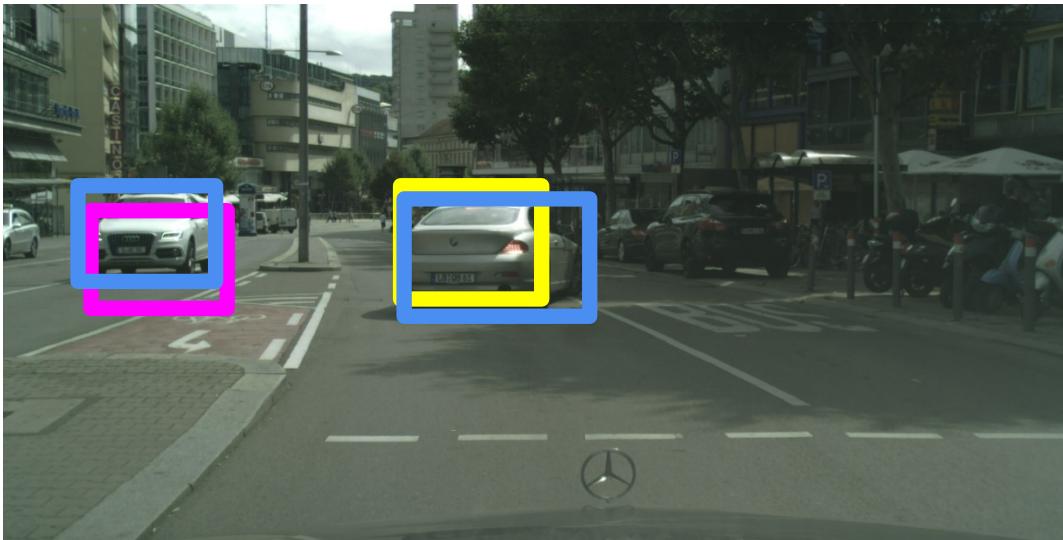
- Each object will have a predefined **motion model** in image space
- **Example:** $p_k = p_{k-1} + v_k \Delta t + \mathcal{N}(0, \Sigma)$

$\mathcal{N}(0, \Sigma)$
Noise



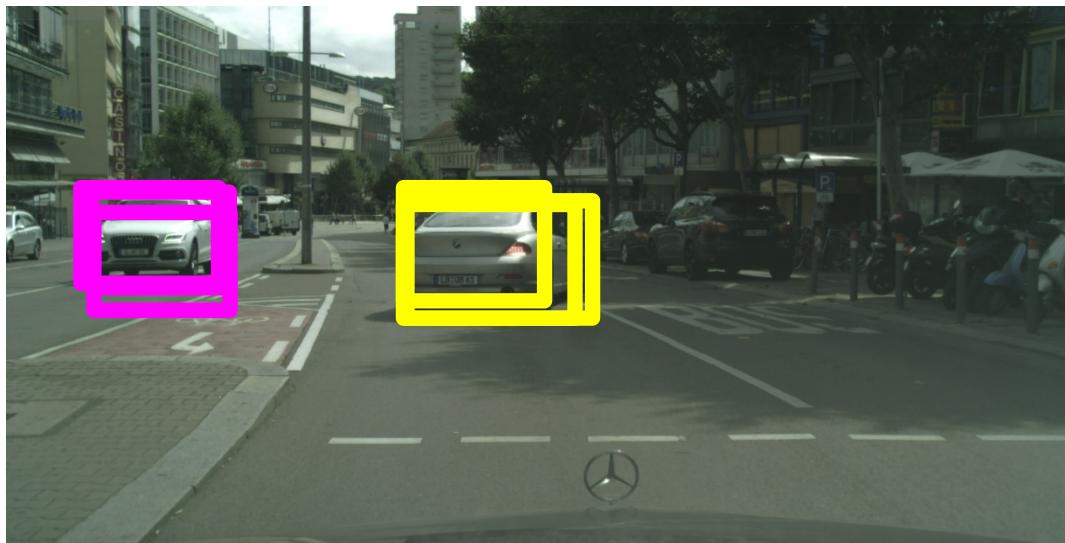
Object Tracking: Correlation

- Get **Measurement Bounding Boxes** from 2D detector.
- Correlate prediction with the **highest IOU** measurement



Object Tracking: Update

- The prediction and measurement are fused as part of the **Kalman Filter Framework**

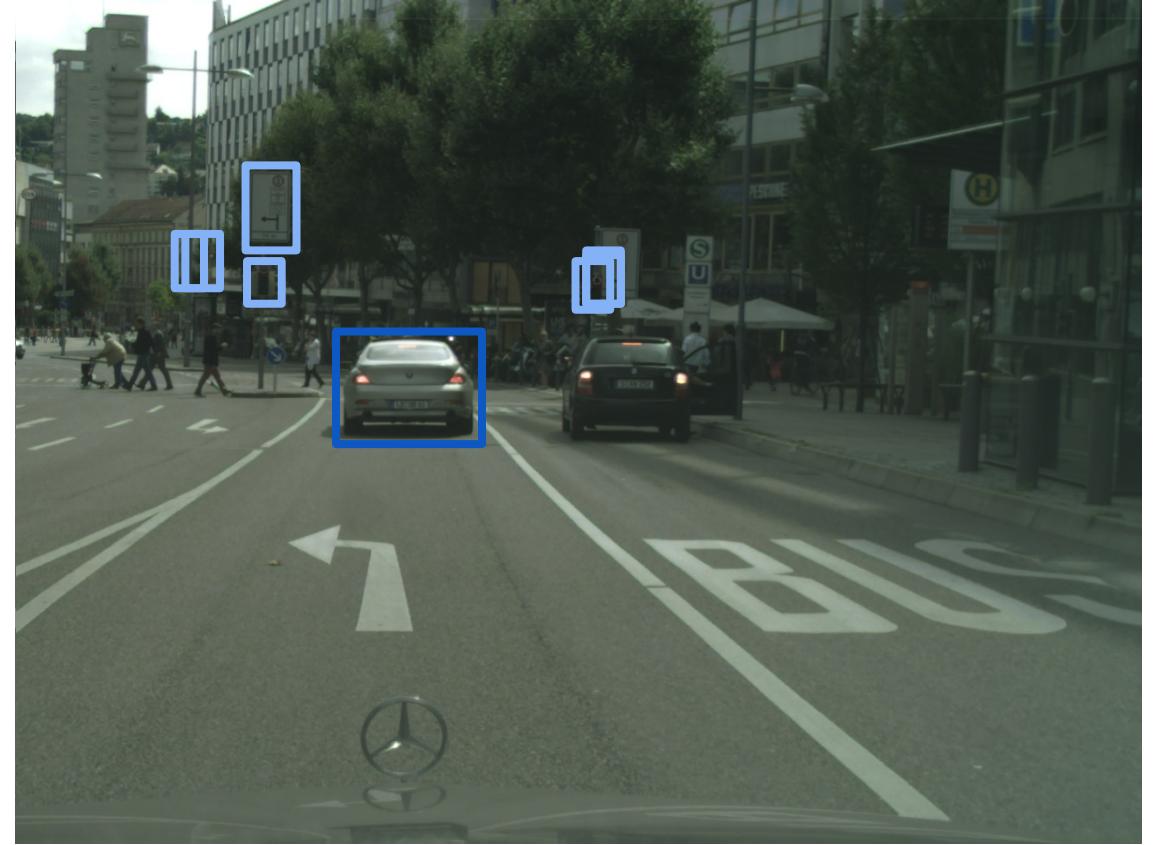


Object Tracking

- For each frame, we **start** new track if a measurement has no correlated prediction
- We also **terminate** inconsistent tracks, if a predicted object does not correlate with a measurement for a **preset** number of frames
- The same methodology can be used to track objects in 3D!

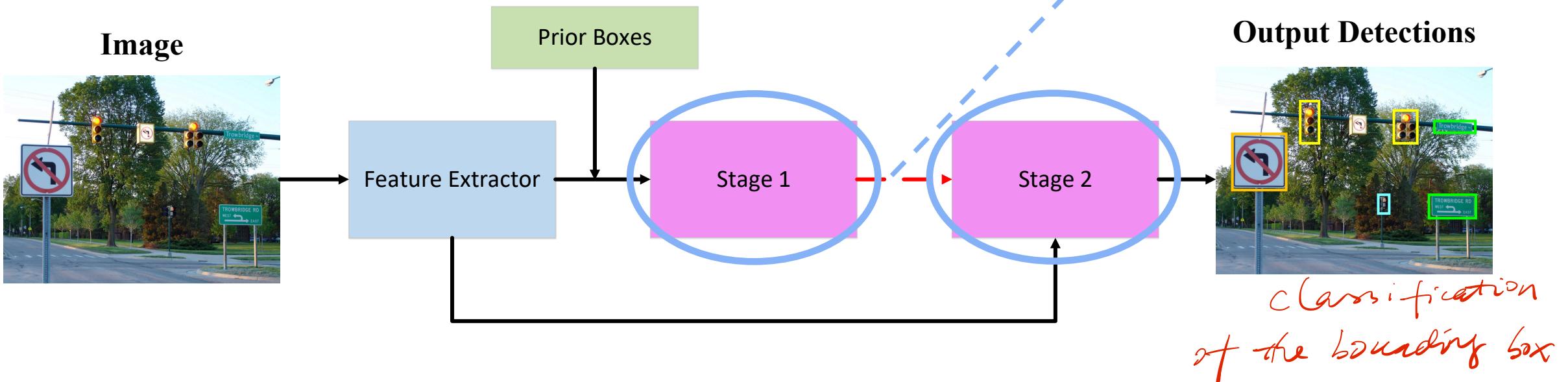
Traffic Sign and Traffic Signal Detection

- Traffic signs and signals appear smaller in size compared to cars, two-wheelers, and pedestrians.
- **Traffic signs** are highly variable with many classes to be trained on.
- **Traffic signals** have different states that are required to be detected.
- In addition, traffic signals change state as the car drives!



Traffic sign and signal detection

- 2D object detectors can be used to perform traffic sign and traffic signal detection without any modifications
- However, **multi-stage hierarchical models** have been shown to outperform the standard single stage object detectors



Summary

- The output of 2D object detectors can be extended to produce 3D object location and dimensions
- The output of 2D object detectors in consecutive frames can be used to track objects in 2D and in 3D
- The output of 2D object detectors can be used to detect traffic signs and signals, and determine the state of traffic signals