

Autonomous Valet Parking (AVP)

Theory and Practice

自主代客泊车理论与实践

Lecture 5: Parking Map Construction



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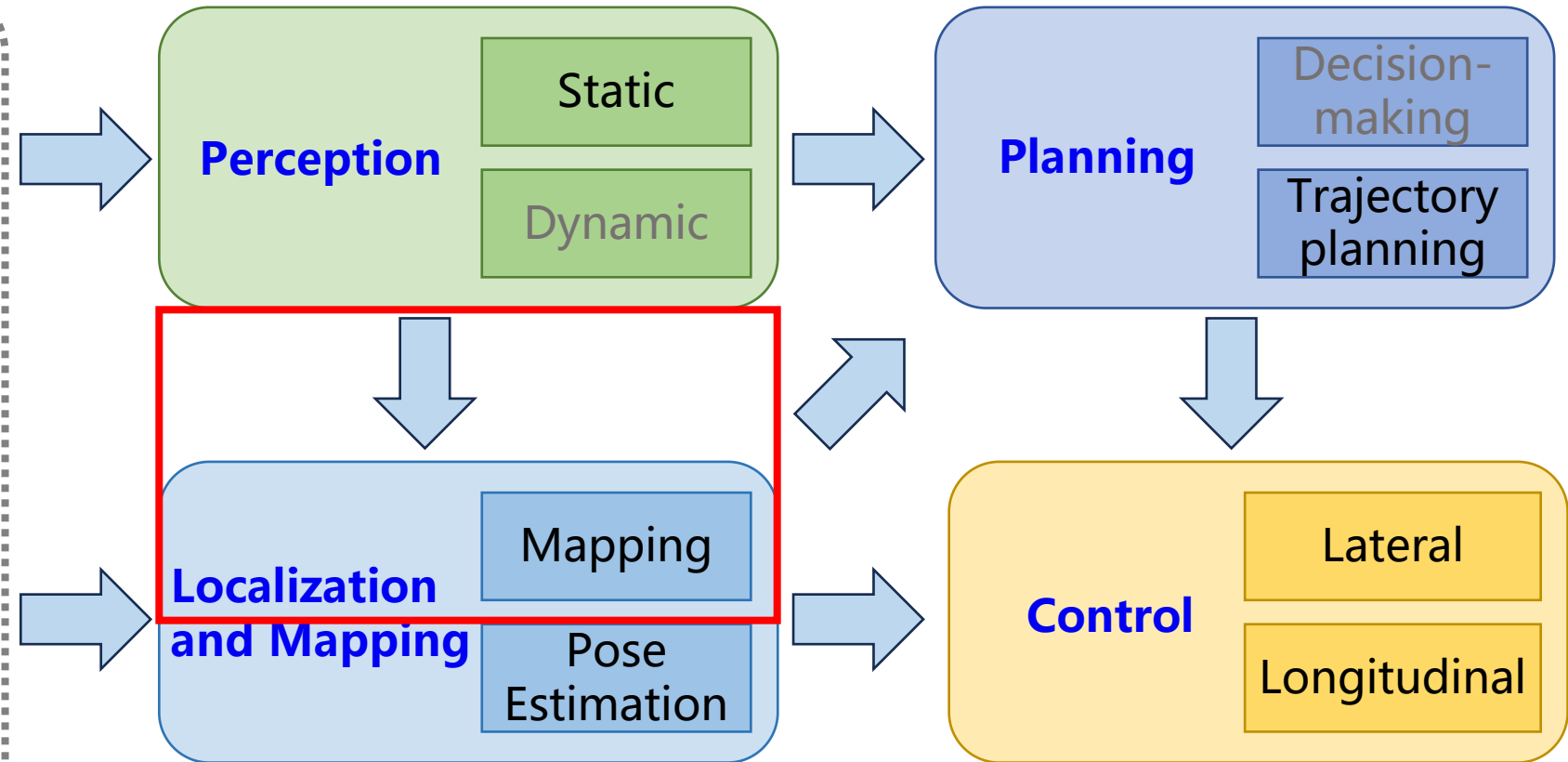
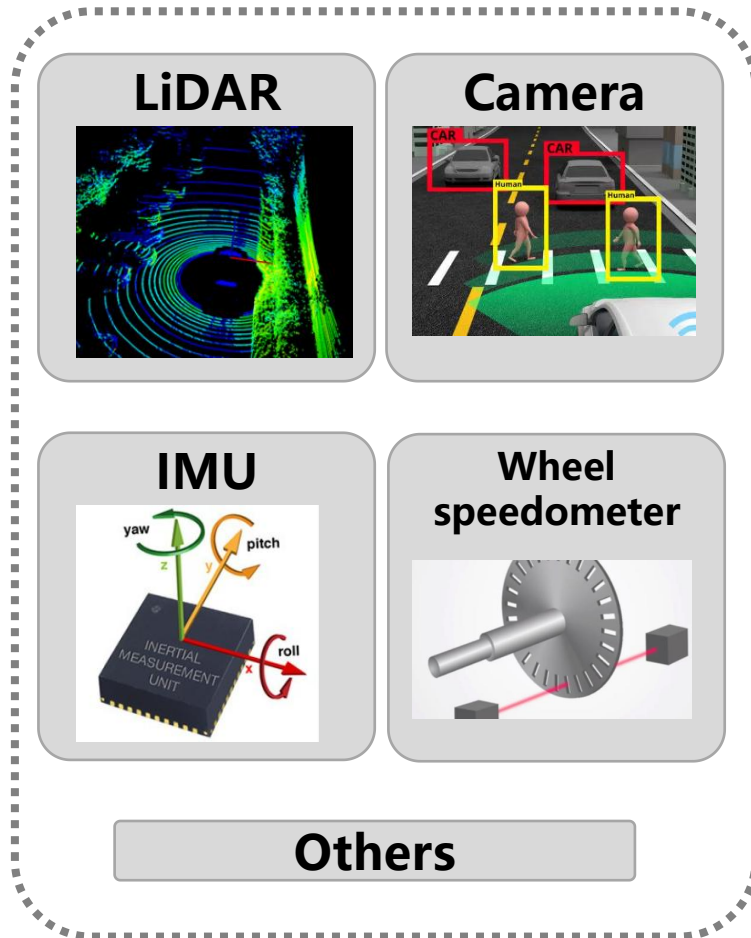








Content

AVP Architecture

Sensors

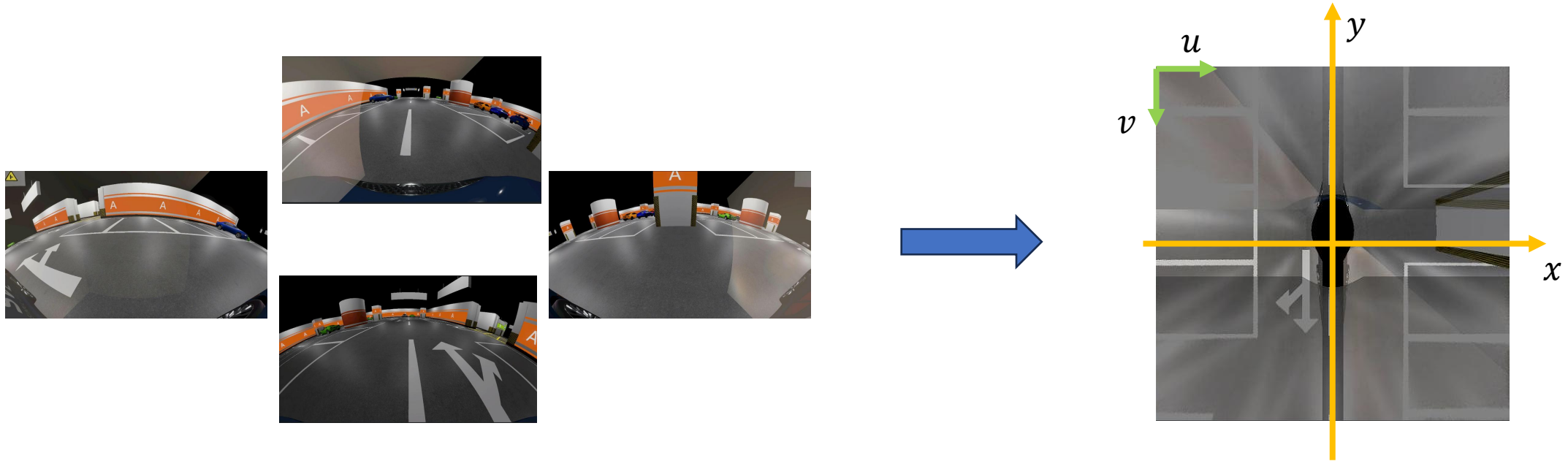




-  1. **Coordinate Transformation**
-  2. **Semantic Map Construction**
-  3. **Parking Spot Detection**
-  4. **Assignment**

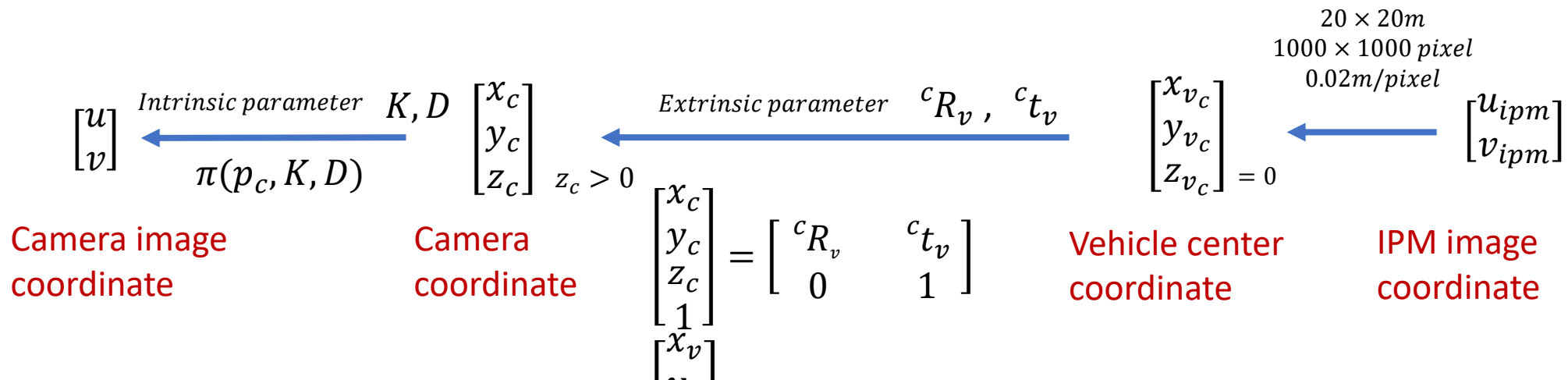


Recall



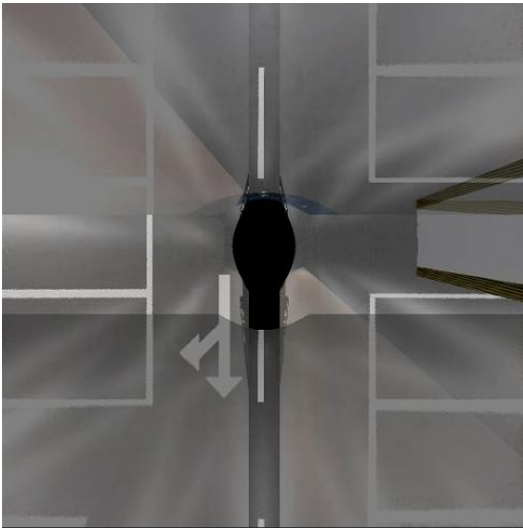
Camera image coordinate

IPM image coordinate

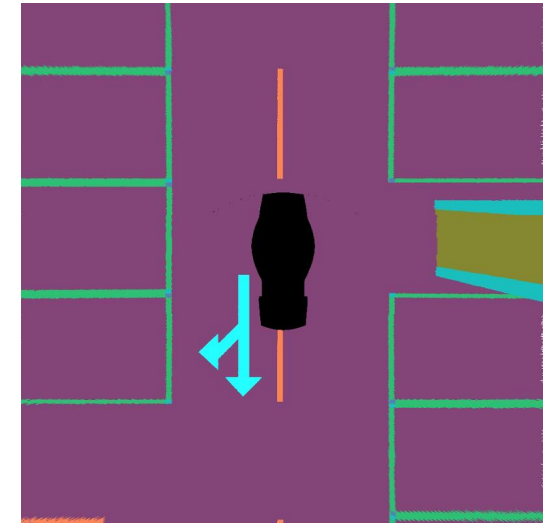
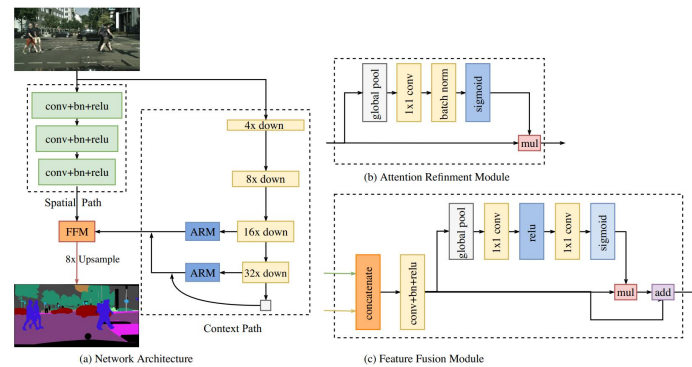




- Semantic feature detection



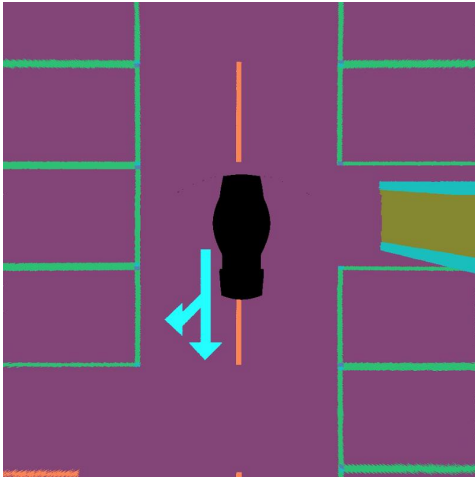
Semantic Segmentation



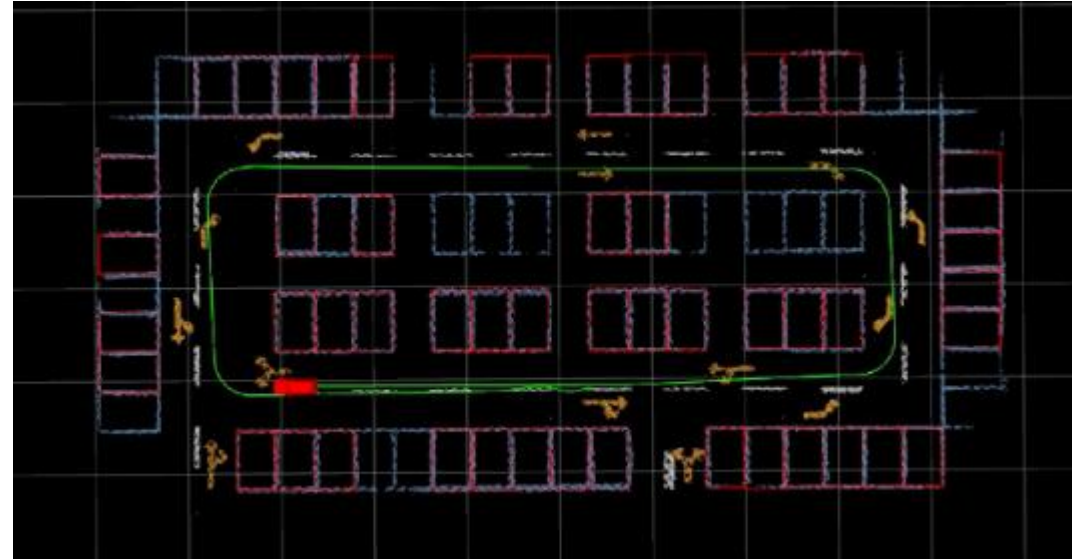


Coordinate Transformation

- Goal



IPM image
coordinate

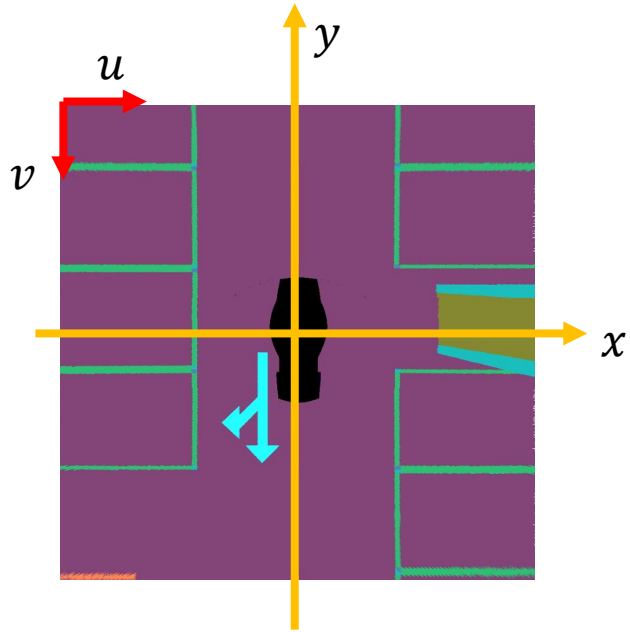


World (Global)
coordinate



Semantic Map Construction

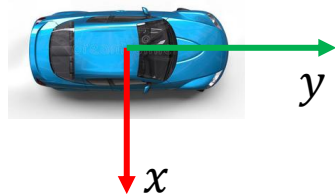
- Coordinates



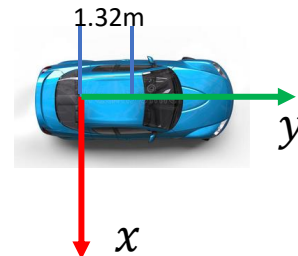
IPM image coordinate



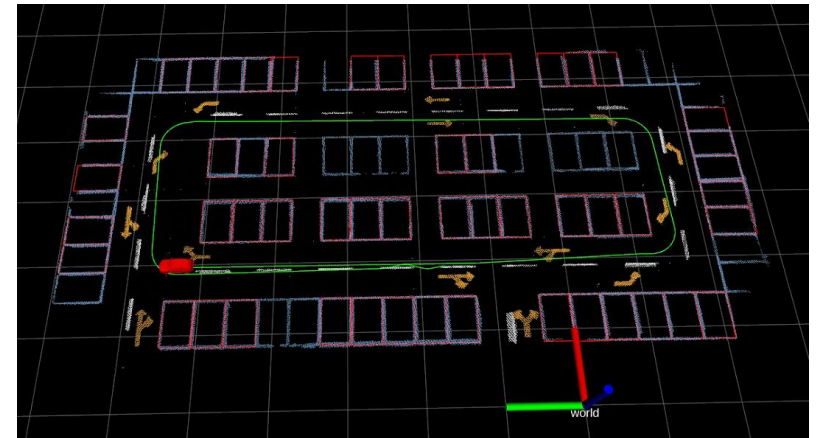
Vehicle center coordinate



Vehicle rear coordinate



World (Global) coordinate

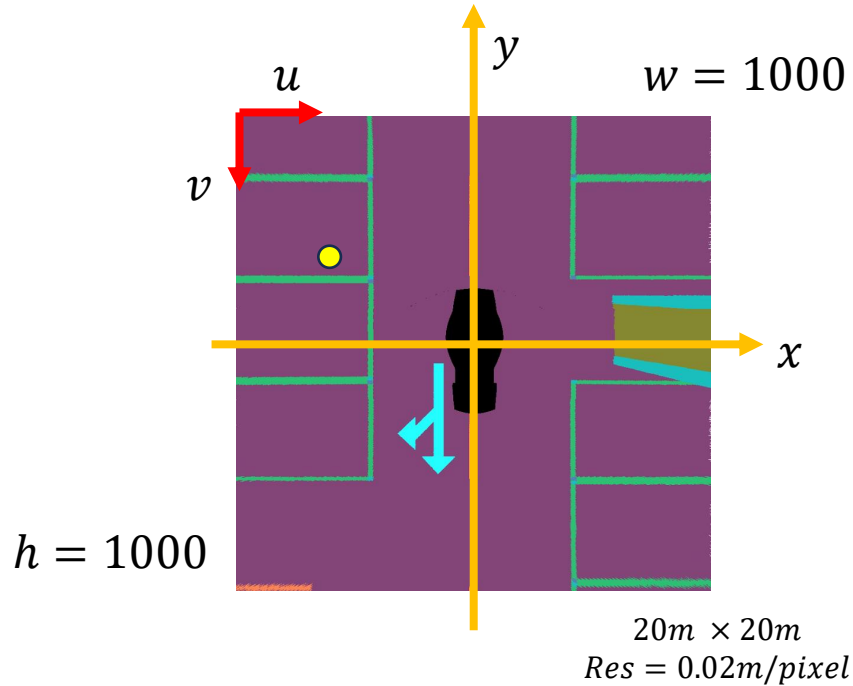




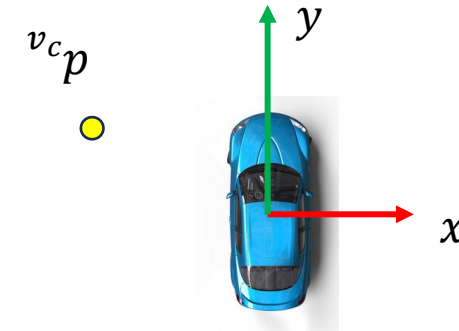
Coordinate Transformation

- Step1

IPM image
coordinate



Vehicle center
coordinate



1. Choose one point on IPM plane, for example: $\begin{bmatrix} u_{ipm} \\ v_{ipm} \end{bmatrix} = \begin{bmatrix} 200 \\ 400 \end{bmatrix}$

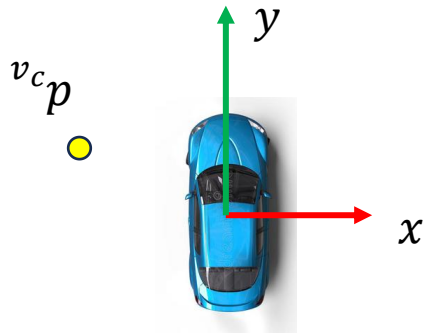
2. Vehicle center coordinate: $v_c p = \begin{bmatrix} x_{v_c} \\ y_{v_c} \\ z_{v_c} \end{bmatrix} = \begin{bmatrix} -(500 - u_{ipm}) * 0.02 \\ (500 - v_{ipm}) * 0.02 \\ 0 \end{bmatrix} = \begin{bmatrix} -6 \\ 2 \\ 0 \end{bmatrix}$



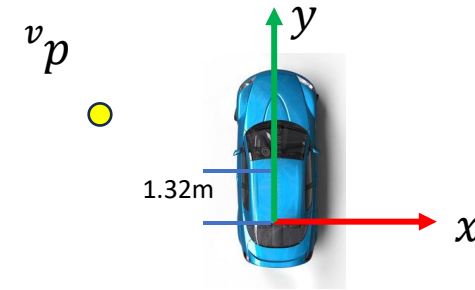
Coordinate Transformation

- Step2

Vehicle center
coordinate



Vehicle (rear)
coordinate



3. Vehicle (rear) coordinate: ${}^v p = \begin{bmatrix} x_v \\ y_v \\ z_v \end{bmatrix} = \begin{bmatrix} x_{v_c} \\ y_{v_c} + 1.32 \\ 0 \end{bmatrix} = \begin{bmatrix} -6 \\ 3.32 \\ 0 \end{bmatrix}$

$$\begin{bmatrix} x_v \\ y_v \\ z_v \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1.32 \\ 0 \\ 1 \end{bmatrix} \begin{bmatrix} x_{v_c} \\ y_{v_c} \\ z_{v_c} \\ 1 \end{bmatrix}$$

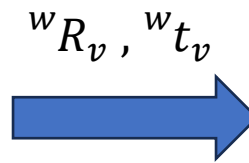
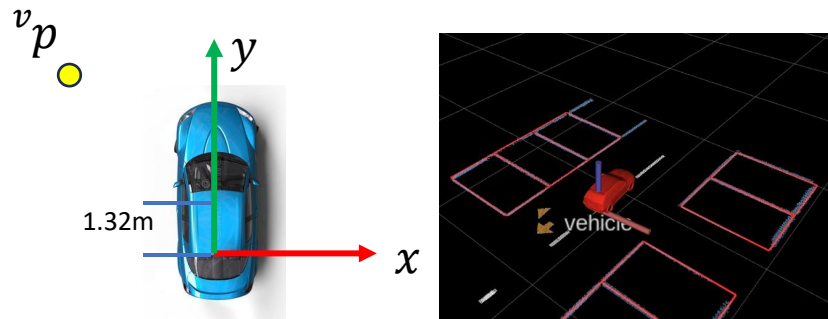


Coordinate Transformation

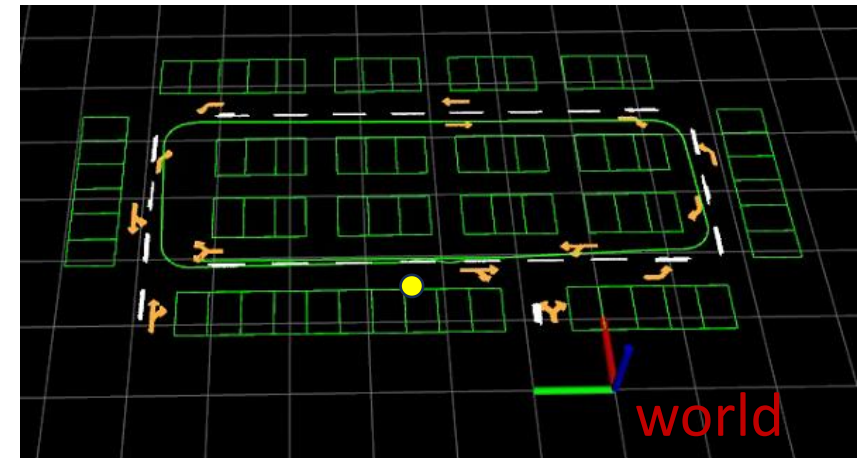
- Step3

- Assume we know vehicle's pose ${}^wR_v, {}^wt_v$
 - odometry

Vehicle rear
coordinate



World (Global)
coordinate

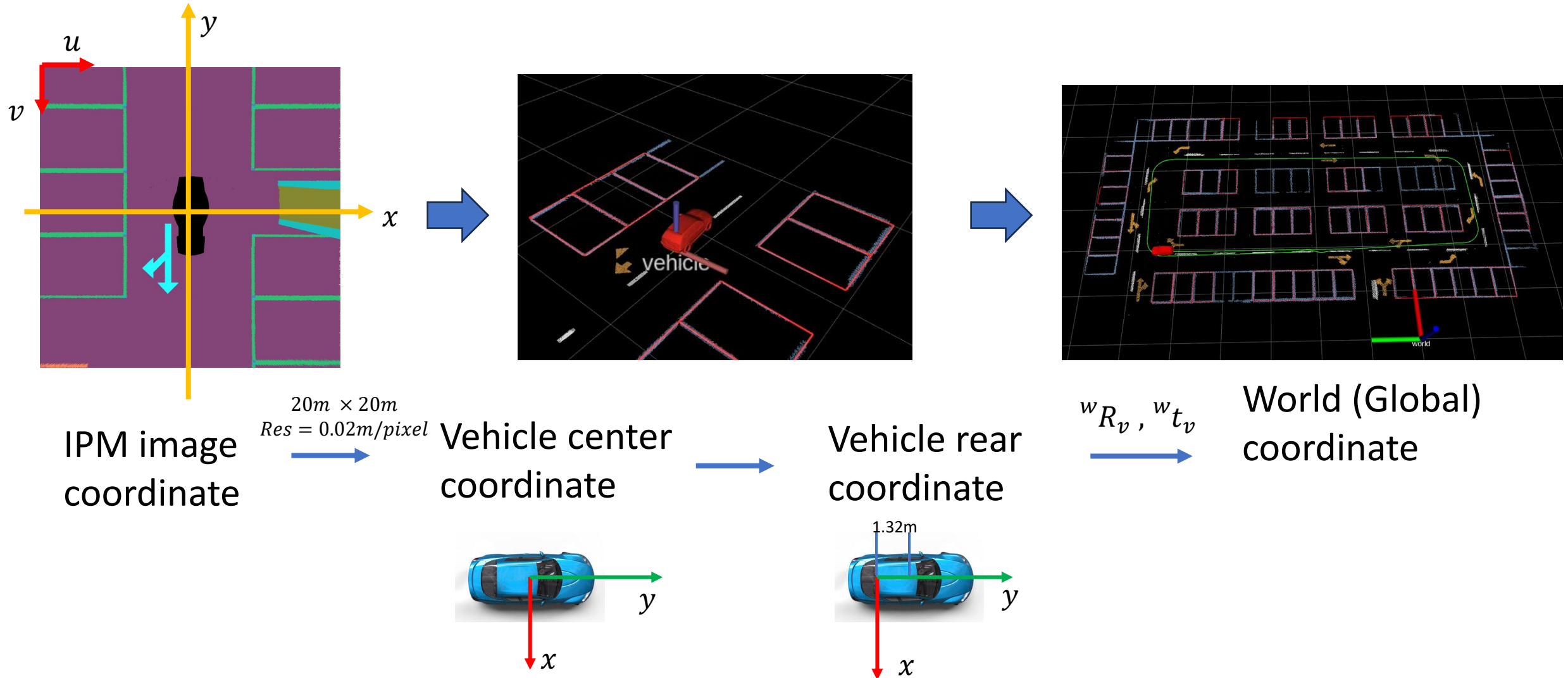


4. Vehicle (rear) coordinate to Global coordinate:
$$\begin{bmatrix} {}^wp \\ 1 \end{bmatrix} = \begin{bmatrix} {}^wR_v & {}^wt_v \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_v \\ y_v \\ z_v \\ 1 \end{bmatrix} {}^vp$$







Coordinate Transformation

- Pipeline



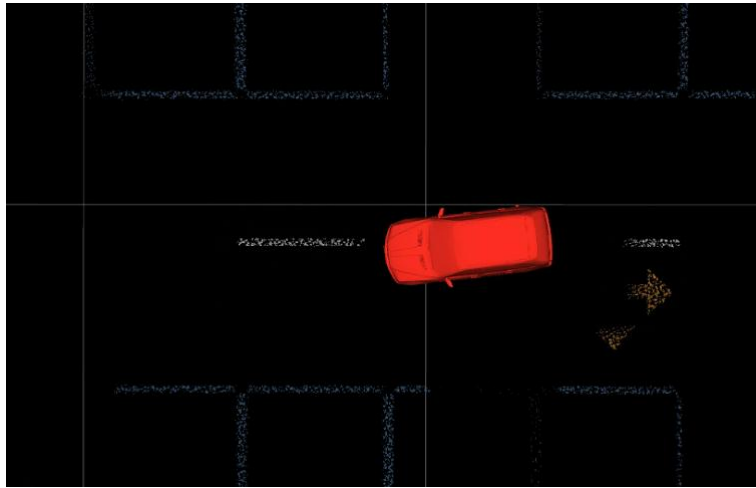


-  1. Coordinate Transformation
-  2. Semantic Map Construction
-  3. Parking Spot Detection
-  4. Assignment

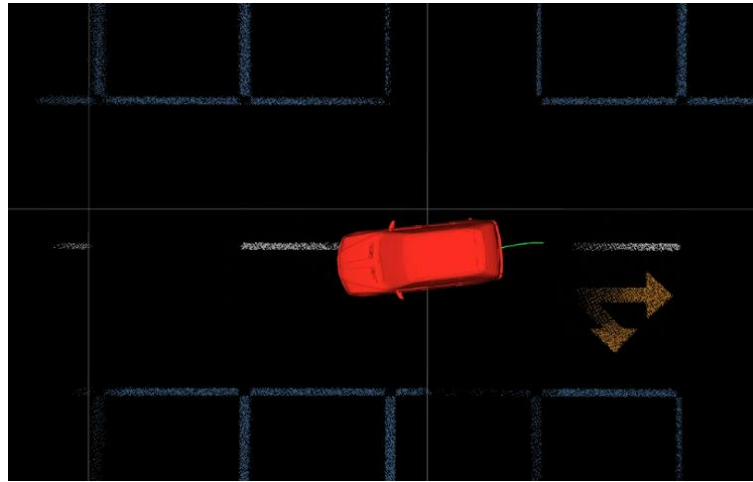


Semantic Map Construction

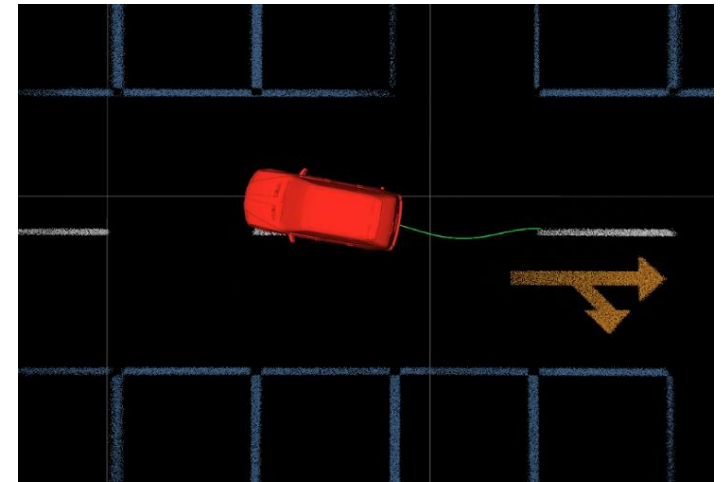
- Map increasing



t



$t + 1$



$t + 2$

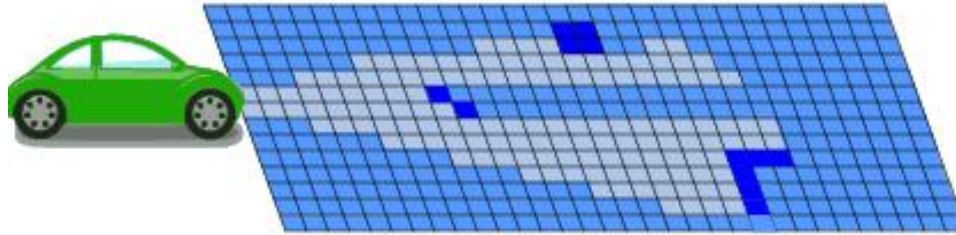
Semantic points increase linearly

- Redundant points

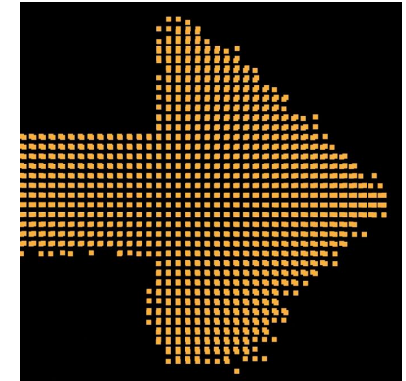
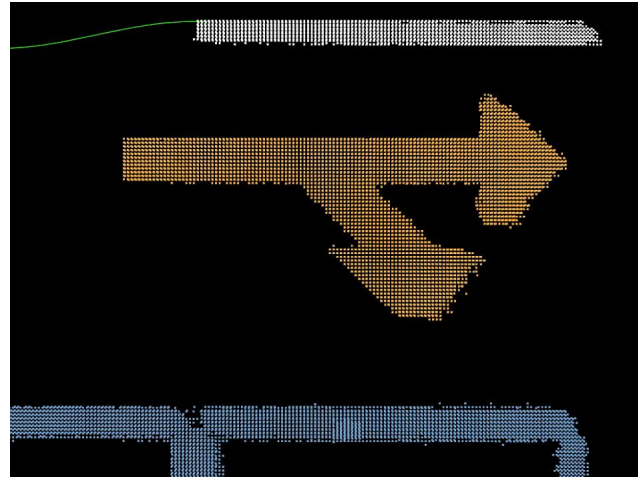
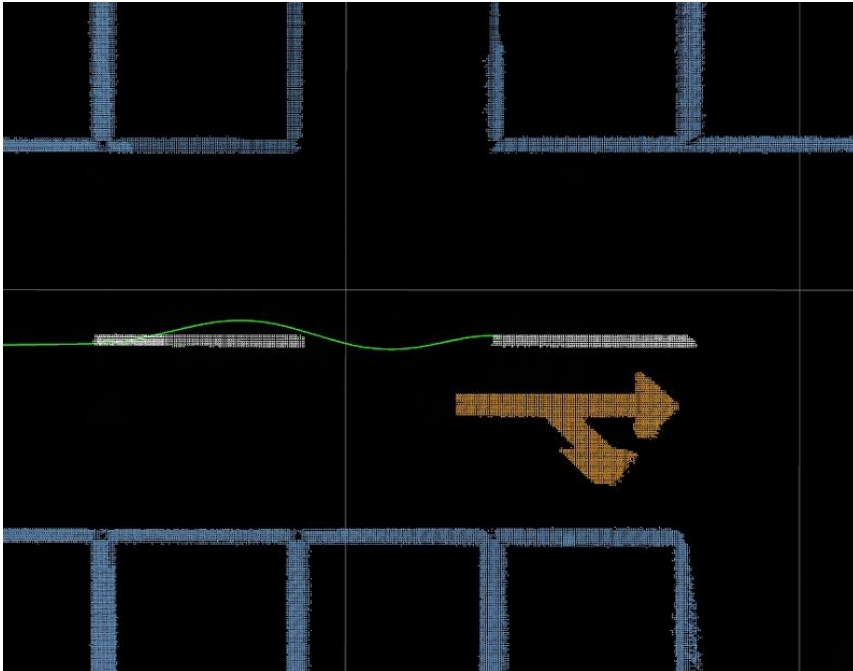


Semantic Map Construction





- Map sparsity
 - Grid map



$0.04 \times 0.04m$
One label in one grid



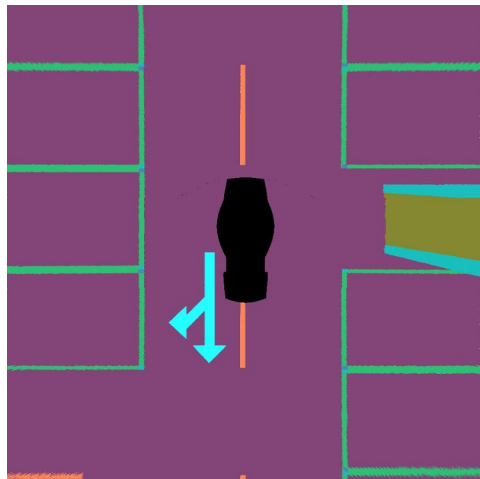


-  1. Coordinate Transformation
-  2. Semantic Map Construction
-  3. **Parking Spot Detection**
-  4. Assignment

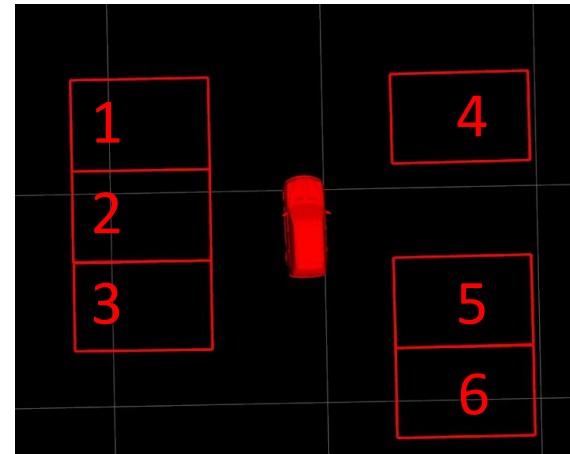


Parking Spot Detection

- Goal
 - Extract the instance of parking spot



points
标量点

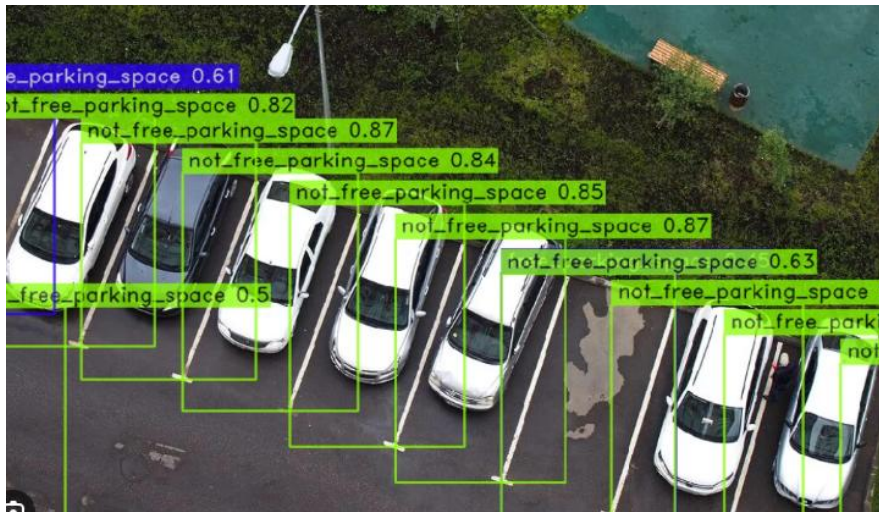


instance
实例

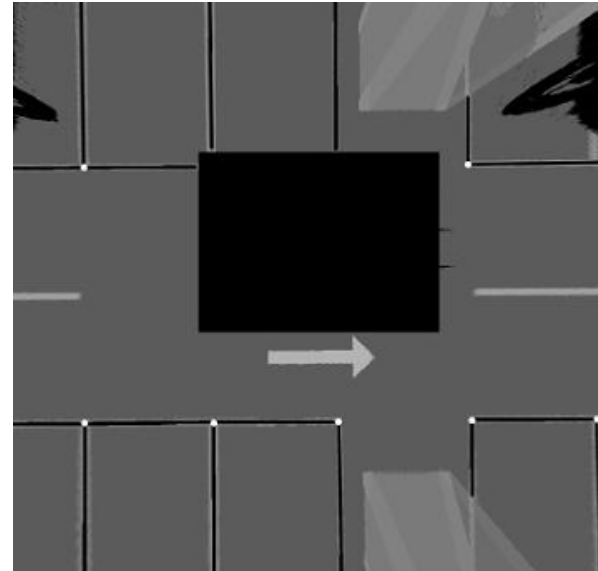


Parking Spot Detection

- Method
 - Neural network-based method



- Rule-based method



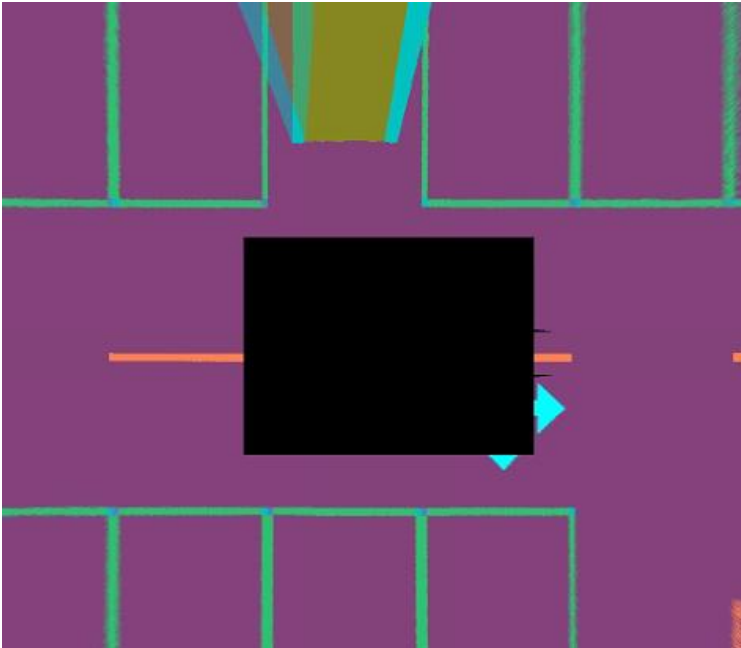
Geometric characteristic

- Line->Corner->Rectangle

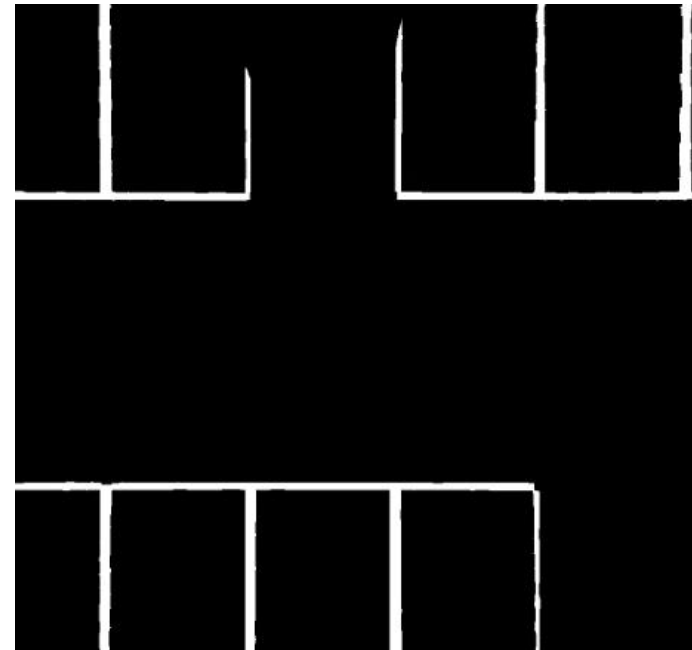


Parking Spot Detection

- Rule-based method
 - Line extraction



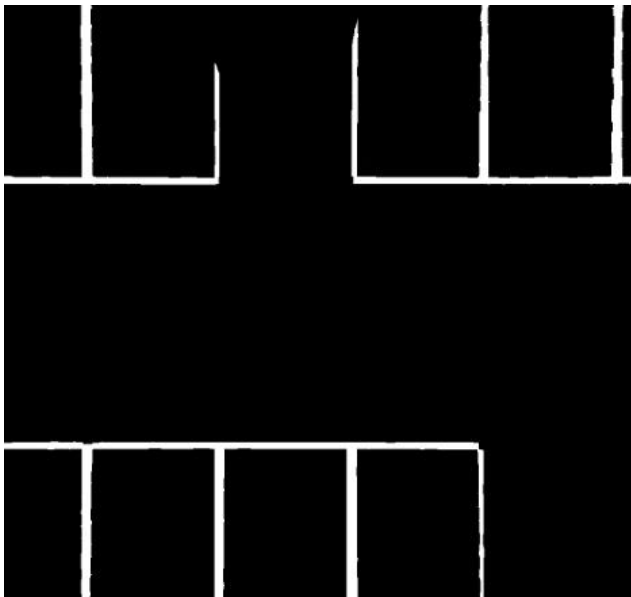
Label = slot line



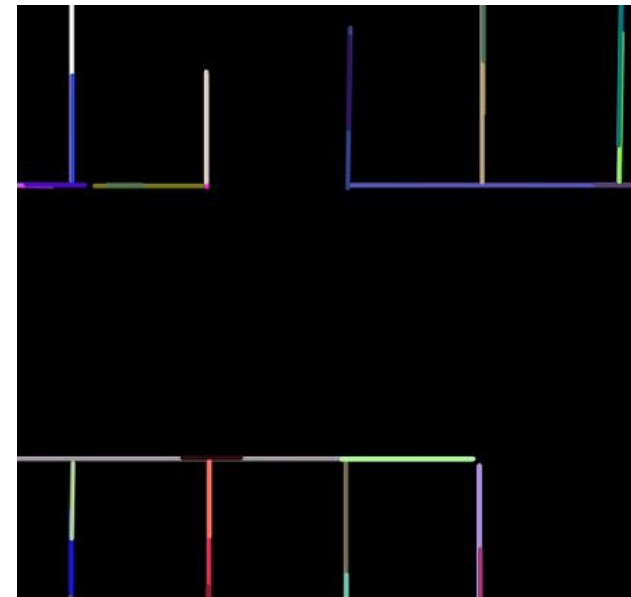


Parking Spot Detection

- Rule-based method
 - Line extraction



Hough Transform



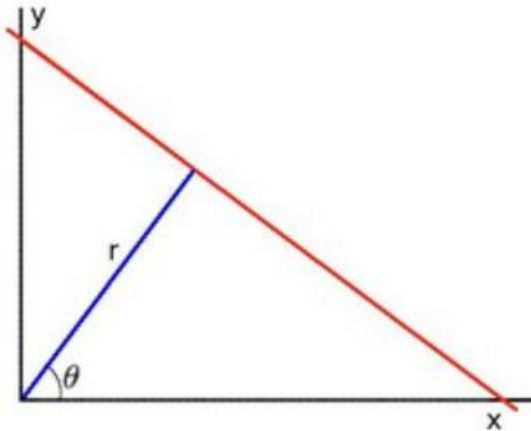


Parking Spot Detection

1. As you know, a line in the image space can be expressed with two variables. For example:

a. In the **Cartesian coordinate system**: Parameters: (m, b) .

b. In the **Polar coordinate system**: Parameters: (r, θ)



For Hough Transforms, we will express lines in the *Polar system*. Hence, a line equation can be written as:

$$y = \left(-\frac{\cos \theta}{\sin \theta} \right) x + \left(\frac{r}{\sin \theta} \right)$$

Arranging the terms: $r = x \cos \theta + y \sin \theta$

1. In general for each point (x_0, y_0) , we can define the family of lines that goes through that point as:

$$r_{\theta} = x_0 \cdot \cos \theta + y_0 \cdot \sin \theta$$

Meaning that each pair (r_{θ}, θ) represents each line that passes by (x_0, y_0) .

- Rule-based method
 - Line extraction
 - Hough Transform



Parking Spot Detection

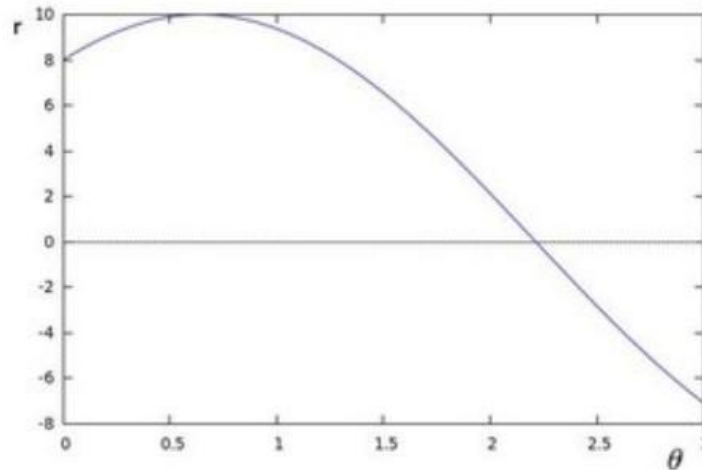
Arranging the terms: $r = x \cos \theta + y \sin \theta$

1. In general for each point (x_0, y_0) , we can define the family of lines that goes through that point as:

$$r_\theta = x_0 \cdot \cos \theta + y_0 \cdot \sin \theta$$

Meaning that each pair (r_θ, θ) represents each line that passes by (x_0, y_0) .

2. If for a given (x_0, y_0) we plot the family of lines that goes through it, we get a sinusoid. For instance, for $x_0 = 8$ and $y_0 = 6$ we get the following plot (in a plane $\theta - r$):



We consider only points such that $r > 0$ and $0 < \theta < 2\pi$.

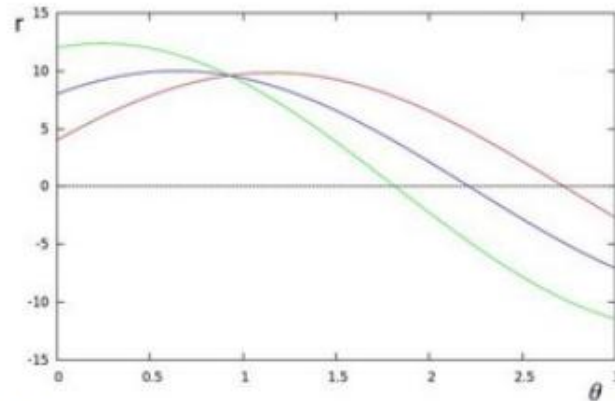
- Rule-based method
 - Line extraction
 - Hough Transform



Parking Spot Detection

- Rule-based method
 - Line extraction
 - Hough Transform

3. We can do the same operation above for all the points in an image. If the curves of two different points intersect in the plane $\theta - r$, that means that both points belong to a same line. For instance, following with the example above and drawing the plot for two more points: $x_1 = 4$, $y_1 = 9$ and $x_2 = 12$, $y_2 = 3$, we get:



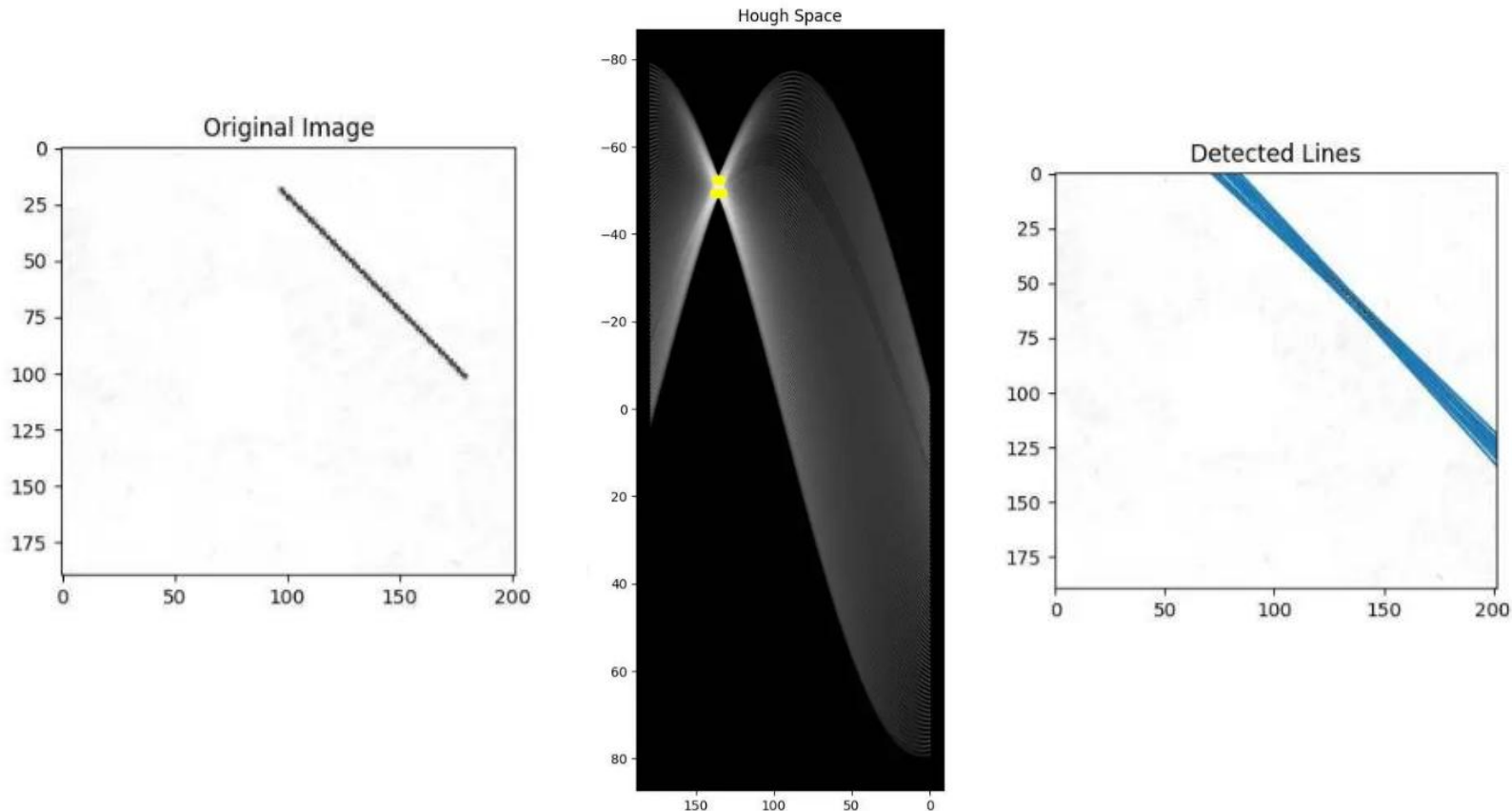
The three plots intersect in one single point $(0.925, 9.6)$, these coordinates are the parameters (θ, r) or the line in which (x_0, y_0) , (x_1, y_1) and (x_2, y_2) lay.



Parking Spot Detection

- Rule-based method
 - Line extraction
 - Hough Transform

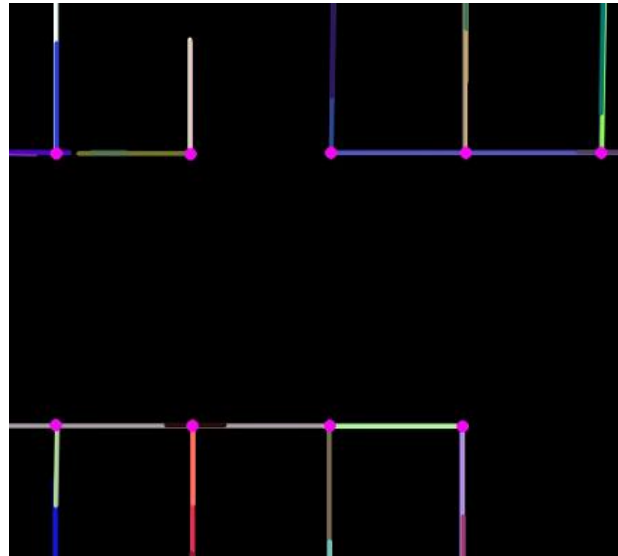
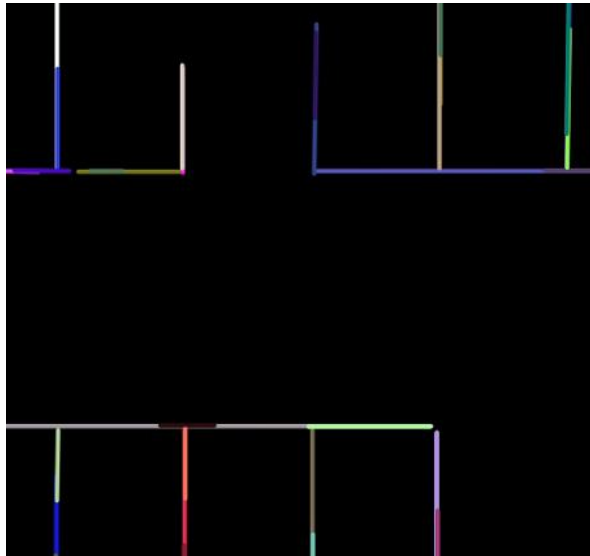
`cv::HoughLinesP`





Parking Spot Detection

- Rule-based method
 - Line extraction
 - Corner extraction
- Iterate every two lanes
 - Distance threshold
 - Intersection angle threshold ($\sim 90^\circ$)
 - Check intersection point label
 - Merge too-closed corner points



Intersection angle

$$\cos \theta = \frac{\overrightarrow{A_1 A_2} \cdot \overrightarrow{B_1 B_2}}{|\overrightarrow{A_1 A_2}| |\overrightarrow{B_1 B_2}|}$$

$$a_1 x + b_1 y = c_1$$

$$a_2 x + b_2 y = c_2$$



Intersection point

$$x = \frac{c_1 b_2 - c_2 b_1}{a_1 b_2 - a_2 b_1}$$

$$y = \frac{a_1 c_2 - a_2 c_1}{a_1 b_2 - a_2 b_1}$$

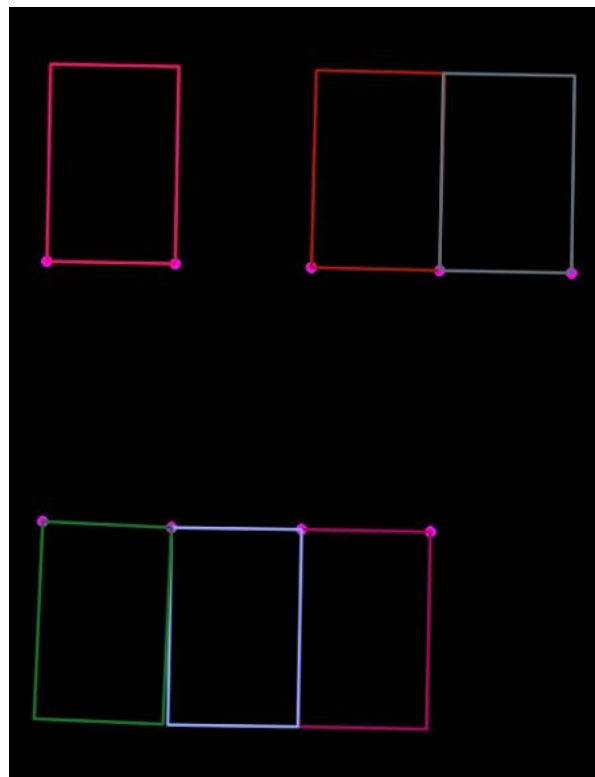
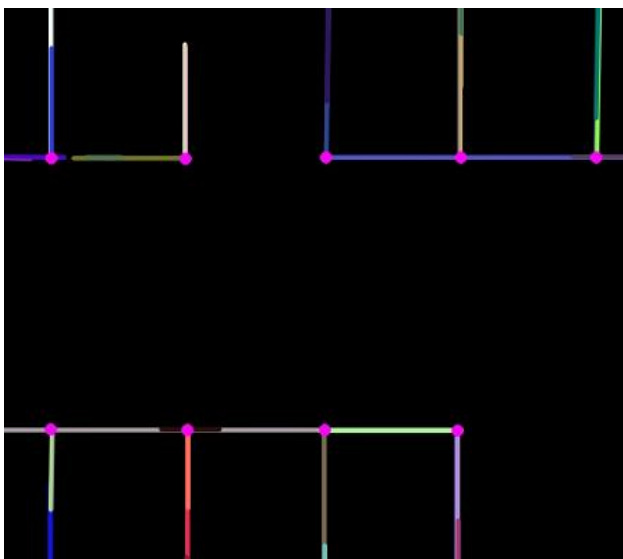


Parking Spot Detection

Slot: 6.4x4.2m 210x320pixel

- Rule-based method
 - Line extraction
 - Corner extraction
 - Slot fitting

- Iterate every two corners
 - 两点距离是否和车位短边接近? (4.2m)
 - & 两点组成的短边是否有一定比例的点的label为库位线
 - 计算短边对应的两个垂线 (注意垂线正、反方向)
 - 垂线上是否有大量的点的label为库位线
 - 若是, 根据车位长度 (6.4m) 推测另外两个库位角点

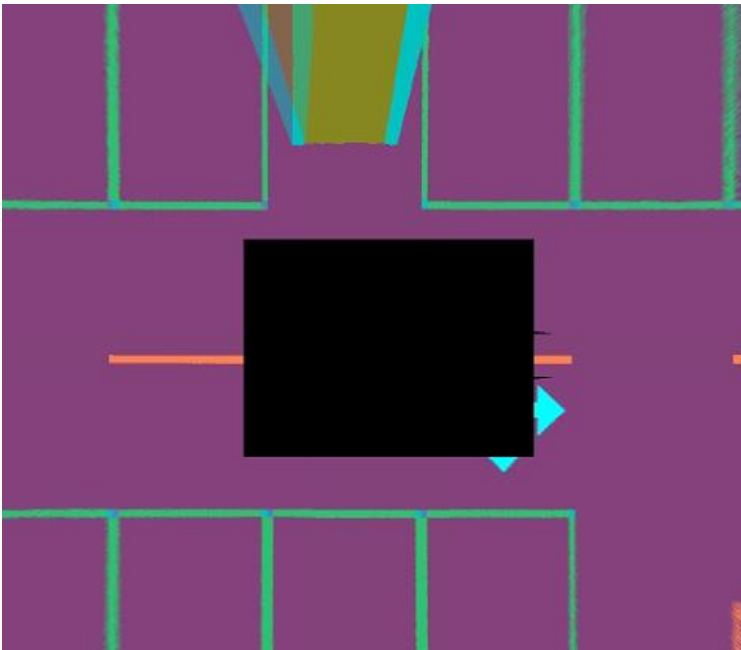




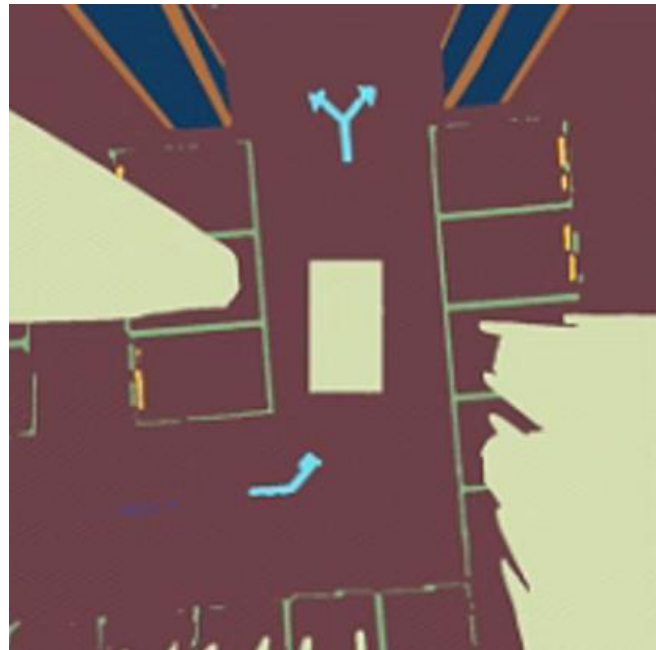
Parking Spot Detection

- Rule-based method
 - Line extraction
 - Corner extraction
 - Slot fitting

Ideal case



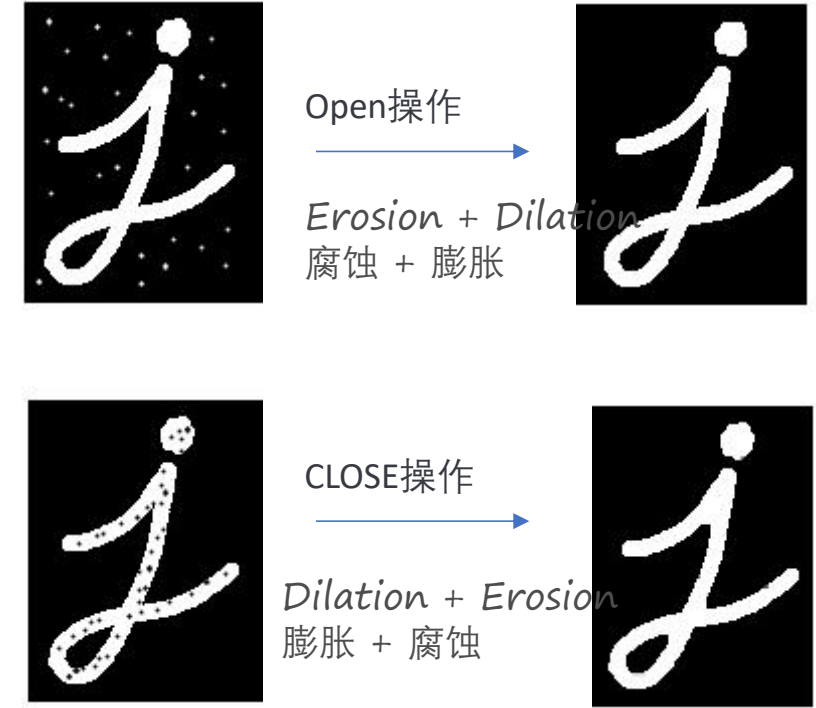
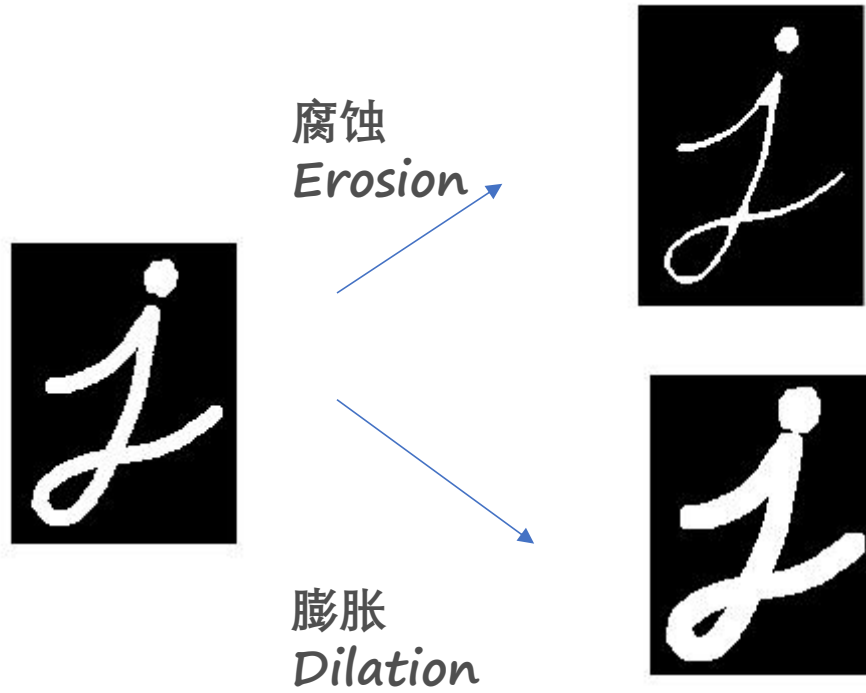
Real case : Noise and blur









Parking Spot Detection

- Rule-based method
 - Enhancement and denoising
 - Line extraction
 - Corner extraction
 - Slot fitting





-  1. Coordinate Transformation
-  2. Semantic Map Construction
-  3. Parking Spot Detection
-  4. Assignment



Assignment

Building your AVP map

Input:

- Segmentation result for every IPM image
- Vehicle pose (vehicle rear in global frame) wR_v , wt_v

• Complete C++ functions:

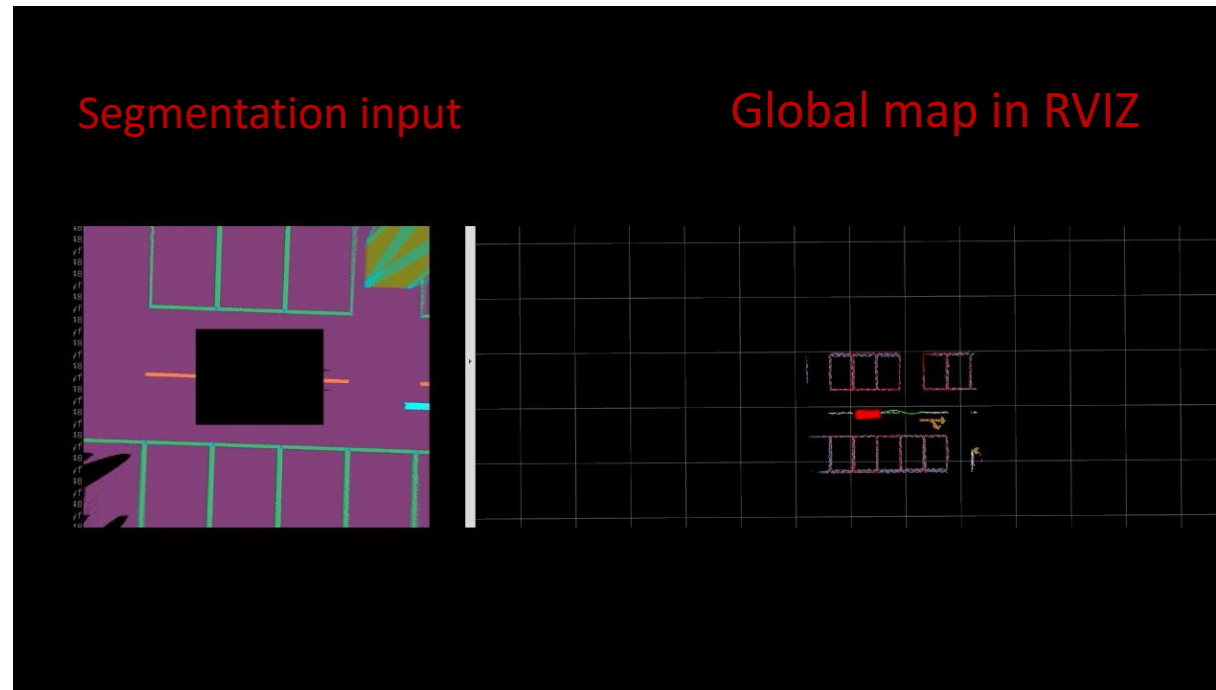
- *ipmPlane2Global()*

- Transformation from IPM image plane to global coordinate

- *detectSlot()*

- Extract slot from lanes

Expected Output:



感谢聆听！
Thanks for Listening

