3D Vehicle Reconstruction via Monocular Camera with Deep Learning Models and Direct Linear Transformation

2024-2 Robot Vision (M3228.003000)

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Abstract—Hello here goes the abstract

I. INTRODUCTION

[1] (Leave this citation for now, otherwise it breaks (if we have 0 citations, and the command at the end))

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II. 3D POSE ESTIMATION

 $(A=U\Sigma V^T)$ [2] In the SVD $Ab=U\Sigma V^Tb=0$ implies $V^Tb=0$ since U is invertible and has no nontrivial solutions for Ub=0. Thus V^Tb must be a vector where only the rows corresponding to zero singular values contribute to the null space. The columns of V that correspond to zero singular values in Σ combined form a basis for the null space of A.

$$\begin{bmatrix} R & t \\ ---- & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

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#Pseudo code of the matching process #cars keeps track of all the cars in the scene #cars_n contains all the cars in the new frame for old_car in cars:

c, d = closest_car(cars_n, old_car)
if d > max_distance:
 #the car left the frame
 cars.remove(old_car)
else:
 old.match(c)
 #every car can only be matched once
 cars_n.remove(new_car)

#all remaining not matched new cars
for new_car in cars_n:
 #a new car has entered the frame
 cars.append(new_car)

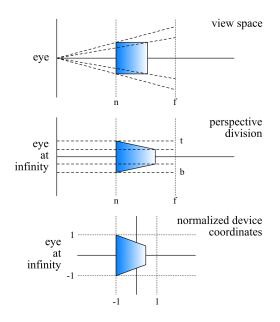


Fig. 1. Perspective projection from view space to NDC

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$$FOV = 2 \cdot \arctan\left(\frac{s_w}{2f}\right), r = n \cdot \tan\left(\frac{FOV}{2}\right)$$
 (1)

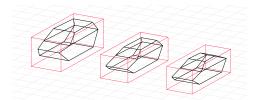


Fig. 2. Different car wireframes used. From left to right Ford Explorer, Volvo V60, Nissan Altima.

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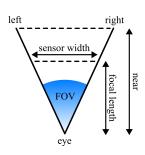


Fig. 3. The sensor width and the focal length are used to calculate the camera's field of view.

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Matrix representation of the standard projection onto the z = 1 plane.

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \\ z \end{bmatrix} \stackrel{\text{div z}}{\equiv} \begin{bmatrix} \frac{x}{z} \\ \frac{y}{z} \\ 1 \end{bmatrix}$$
 (2)

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The API projection Matrix

$$M_{P} = M_{P''}M_{P'} = \begin{bmatrix} \frac{2}{r-l} & 0 & 0 & \frac{-(r+l)}{r-l} \\ 0 & \frac{2}{t-b} & 0 & \frac{-(t+b)}{t-b} \\ 0 & 0 & \frac{2}{f-n} & \frac{-(f+n)}{f-n} \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} n & 0 & 0 & 0 \\ 0 & n & 0 & 0 \\ 0 & 0 & n+f & -nf \\ 0 & 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} c_{1} & 0 & 0 & 0 \\ 0 & c_{2} & 0 & 0 \\ 0 & 0 & c_{3} & c_{4} \\ 0 & 0 & 1 & 0 \end{bmatrix}$$
(3)

To get the screen-space position of a vertex, we multiply it by its Modelview Matrix and API Projection Matrix before doing the projective division. We have all the parameters except the Model Matrix. $p'=M_{API}M_{Modelview}p$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} \stackrel{\text{div w}}{\equiv} \begin{bmatrix} c_1 & 0 & 0 & 0 \\ 0 & c_2 & 0 & 0 \\ 0 & 0 & c_3 & c_4 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

If we multiply that out, we get for x' and y':

$$x' = \frac{c_1 ax + c_1 by + c_1 cz + c_1 d}{ix + jy + kz + l}$$
$$y' = \frac{c_2 ex + c_2 fy + c_2 gz + c_2 h}{ix + jy + kz + l}$$

We can multiply by the denuminator and solve for 0. Since this holds for every pair of points $(p_1, p'_1), (p_2, p'_2), ..., (p_n, p'_n)$ that we track we can create the following system of equation: 4

REFERENCES

- [1] J.-B. Michel, Y. K. Shen, A. P. Aiden, A. Veres, M. K. Gray, T. G. B. Team, J. P. Pickett, D. Hoiberg, D. Clancy, P. Norvig, J. Orwant, S. Pinker, M. A. Nowak, and E. L. Aiden, "Quantitative analysis of culture using millions of digitized books," *Science*, vol. 331, no. 6014, pp. 176–182, 2011. [Online]. Available: https://www.science.org/doi/abs/10.1126/science.1199644
- [2] "Singular value decomposition," https://en.wikipedia.org/wiki/Singular_value_decomposition, accessed: 2024-12-06.