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OF ENGINEERING



Robot Vision

Two View Geometry & RGBD

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Overview

- RGBD
 - Two-view geometry
 - Epipolar geometry
 - Fundamental/Essential matrix
 - RGBD sensors and principles
- Vanishing Points



References

- Hartley & Zisserman 2003:
 - Section 9.1, 9.2, 9.5, 9.6
- Corke 2011:
 - Section 14.2, 14.3
- Forsyth & Ponce 2011
 - Chapter 7
- Szeliski 2011:
 - Section 11.1, 12.2, 12.3



Let us start from the movie: Lord of the Rings

- What is the paradox when comparing the two images?





Have you heard of the Ames Room?



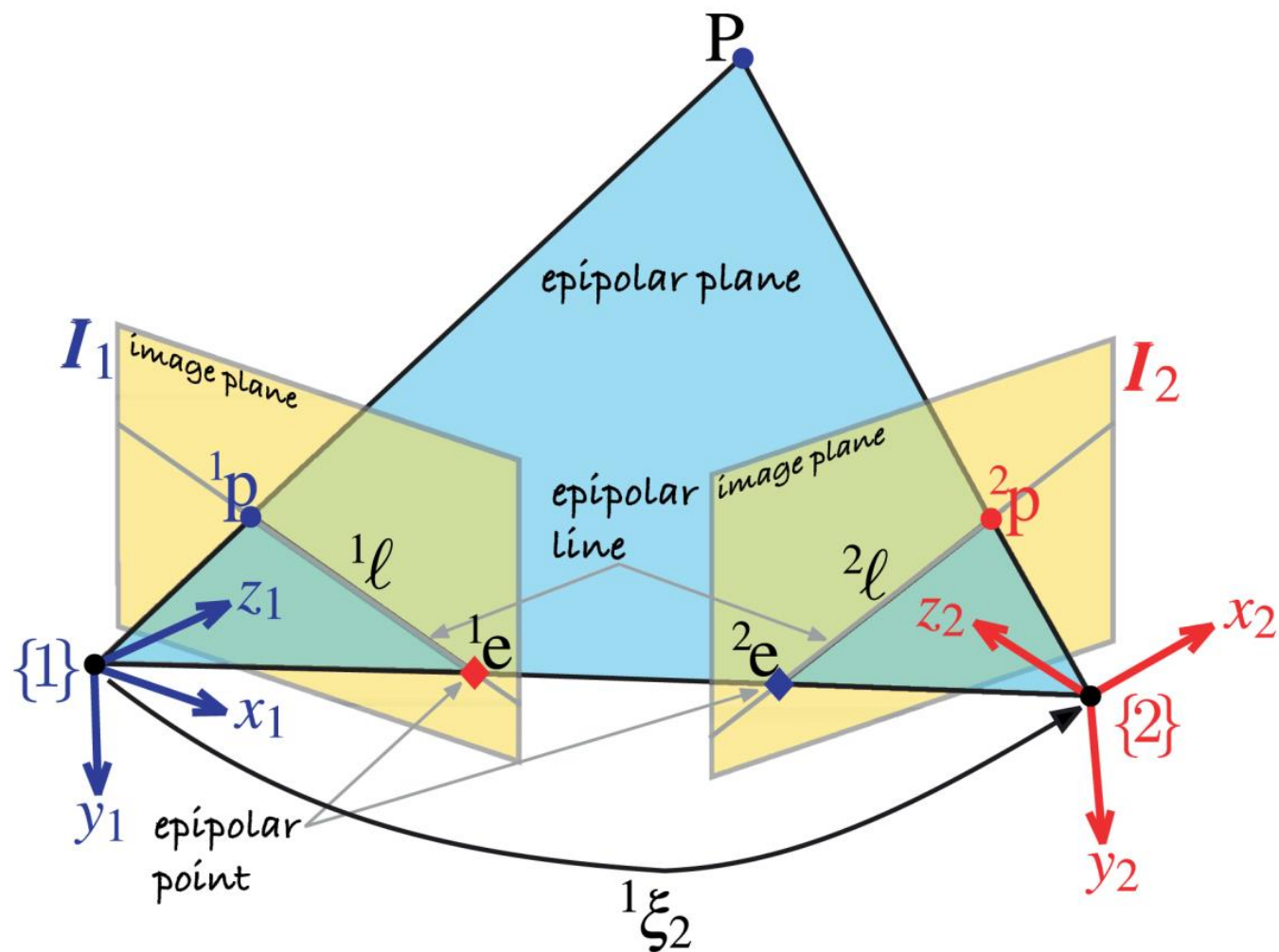


How did they implement the illusion in the movie?



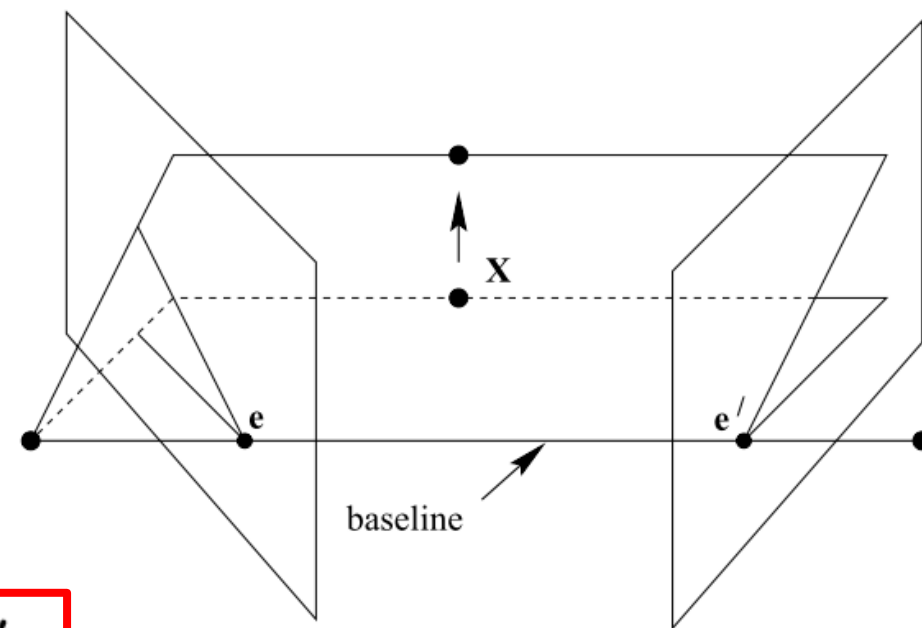
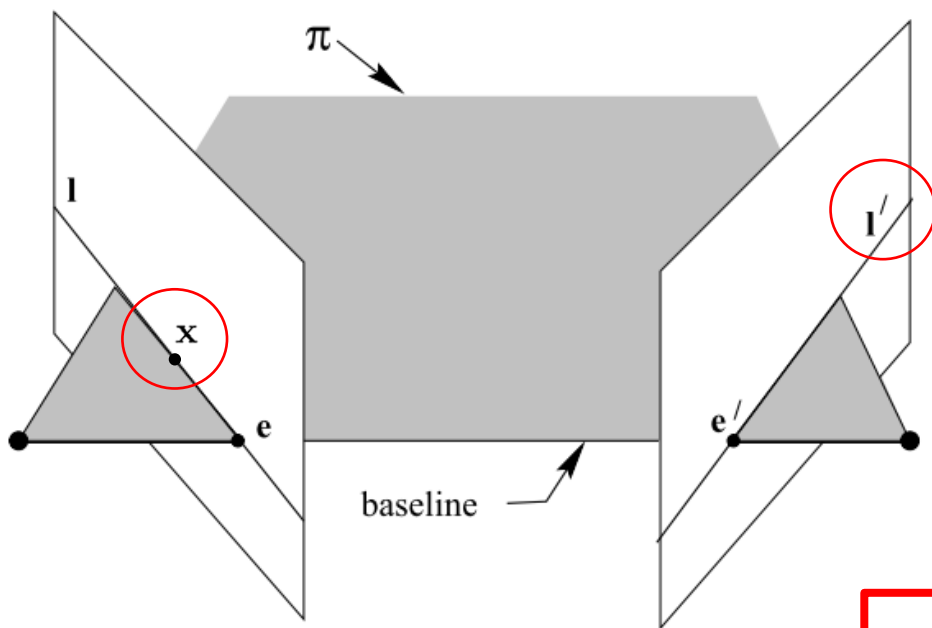
https://youtu.be/QWMFpxkGO_s

Two-view Geometry



Epipolar Geometry

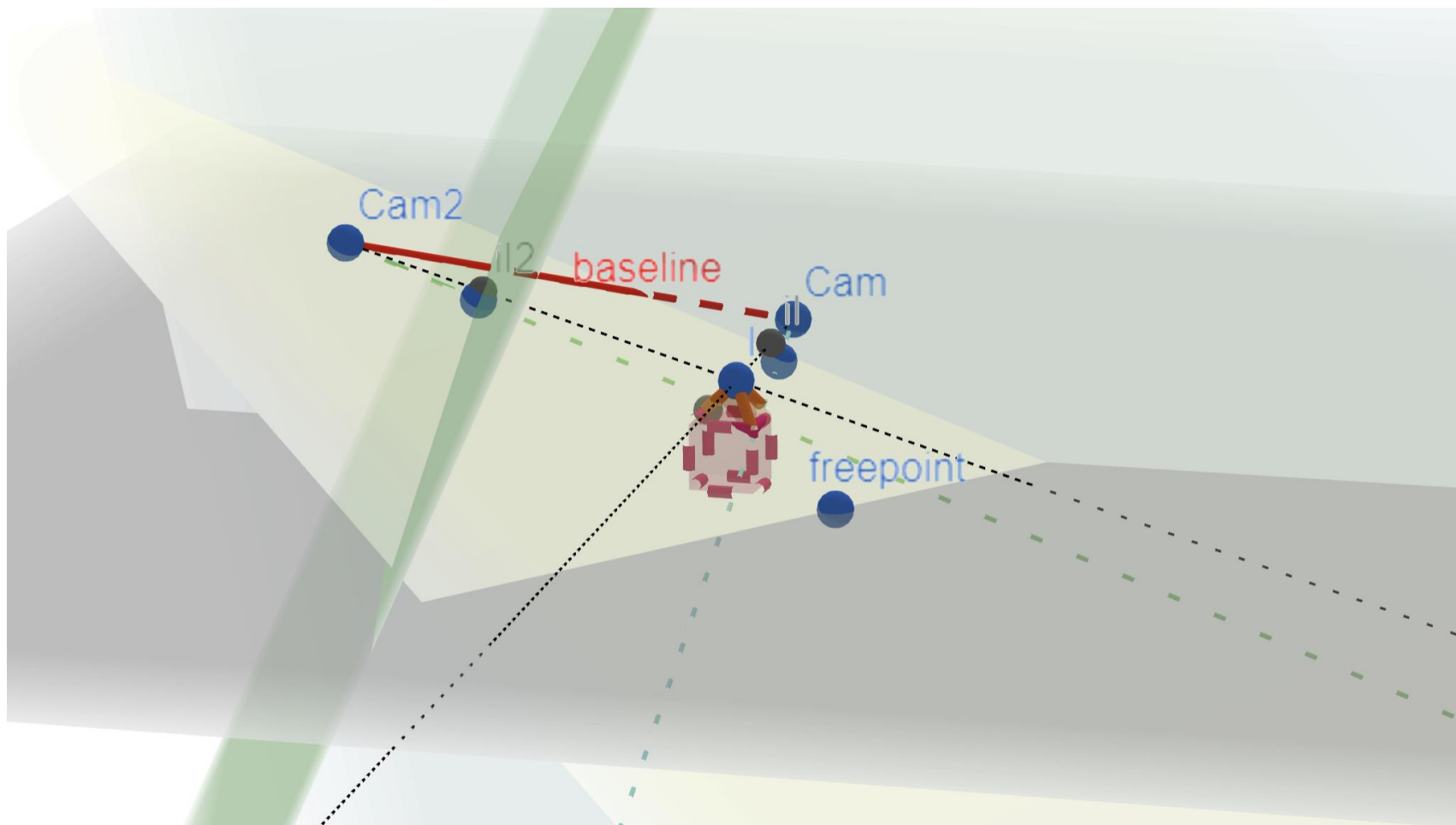
- Independent of scene structure
- Only depends on the **cameras' internal parameters and relative pose**



$$x \mapsto l'$$



Epipolar Geometry: Interactive Demo





Fundamental Matrix

Result 9.3. *The fundamental matrix satisfies the condition that for any pair of corresponding points $\mathbf{x} \leftrightarrow \mathbf{x}'$ in the two images*

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0.$$

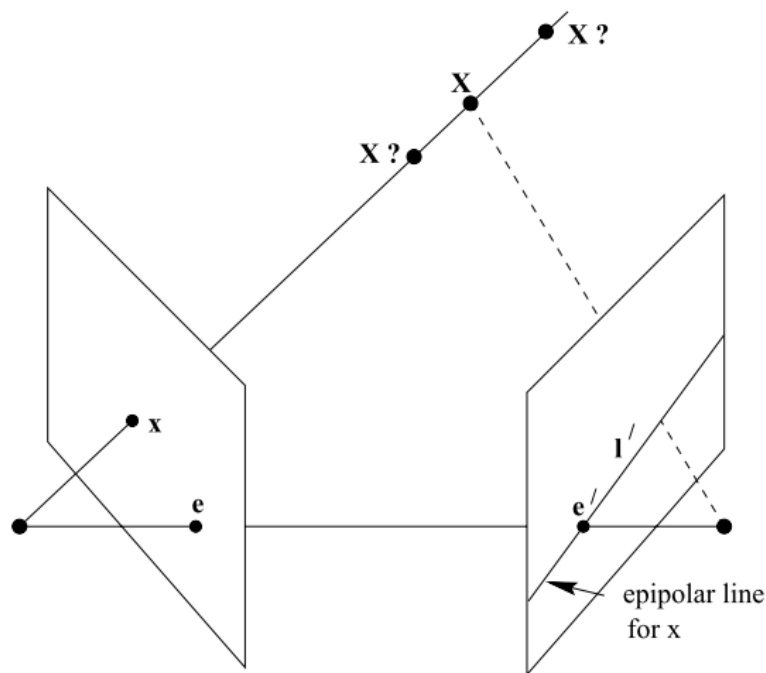
[LonguetHiggins-81] H. C. Longuet-Higgins. A computer algorithm for reconstructing a scene from two projections. *Nature*, 293:133–135, September 1981.



Fundamental Matrix

Result 9.3. *The fundamental matrix satisfies the condition that for any pair of corresponding points $\mathbf{x} \leftrightarrow \mathbf{x}'$ in the two images*

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0.$$



$$\mathbf{x} \mapsto \mathbf{l}'$$



Normalized Coordinate and Essential Matrix

$$\mathbf{x} = \mathbf{P}\mathbf{X}$$

$$\mathbf{P} = \mathbf{K}[\mathbf{R} \mid \mathbf{t}]$$

Normalized coordinate $\hat{\mathbf{x}} = \mathbf{K}^{-1}\mathbf{x} = [\mathbf{R} \mid \mathbf{t}]\mathbf{X}$

$$\hat{\mathbf{x}}'^T \mathbf{E} \hat{\mathbf{x}} = 0$$

Result 9.17. *A 3×3 matrix is an essential matrix if and only if two of its singular values are equal, and the third is zero.*

$$\mathbf{E} = \mathbf{K}'^T \mathbf{F} \mathbf{K}$$

$$\boxed{\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}}$$



Estimating F-matrix

- Find multiple $\mathbf{X} \leftrightarrow \mathbf{X}'$ correspondences (≥ 7) between two images

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0$$

$$x'x f_{11} + x'y f_{12} + x' f_{13} + y'x f_{21} + y'y f_{22} + y' f_{23} + x f_{31} + y f_{32} + f_{33} = 0$$

$$\mathbf{A} \mathbf{f} = \begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_n x_n & x'_n y_n & x'_n & y'_n x_n & y'_n y_n & y'_n & x_n & y_n & 1 \end{bmatrix} \mathbf{f} = \mathbf{0}$$

- Enforce rank-2 constraint by SVD



The Fundamental Matrix Song – Daniel Wedge

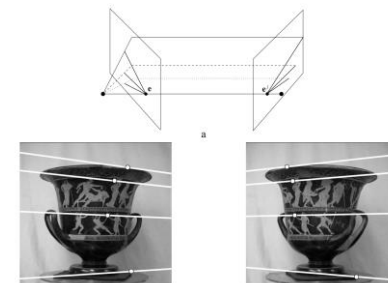


Video from: <https://youtu.be/DgGV3l82NTk>

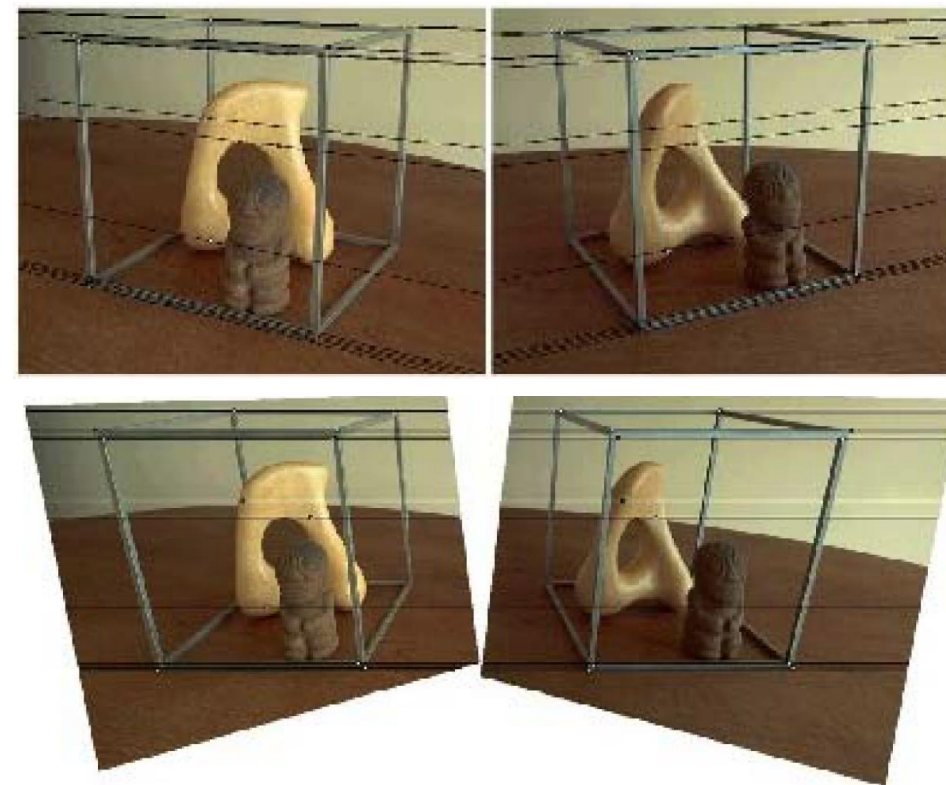


Depth from Stereo Images

- Epipolar geometry: stereo calibration
- Finding disparity by searching along epipolar lines
 - Low texture: bad
 - Repeated texture: bad



Hartley & Zisserman 2003

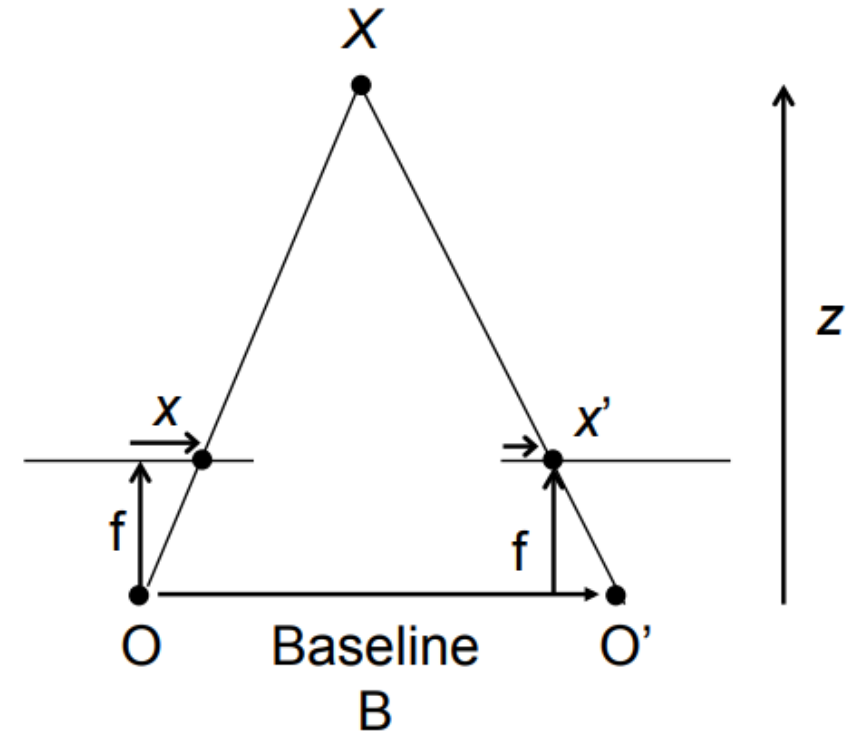


Szeliski 2011

Depth from Stereo Images

- Disparity and depth
 - Inversely related
- Stereo camera infers pixel depth from disparity
- Longer baseline==better depth accuracy

$$\frac{x - x'}{O - O'} = \frac{f}{z}$$

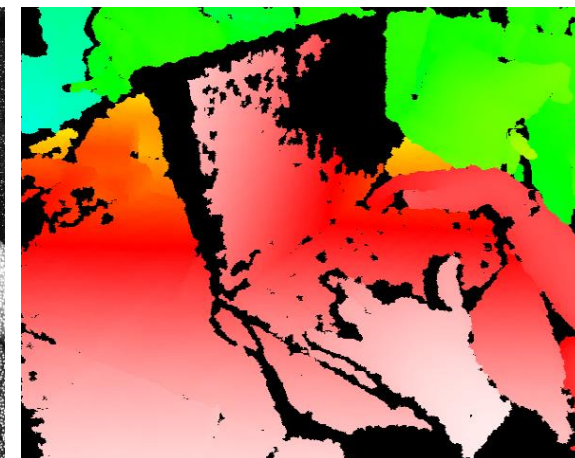
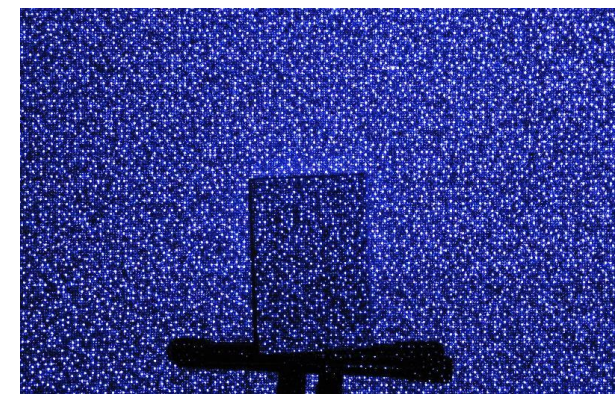
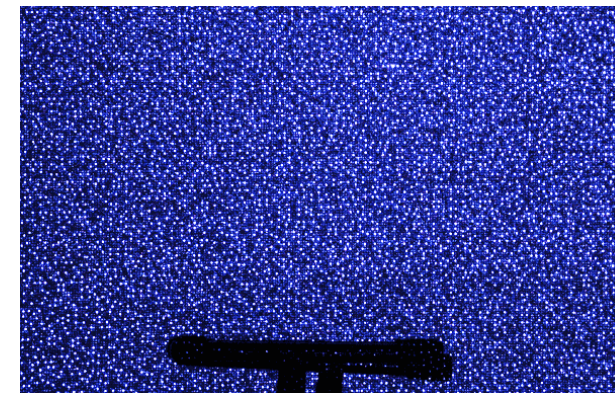
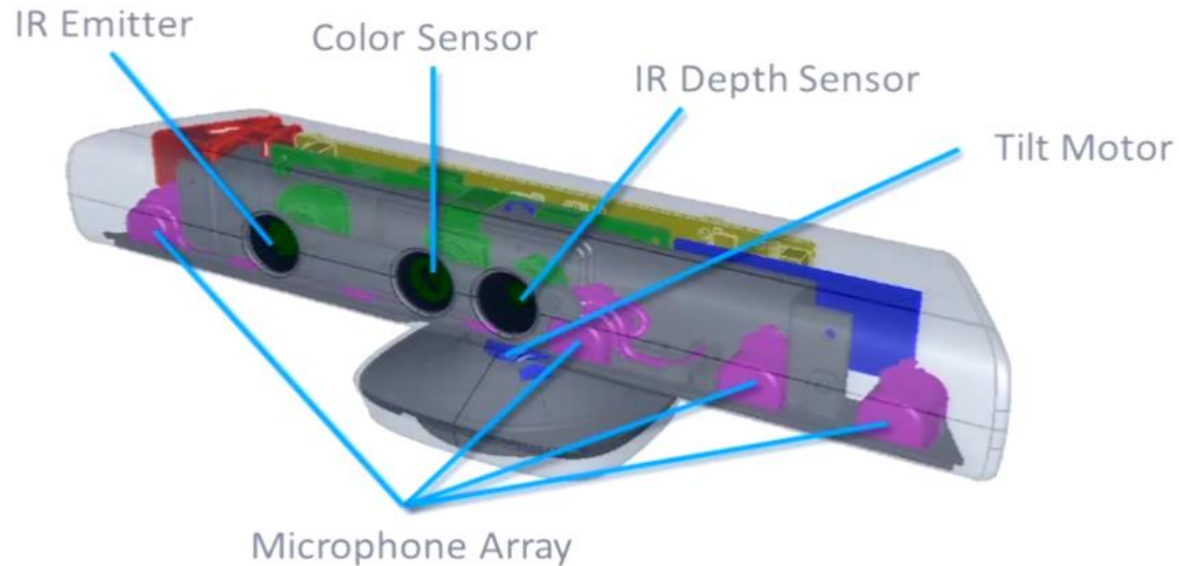


$$disparity = x - x' = \frac{B \cdot f}{z} \quad z = \frac{B \cdot f}{x - x'}$$



Depth from Structured Light

- projector + camera
 - Microsoft Kinect
 - Texture in the environment: generally not important
 - But cannot be used under direct sunlight





Other RGBD Sensor Alternatives

- Intel RealSense
- Asus Xtion Pro
- Microsoft Kinect V2
- Structure Sensor
- If you do not have RGBD sensor
 - <http://www.michaelfirman.co.uk/RGBDdatasets/>





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Projective Geometry II (cfeng@nyu.edu)

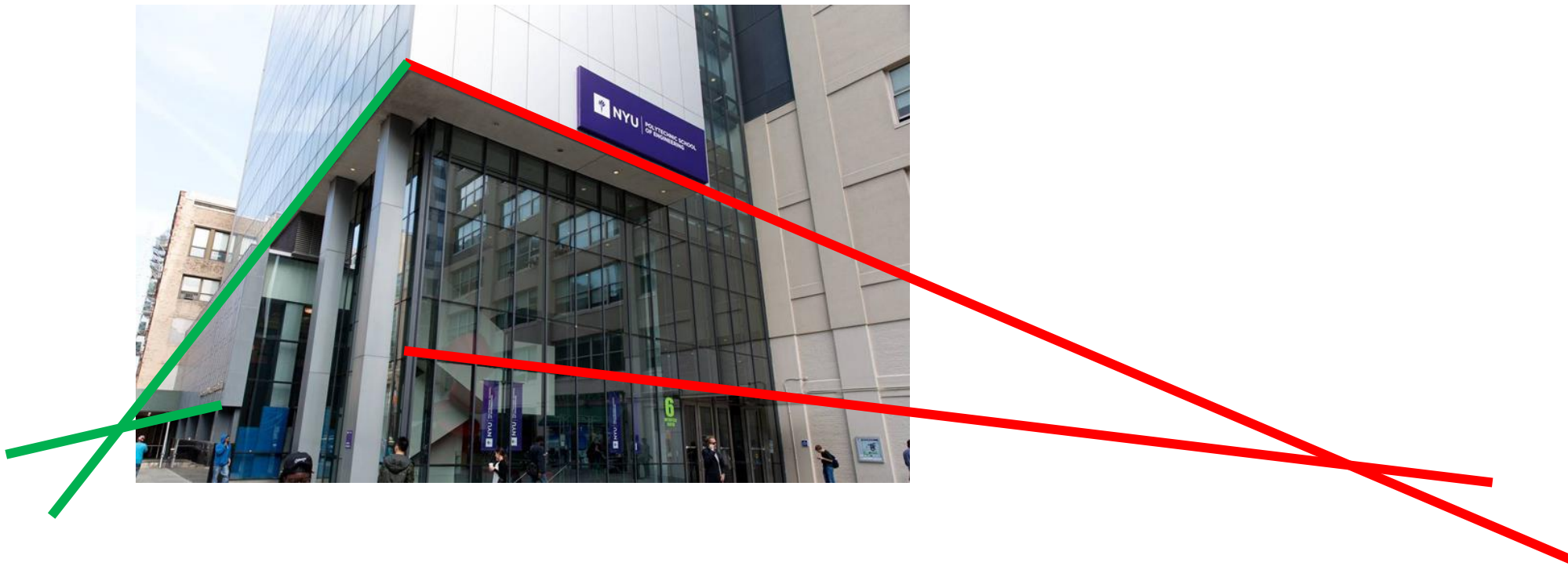


Vanishing Point



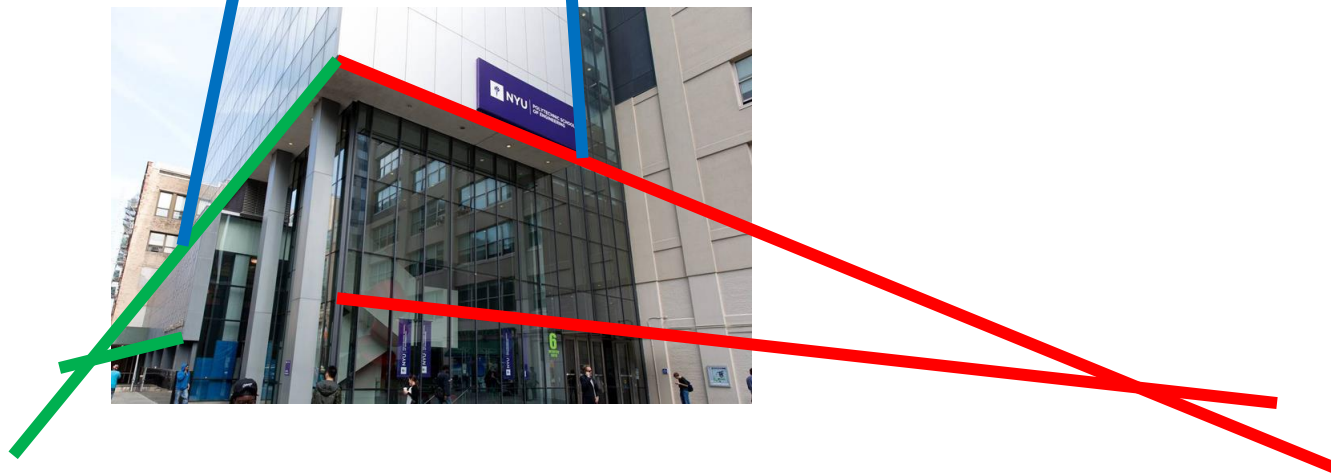


Vanishing Points



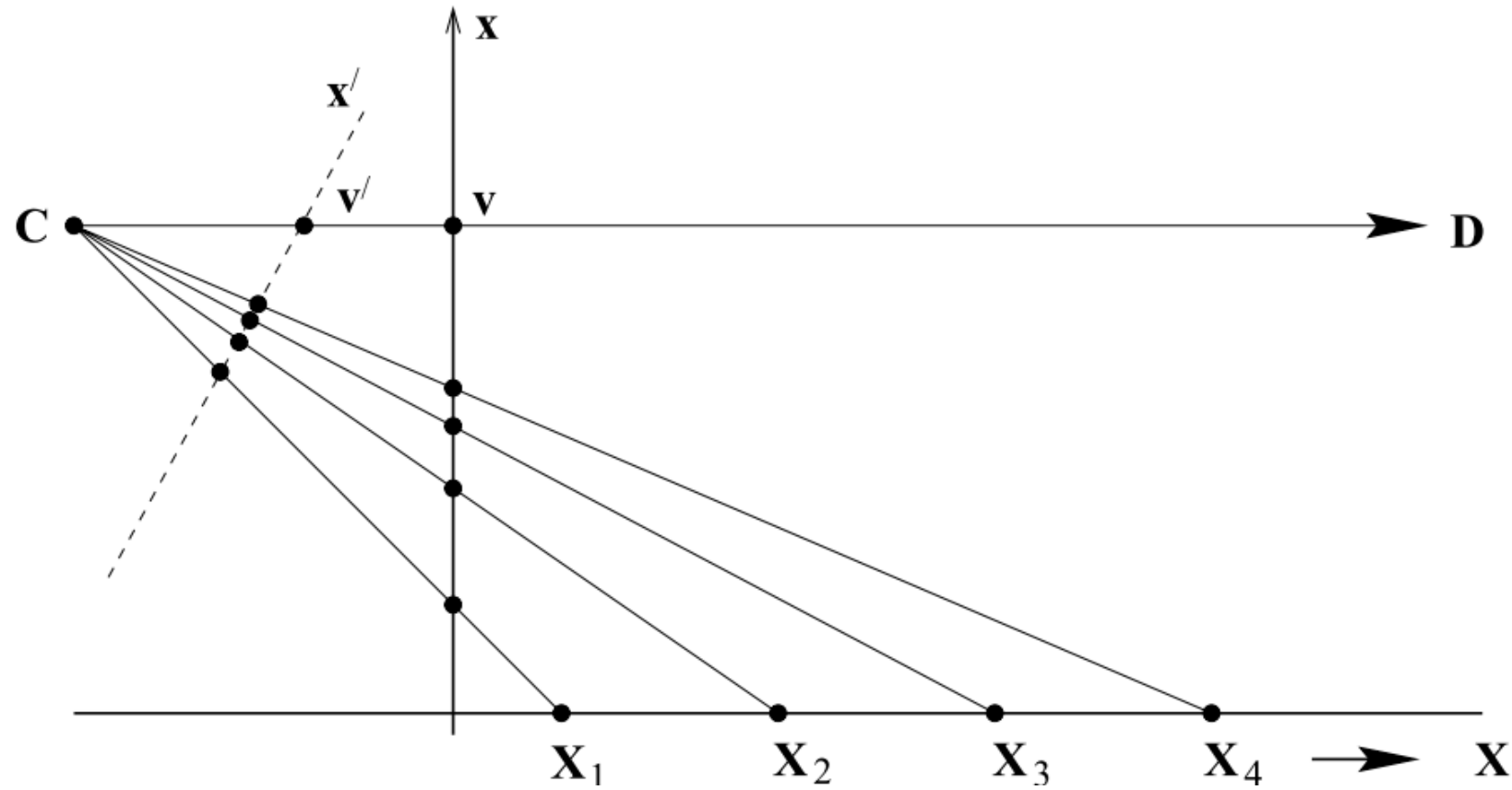


Vanishing Points





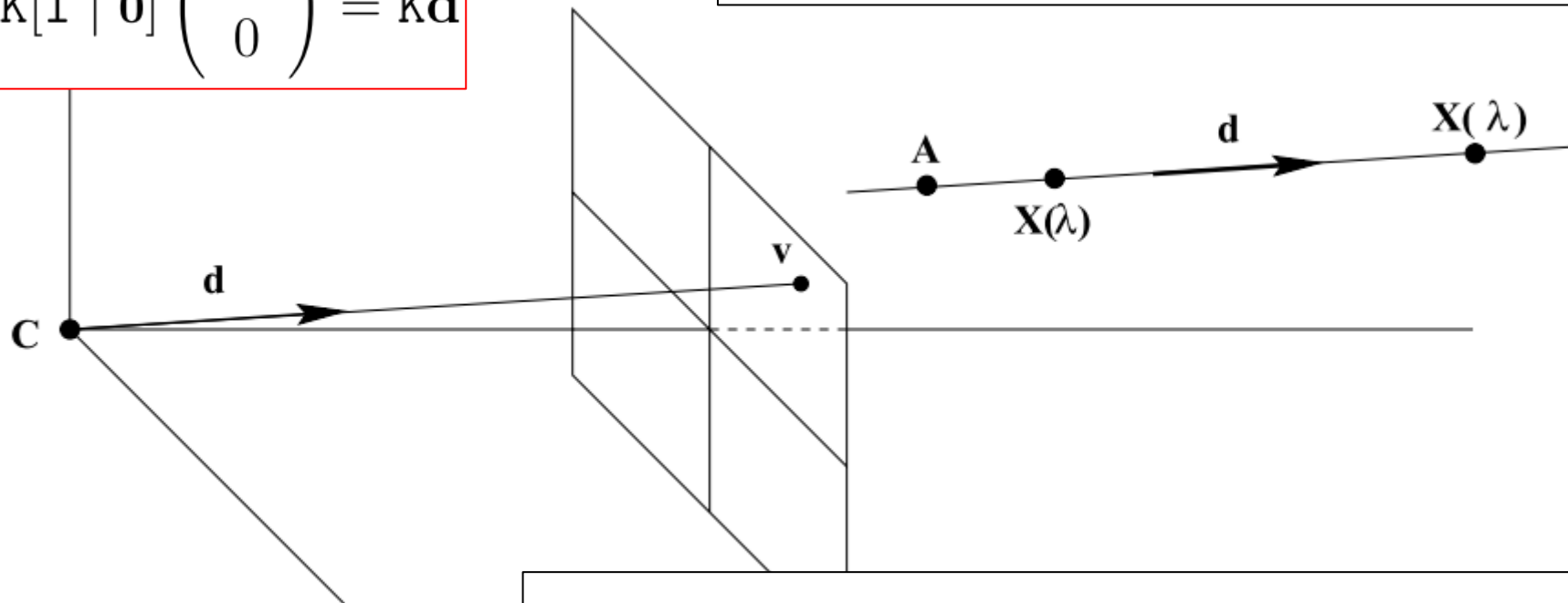
Vanishing Point – 2D





Vanishing Point – 3D

$$\mathbf{v} = P\mathbf{X}_{\infty} = K[\mathbf{I} \mid \mathbf{0}] \begin{pmatrix} \mathbf{d} \\ 0 \end{pmatrix} = K\mathbf{d}$$

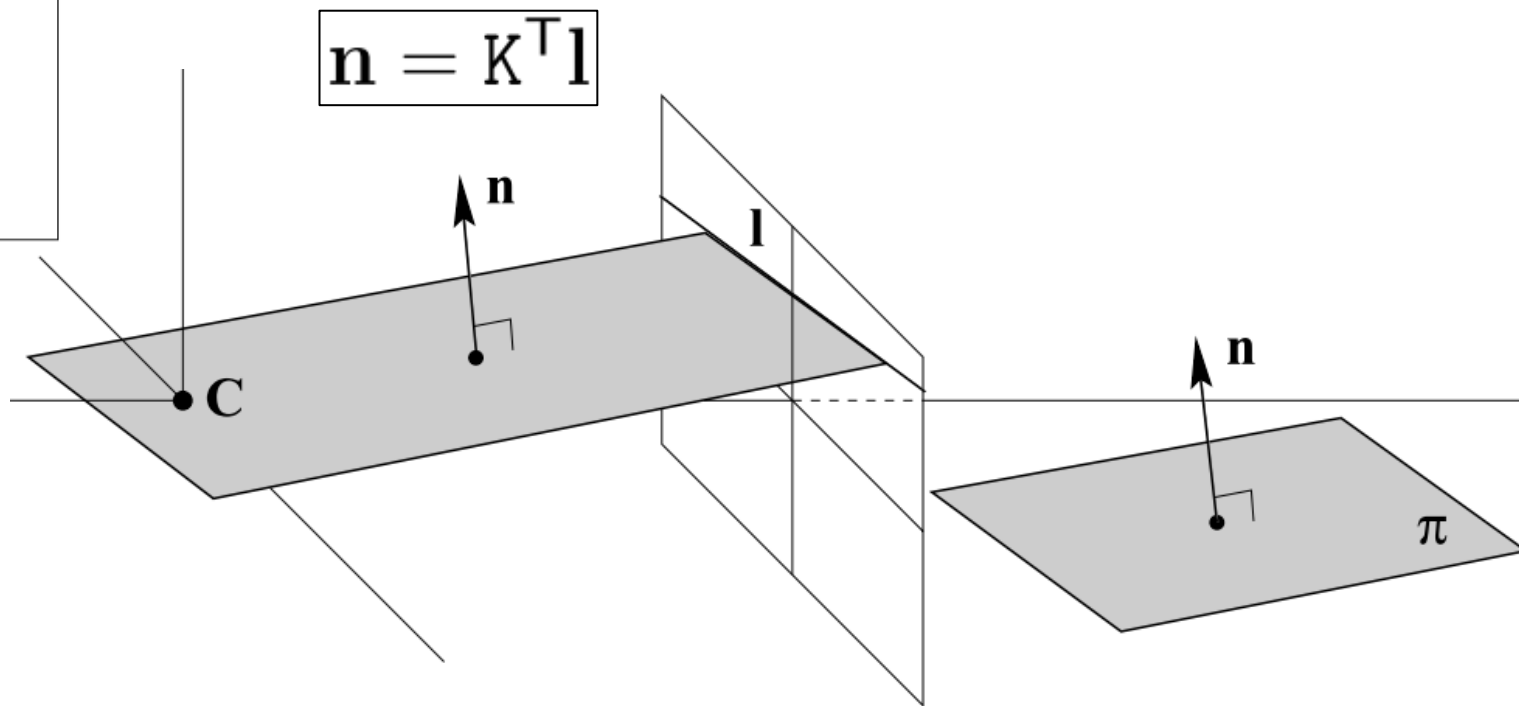
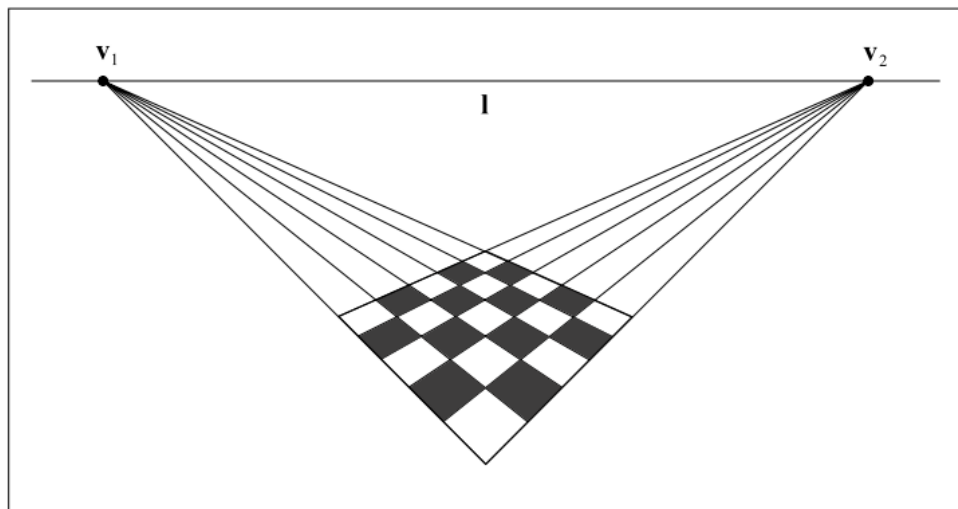


$$\mathbf{x}(\lambda) = P\mathbf{X}(\lambda) = P\mathbf{A} + \lambda P\mathbf{D} = \mathbf{a} + \lambda K\mathbf{d}$$

$$\mathbf{v} = \lim_{\lambda \rightarrow \infty} \mathbf{x}(\lambda) = \lim_{\lambda \rightarrow \infty} (\mathbf{a} + \lambda K\mathbf{d}) = K\mathbf{d}$$



Vanishing Line



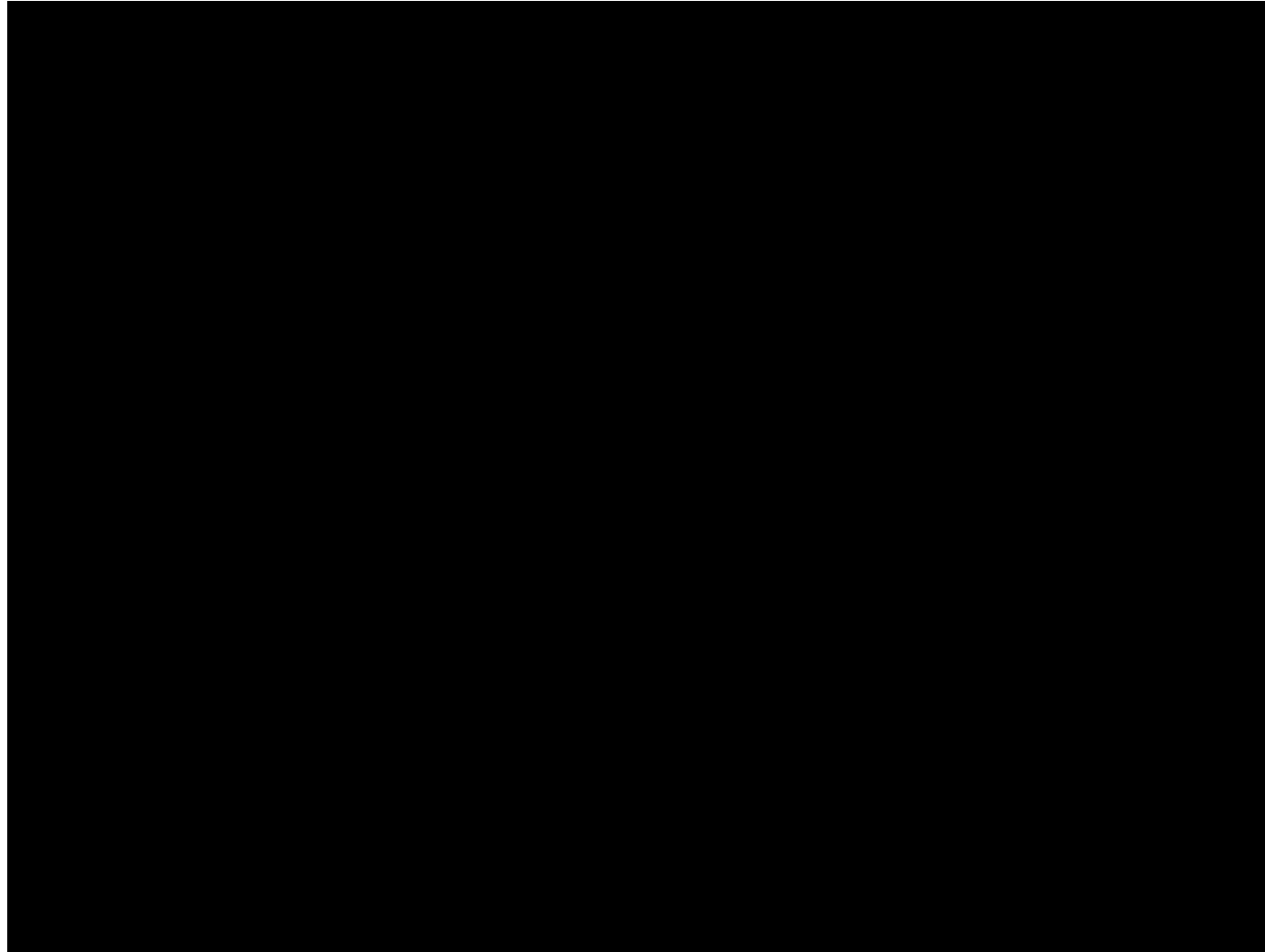


Vanishing Point for Robots





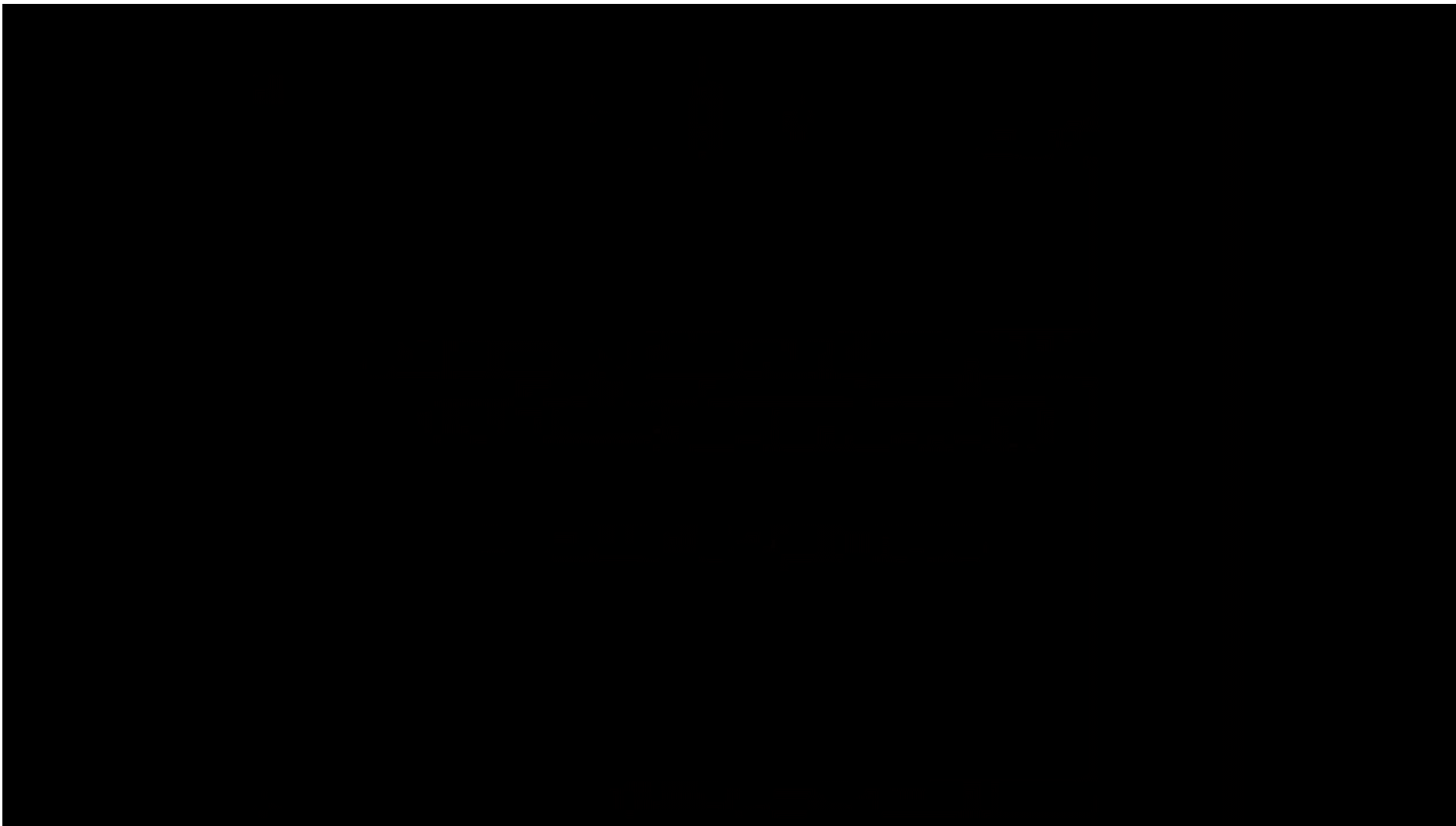
Vanishing Point for Autonomous Driving



Video from: <https://youtu.be/lws6-q9ji0o>



Vanishing Point for Traffic Surveillance



Video from: <https://youtu.be/S3msCdn3fNM>



References for Next Week

- Forsyth & Ponce 2011
 - Chapter 8
- Szeliski 2011:
 - Chapter 11
- Corke 2022:
 - Section 14.4
- Hartley & Zisserman 2003:
 - Section 5.2, 18.1, A6
- Chen Feng, Vineet R. Kamat, and Carol C. Menassa. "Marker-Assisted Structure from Motion for 3D Environment Modeling and Object Pose Estimation." (2016).