

An Invitation to 3D Vision: Two-View Geometry

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Review) Absolute Camera Pose Estimation

Example) Pose estimation (book) + camera calibration - initially given K [pose_estimation_book3.py]
(due to wrong initial K)

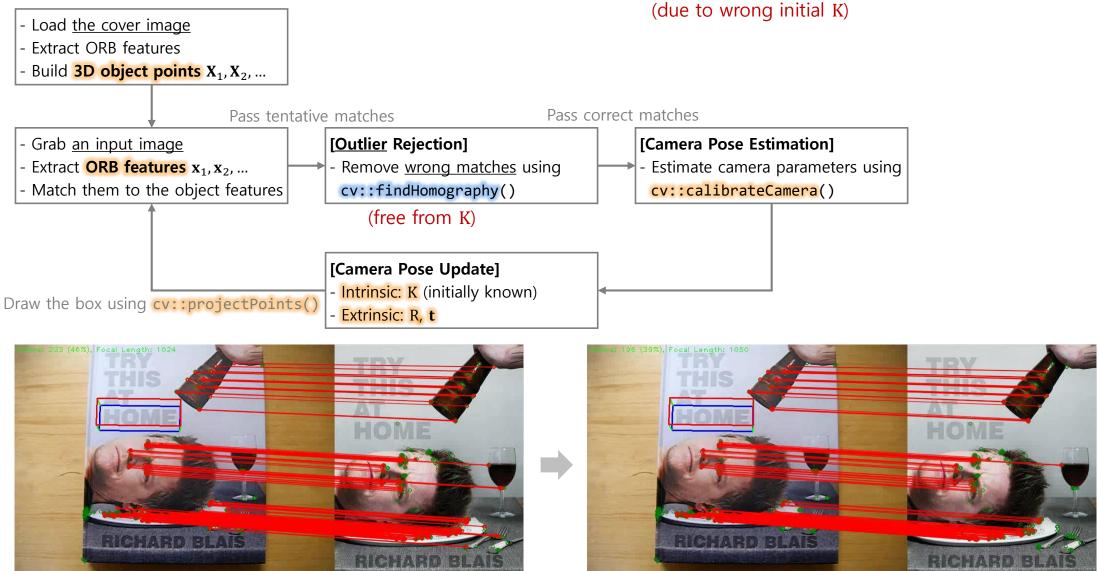
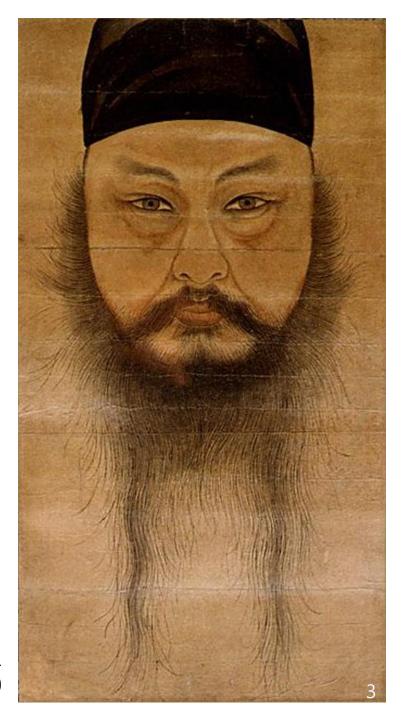
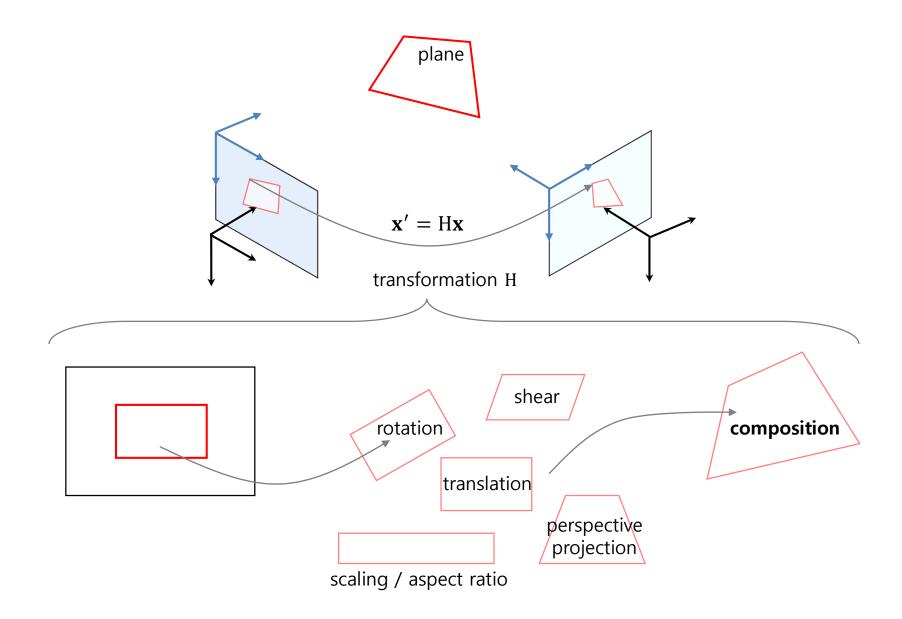


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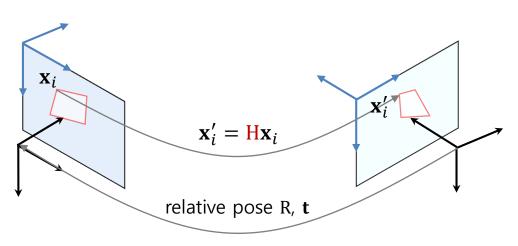
윤두서(1668-1715) 자화상, 국보 제240호 Korean National Treasure No. 240



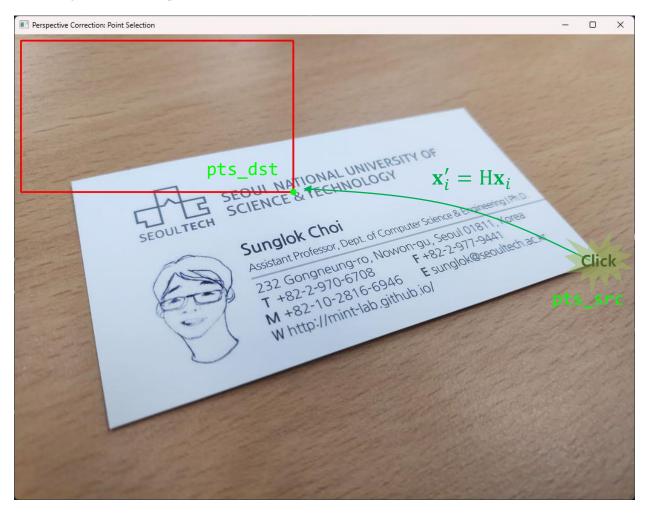
| | Euclidean Transform (a.k.a. Rigid Transform) | Similarity Transform | Affine Transform | Projective Transform (a.k.a. Planar Homography) |
|--|--|--|---|---|
| Matrix Forms H | $\begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{bmatrix}$ | $\begin{bmatrix} s\cos\theta & -s\sin\theta & t_x \\ s\sin\theta & s\cos\theta & t_y \\ 0 & 0 & 1 \end{bmatrix}$ | $\begin{bmatrix} a_{11} & a_{12} & t_x \\ a_{21} & a_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$ | $\begin{bmatrix} a_{11} & a_{12} & t_x \\ a_{21} & a_{22} & t_y \\ v_1 & v_2 & 1 \end{bmatrix}$ |
| DOF | 3 | 4 | 6 | 8 |
| Transformations - rotation - translation - scaling - aspect ratio - shear - perspective projection | O O X X X X | O O O X X X | O O O O X | 0 0 0 0 0 0 |
| Invariants - length - angle - ratio of lengths - parallelism - incidence - cross ratio | 0 0 0 0 0 | X O O O O | X X X O O | X X X O O |
| OpenCV Functions | | | <pre>cv::getAffineTransform() cv::estimateRigidTransform() - cv::warpAffine()</pre> | <pre>cv::getPerspectiveTransform() - cv::findHomography() cv::warpPerspective()</pre> |

Planar homography estimation

- Unknown: Planar homography H (8 DOF)
- Given: Point correspondence $(\mathbf{x}_1, \mathbf{x}_1'), ..., (\mathbf{x}_n, \mathbf{x}_n')$
- Constraints: $n \times projective transformation \mathbf{x}'_i = H\mathbf{x}_i$
- Solutions $(n \ge 4) \rightarrow 4$ -point algorithm
 - OpenCV: cv.getPerspectiveTransform() and cv.findHomography()
 - Note) More simplified transformations need less number of minimal correspondence.
 - Affine $(n \ge 3)$, similarity $(n \ge 2)$, Euclidean $(n \ge 2)$
- Note) Planar homography can be decomposed as relative camera pose.
 - OpenCV: cv.decomposeHomographyMat()
 - The decomposition needs to know camera matrices.



Example) Perspective distortion correction [perspective_correction.py]





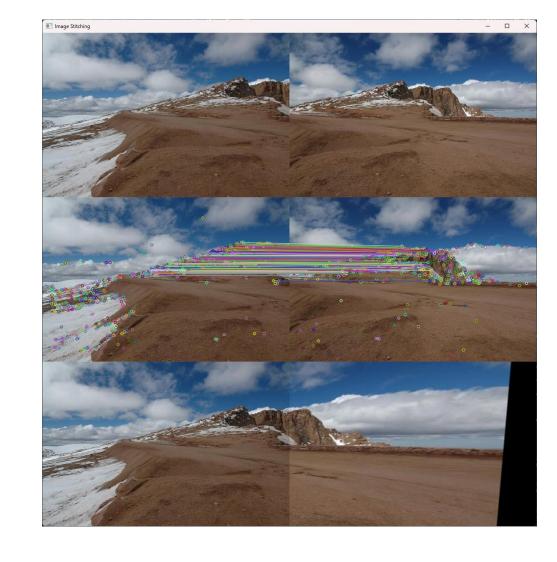
Example) **Perspective distortion correction** [perspective_correction.py] def mouse_event_handler(event, x, y, flags, param): if event == cv.EVENT_LBUTTONDOWN: param.append((x, y)) if name__ == '__main__': img file = '.../data/sunglok card.jpg' card size = (450, 250)offset = 10# Prepare the rectified points $pts_dst = np.array([[0, 0], [card_size[0], 0], [0, card_size[1]], [card_size[0], card_size[1]]])$ # Load an image img = cv.imread(img file) # Get the matched points from mouse clicks pts src = [] wnd name = 'Perspective Correction: Point Selection' cv.namedWindow(wnd_name) cv.setMouseCallback(wnd_name, mouse event_handler, pts_src) while len(pts src) < 4:</pre> img_display = img.copy() cv.rectangle(img_display, (offset, offset), (offset + card_size[0], offset + card_size[1]), (0, 0, 255), 2) idx = min(len(pts_src), len(pts_dst)) cv.circle(img_display, offset + pts_dst[idx], 5, (0, 255, 0), -1) 8 cv.imshow(wnd_name, img_display)

Example) Perspective distortion correction [perspective_correction.py]

```
if name == ' main ':
    img_file = '../data/sunglok_card.jpg'
    card size = (450, 250)
    offset = 10
   # Prepare the rectified points
    pts_dst = np.array([[0, 0], [card_size[0], 0], [0, card_size[1]], [card_size[0], card_size[1]]])
   # Load an image
    img = cv.imread(img file)
   # Get the matched points from mouse clicks
   pts src = []
    if len(pts src) == 4:
        # Calculate planar homography and rectify perspective distortion
        H, _ = cv.findHomography(np.array(pts_src), pts_dst)
       img_rectify = cv.warpPerspective(img, H, card_size)
       # Show the rectified image
        cv.imshow('Perspective Correction: Rectified Image', img rectify)
        cv.waitKey(0)
    cv.destroyAllWindows()
```

Example) Planar image stitching [image_stitching.py]

```
# Load two images
img1 = cv.imread('../data/hill01.jpg')
img2 = cv.imread('../data/hill02.jpg')
# Retrieve matching points
brisk = cv.BRISK create()
keypoints1, descriptors1 = brisk.detectAndCompute(img1, None)
keypoints2, descriptors2 = brisk.detectAndCompute(img2, None)
fmatcher = cv.DescriptorMatcher create('BruteForce-Hamming')
match = fmatcher.match(descriptors1, descriptors2)
# Calculate planar homography and merge them
pts1, pts2 = [], []
for i in range(len(match)):
    pts1.append(keypoints1[match[i].queryIdx].pt)
    pts2.append(keypoints2[match[i].trainIdx].pt)
pts1 = np.array(pts1, dtype=np.float32)
pts2 = np.array(pts2, dtype=np.float32)
```



```
Example) 2D video stabilization [video_stabilization.py]
  # Open a video and get the reference image and feature points
  video = cv.VideoCapture('../data/traffic.avi')
  _, gray_ref = video.read()
  if gray ref.ndim >= 3:
      gray_ref = cv.cvtColor(gray_ref, cv.COLOR_BGR2GRAY)
  pts_ref = cv.goodFeaturesToTrack(gray_ref, 2000, 0.01, 10)
  # Run and show video stabilization
                                                                     A shaking CCTV video
  while True:
      # Read an image from `video`
      valid, img = video.read()
      if not valid:
          break
      if img.ndim >= 3:
          gray = cv.cvtColor(img, cv.COLOR BGR2GRAY)
      else:
          gray = img.copy()
      # Extract optical flow and calculate planar homography
      pts, status, err = cv.calcOpticalFlowPyrLK(gray_ref, gray, pts_ref, None)
      H, inlier_mask = <a href="mask">cv.findHomography</a>(pts, pts_ref, cv.RANSAC)
      # Synthesize a stabilized image
      warp = cv.warpPerspective(img, H, (img.shape[1], img.shape[0]))
```

- Assumption) A plane is observed by two views.
 - Perspective distortion correction: A complete plane
 - Planar image stitching: An approximated plane (← distance ≫ depth variation)
 - 2D video stabilization: An approximated plane (← small motion)











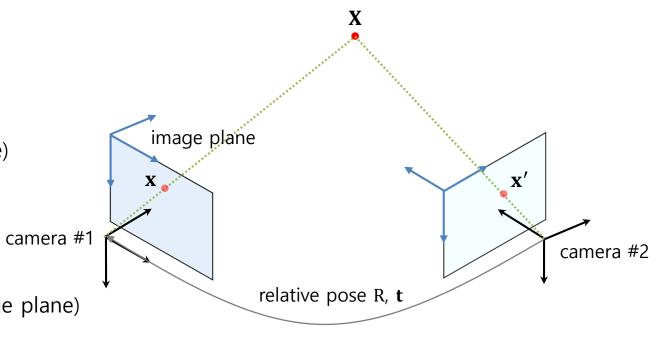
Epipolar constraint

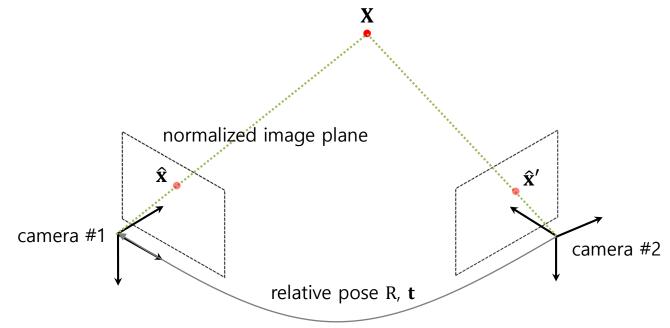
 $\mathbf{x'}^{\mathsf{T}}\mathbf{F}\mathbf{x} = \mathbf{0}$ (F: <u>Fundamental matrix</u> on the image plane) $(\mathbf{x} = \mathbf{K}\hat{\mathbf{x}})$

 $\hat{\mathbf{x}}^{\prime \mathsf{T}} \mathbf{K}^{\prime \mathsf{T}} \mathbf{F} \mathbf{K} \hat{\mathbf{x}} = \mathbf{0}$

 $\updownarrow \quad (E = K'^{\mathsf{T}}FK)$

 $\hat{\mathbf{x}}'^{\mathsf{T}} \mathbf{E} \hat{\mathbf{x}} = \mathbf{0}$ (E: <u>Essential matrix</u> on the normalized image plane)





Epipolar constraint: Derivation

$$\mathbf{X}' = \mathbf{R}\mathbf{X} + \mathbf{t}$$

$$\downarrow \quad (\mathbf{X} = \lambda \hat{\mathbf{x}})$$

$$\lambda' \hat{\mathbf{x}}' = \lambda \mathbf{R} \hat{\mathbf{x}} + \mathbf{t}$$

$$\mathbf{t} \times \downarrow$$

$$\lambda' \mathbf{t} \times \hat{\mathbf{x}}' = \lambda \mathbf{t} \times \mathbf{R} \hat{\mathbf{x}}$$

$$\hat{\mathbf{x}}' \cdot \downarrow$$

$$\lambda' \hat{\mathbf{x}}' \cdot (\mathbf{t} \times \hat{\mathbf{x}}') = \lambda \hat{\mathbf{x}}' \cdot \mathbf{t} \times \mathbf{R} \hat{\mathbf{x}}$$

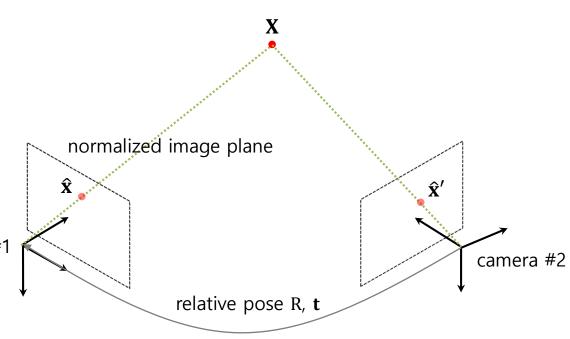
$$\downarrow \quad (\hat{\mathbf{x}}' \cdot (\mathbf{t} \times \hat{\mathbf{x}}') = 0)$$

$$\lambda \hat{\mathbf{x}}' \cdot \mathbf{t} \times \mathbf{R} \hat{\mathbf{x}} = 0$$

$$\downarrow \quad (\mathbf{E} = \mathbf{t} \times \mathbf{R} \text{ or } \mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}) \quad \text{Note)} \quad [\mathbf{t}]_{\times} = \begin{bmatrix} 0 & -t_{z} & t_{y} \\ t_{z} & 0 & -t_{x} \\ -t_{y} & t_{x} & 0 \end{bmatrix}$$

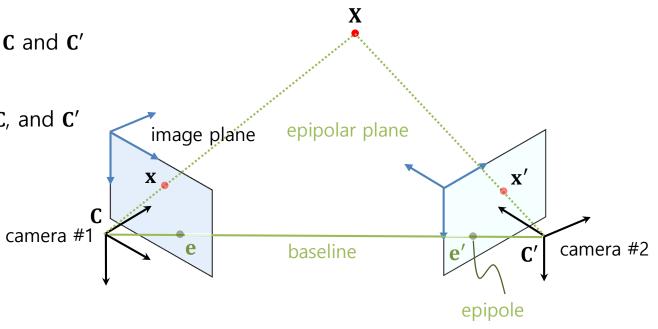
$$\hat{\mathbf{x}}' \mathbf{E} \hat{\mathbf{x}} = 0$$

$$\text{camera #1}$$



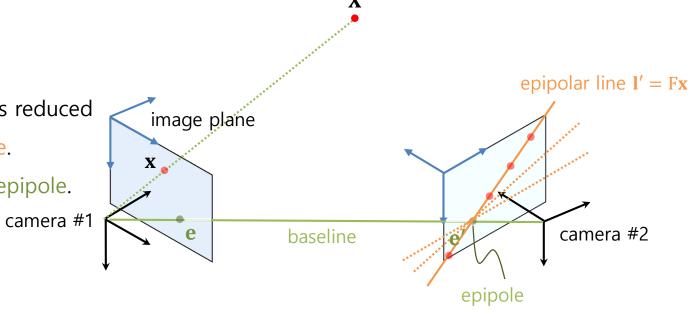
Epipolar geometry

- Baseline
 - Distance between two camera centers, C and C'
- Epipolar plane
 - Plane generated from three points, X, C, and C'
- Epipole (a.k.a. epipolar point)
 - Projection of other camera centers
 - $\mathbf{e} = P\mathbf{C}'$ and $\mathbf{e}' = P'\mathbf{C}$
 - e.g. P = K[I|0] and P' = K'[R|t]
 - $-\mathbf{e} = -\mathbf{K}\mathbf{R}^{\mathsf{T}}\mathbf{t}$ and $\mathbf{e}' = \mathbf{K}'\mathbf{t}$

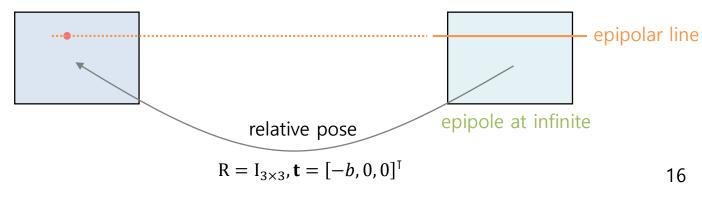


Epipolar geometry

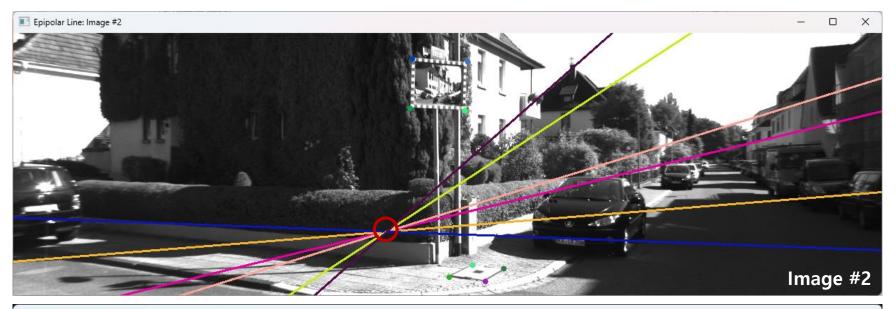
- Epipolar line
 - $\mathbf{x'}^{\mathsf{T}} \mathbf{F} \mathbf{x} = 0 \rightarrow \mathbf{x'}^{\mathsf{T}} \mathbf{l'} = 0$ where $\mathbf{l'} = \mathbf{F} \mathbf{x}$
 - x' will lie on the line 1'.
 - The search space of feature matching is reduced from a image plane to the epipolar line.
- Note) Every epipolar line intersects at the epipole.
 - Fe = 0
 - e is the null space of F.



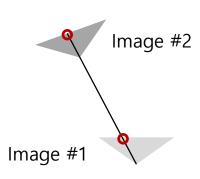
Special case) Stereo cameras



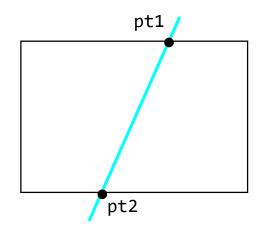
• Example) Epipolar line visualization [epipolar_line_visualization.py]







Example) **Epipolar line visualization** [epipolar_line_visualization.py] def mouse_event_handler(event, x, y, flags, param): if event == cv.EVENT_LBUTTONDOWN: param.append((x, y)) def draw straight line(img, line, color, thickness=1): h, w, * = img.shapea, b, c = line # Line: ax + by + c = 0if abs(a) > abs(b): pt1 = (int(c / -a), 0)pt2 = (int((b*h + c) / -a), h)else: cv.line(img, pt1, pt2, color, thickness) if name__ == '__main__': # Load two images img1 = cv.imread('../data/KITTI07/image 0/000000.png', cv.IMREAD COLOR) img2 = cv.imread('../data/KITTI07/image 0/000023.png', cv.IMREAD COLOR) # Note) `F` is derived from `fundamental mat estimation.py`. F = np.array([[3.34638533e-07, 7.58547151e-06, -2.04147752e-03], ...])# Register event handlers and show images wnd1_name, wnd2_name = 'Epipolar Line: Image #1', 'Epipolar Line: Image #2' img1_pts, img2_pts = [], [] cv.namedWindow(wnd1_name) cv.setMouseCallback(wnd1_name, mouse_event_handler, img1_pts)

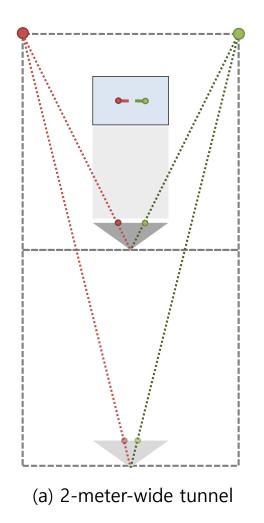


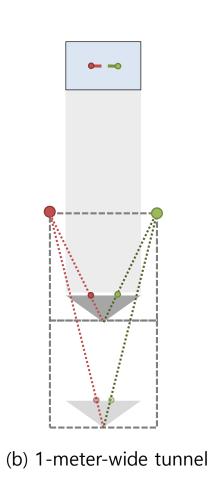
Example) **Epipolar line visualization** [epipolar_line_visualization.py] if __name__ == '__main__': # Load two images img1 = cv.imread('../data/KITTI07/image_0/000000.png', cv.IMREAD COLOR) img2 = cv.imread('../data/KITTI07/image 0/000023.png', cv.IMREAD COLOR) F = np.arrav([[3.34638533e-07, 7.58547151e-06, -2.04147752e-03], ...])# Register event handlers and show images # Get a point from a image and draw its correponding epipolar line on the other image while True: if len(img1 pts) > 0: for x, y in img1 pts: color = (random.randrange(256), random.randrange(256), random.randrange(256)) cv.circle(img1, (x, y), 4, color, -1) $\mathbf{I'} = \mathbf{F}\mathbf{x}$ epipolar_line = F @ [[x], [y], [1]] draw_straight_line(img2, epipolar line, color, 2) img1 pts.clear() if len(img2_pts) > 0: for x, y in img2_pts: color = (random.randrange(256), random.randrange(256), random.randrange(256)) cv.circle(img2, (x, y), 4, color, -1)epipolar_line = F.T @ [[x], [y], [1]] draw_straight_line($\frac{img1}{img1}$, $\frac{epipolar_line}{img1}$, color, 2) $I = F^T x'$ img2_pts.clear() cv.imshow(wnd2 name, img2)

- Relative camera pose estimation (~ fundamental/essential matrix estimation)
 - Unknown: Rotation and translation R, t (5 DOF; up-to scale "scale ambiguity")
 - Given: Point correspondence $(\mathbf{x}_1, \mathbf{x}_1')$, ..., $(\mathbf{x}_n, \mathbf{x}_n')$ and camera matrices K, K'
 - Constraints: $n \times \text{epipolar constraint } (\mathbf{x'}^\mathsf{T} \mathbf{F} \mathbf{x} = 0 \text{ or } \hat{\mathbf{x}}'^\mathsf{T} \mathbf{E} \hat{\mathbf{x}} = 0)$
 - Solutions) Fundamental matrix: 7/8-point algorithm (7 DOF; intrinsic+extrinsic)
 - **Properties**: det(F) = 0 (or rank(F) = 2)
 - **Estimation**: cv.findFundamentalMat() → 1 solution
 - Conversion to E: $E = K'^{\mathsf{T}}FK$
 - Solutions) Essential matrix: 5-point algorithm (5 DOF; extrinsic)
 - **Properties**: det(E) = 0 and $2EE^{T}E tr(EE^{T})E = 0$
 - **Estimation**: $cv.findEssentialMat() \rightarrow k$ solutions
 - **Decomposition**: cv.decomposeEssentialMat() → 4 solutions "relative pose ambiguity"
 - **Decomposition with positive-depth check**: cv.recoverPose() → 1 solution

```
R, t
\updownarrow (E = [t]_{\times}R)
E
\updownarrow (E = K'^{\mathsf{T}}FK)
F
```

Relative Camera Pose Estimation: Scale Ambiguity







The Sandcrawler @ Star Wars IV: A New Hope (1977)

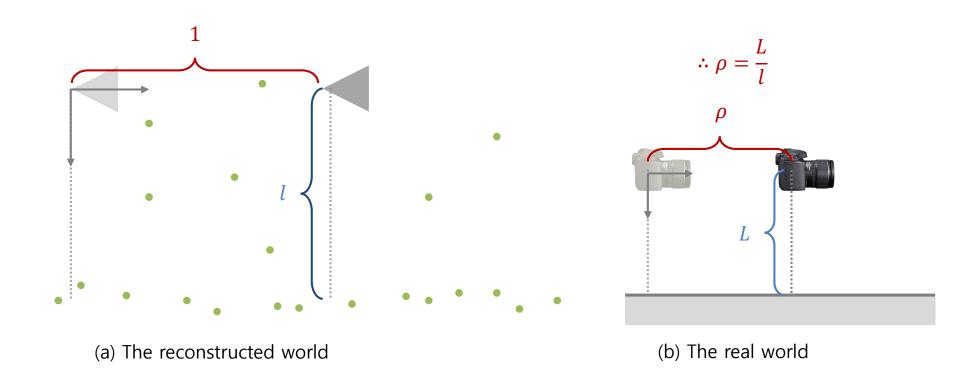




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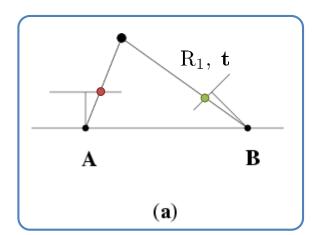
Relative Camera Pose Estimation: Scale Ambiguity

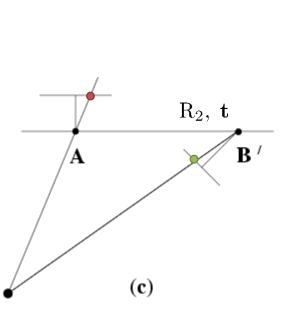
- How to resolve scale ambiguity
 - Additional sensors: Speedometers (odometers), IMUs, GPSs, depth/distance (stereo, RGB-D, LiDAR, ...)
 - Motion constraints: Known initial translation, Ackerman's steering kinematics
 - Observation constraints: Known size of objects, known and constant height of camera

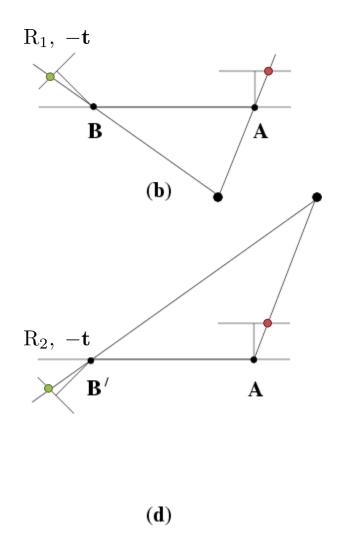


Relative Camera Pose Estimation: Relative Pose Ambiguity

- How to resolve pose ambiguity
 - Positive depth constraint





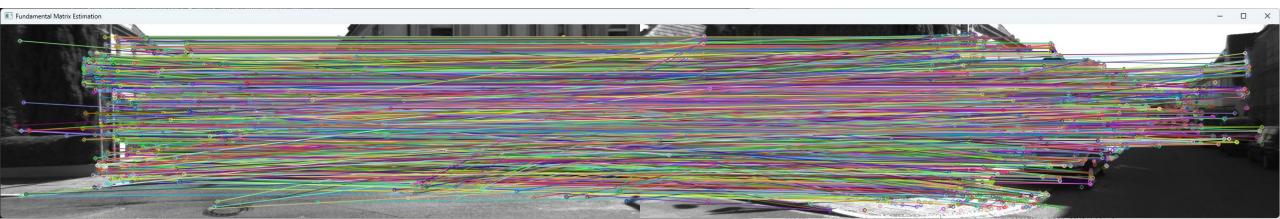


Example) **Fundamental matrix estimation** [fundamental_mat_estimation.py]

```
# Load two images
img1 = cv.imread('../data/KITTI07/image_0/000000.png')
img2 = cv.imread('../data/KITTI07/image_0/000023.png')
f, cx, cy = 707.0912, 601.8873, 183.1104 # From the KITTI dataset
K = np.array([[f, 0, cx], [0, f, cy], [0, 0, 1]])

# Retrieve matching points
brisk = cv.BRISK_create()
keypoints1, descriptors1 = brisk.detectAndCompute(img1, None)
keypoints2, descriptors2 = brisk.detectAndCompute(img2, None)

fmatcher = cv.DescriptorMatcher_create('BruteForce-Hamming')
match = fmatcher.match(descriptors1, descriptors2)
```

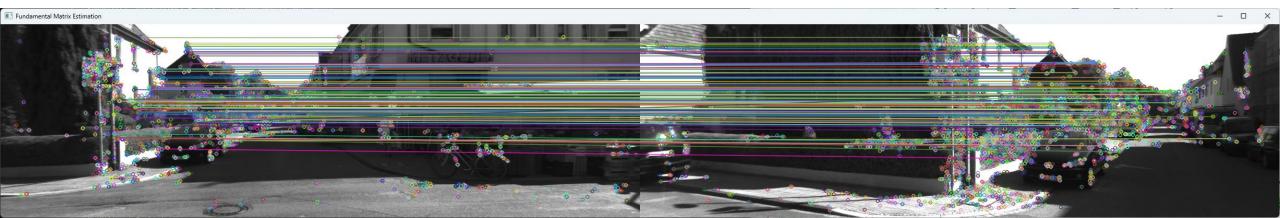


Example) **Fundamental matrix estimation** [fundamental_mat_estimation.py]

```
# Load two images
...

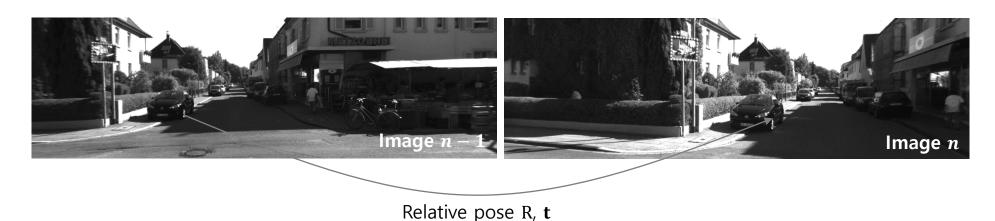
# Retrieve matching points
...

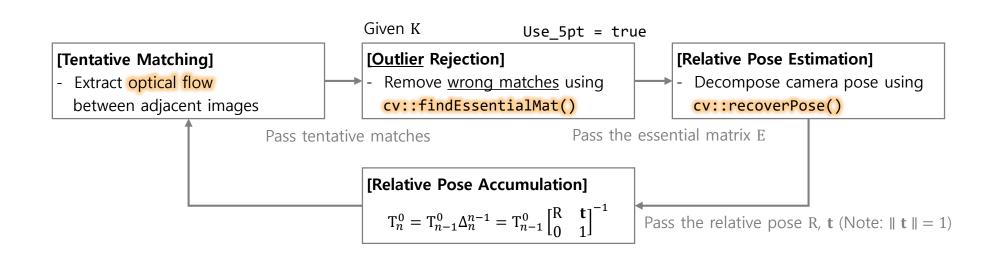
# Calculate the fundamental matrix
pts1, pts2 = [], []
for i in range(len(match)):
    pts1.append(keypoints1[match[i].queryIdx].pt)
    pts2.append(keypoints2[match[i].trainIdx].pt)
pts1 = np.array(pts1, dtype=np.float32)
pts2 = np.array(pts2, dtype=np.float32)
F, inlier_mask = cv.findFundamentalMat(pts1, pts2, cv.FM_RANSAC, 0.5, 0.999)
```



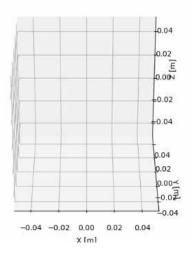
```
Example) Fundamental matrix estimation [fundamental_mat_estimation.py]
    # Load two images
    f, cx, cy = 707.0912, 601.8873, 183.1104 # From the KITTI dataset
    K = np.array([[f, 0, cx], [0, f, cy], [0, 0, 1]])
    # Retrieve matching points
    # Calculate the fundamental matrix
    F, inlier mask = <a href="mask">cv.findFundamentalMat</a>(pts1, pts2, cv.FM RANSAC, 0.5, 0.999)
                                                                                                           Image #2
    # Extract relative camera pose between two images
                                                                                                           [-0.57, 0.09, 0.82]
    E = K.T @ F @ K
    positive_num, R, t, positive_mask = cv.recoverPose(E, pts1, pts2, K, mask=inlier_mask)
    print(f'* The position of Image #2 = \{-R.T @ t\}') # [-0.57, 0.09, 0.82]
                                                                                                Image #1
```

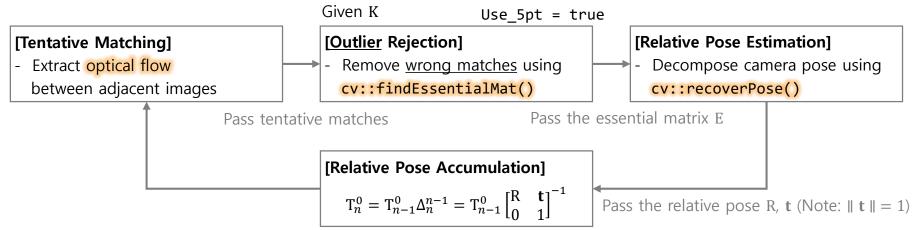
[0, 0, 0]











```
video file = '../data/KITTI07/image 0/%06d.png'
f, cx, cy = 707.0912, 601.8873, 183.1104
use 5pt = True
min inlier num = 100
min inlier ratio = 0.2
traj file = '../data/vo epipolar.xyz'
# Open a video and get an initial image
video = cv.VideoCapture(video file)
  gray_prev = video.read()
if gray_prev.ndim >= 3 and gray_prev.shape[2] > 1:
    gray prev = cv.cvtColor(gray prev, cv.COLOR BGR2GRAY)
# Prepare a plot to visualize the camera trajectory
# Run the monocular visual odometry
K = np.array([[f, 0, cx], [0, f, cy], [0, 0, 1]])
camera_traj = np.zeros((1, 3))
camera pose = np.eye(4)
while True:
    # Grab an image from the video
    valid, img = video.read()
    if not valid:
        break
```

```
# Run the monocular visual odometry
K = np.array([[f, 0, cx], [0, f, cy], [0, 0, 1]])
camera_traj = np.zeros((1, 3))
camera pose = np.eye(4)
while True:
    # Grab an image from the video
    valid, img = video.read()
    if not valid:
        break
    if img.ndim >= 3 and img.shape[2] > 1:
        gray = cv.cvtColor(img, cv.COLOR BGR2GRAY)
    else:
        gray = img.copy()
    # Extract optical flow
    pts prev = cv.goodFeaturesToTrack(gray prev, 2000, 0.01, 10)
    pts, status, err = cv.calcOpticalFlowPyrLK(gray prev, gray, pts prev, None)
    gray prev = gray
    # Calculate relative pose
    if use 5pt:
        E, inlier_mask = cv.findEssentialMat(pts_prev, pts, f, (cx, cy), cv.FM_RANSAC, 0.99, 1)
    else:
        F, inlier mask = cv.findFundamentalMat(pts_prev, pts, cv.FM_RANSAC, 1, 0.99)
        E = K.T @ F @ K
    inlier_num, R, t, inlier_mask = cv.recoverPose(E, pts_prev, pts, focal=f, pp=(cx, cy), mask=inlier_mask)
```

```
# Run the monocular visual odometry
while True:
    # Grab an image from the video
    . . .
    # Extract optical flow
    . . .
    # Calculate relative pose
    if use 5pt:
         E, inlier mask = cv.findEssentialMat(pts prev, pts, f, (cx, cy), cv.FM RANSAC, 0.99, 1)
    else:
         F, inlier_mask = <a href="mask">cv.findFundamentalMat</a>(pts_prev, pts, cv.FM_RANSAC, 1, 0.99)
         E = K.T @ F @ K
    inlier_num, R, t, inlier_mask = cv.recoverPose(E, pts_prev, pts, focal=f, pp=(cx, cy), mask=inlier_mask)
    inlier ratio = inlier num / len(pts)
    # Accumulate relative pose if result is reliable
    info color = (0, 255, 0)
    if inlier_num > min_inlier_num and inlier_ratio > min_inlier_ratio:
         T = np.eye(4)
        T[:3, :3] = R
         T[:3, 3] = t.flatten()
                                                                T_n^0 = T_{n-1}^0 \Delta_n^{n-1} = T_{n-1}^0 \begin{bmatrix} R & \mathbf{t} \\ 0 & 1 \end{bmatrix}^{-1}
         camera_pose = camera_pose @ np.linalg.inv(T)
         info color = (0, 0, 255)
```

- Relative camera pose estimation (~ fundamental/essential matrix estimation)
 - Unknown: Rotation and translation R, t (5 DOF; up-to scale "scale ambiguity")
 - Given: Point correspondence $(\mathbf{x}_1, \mathbf{x}_1')$, ..., $(\mathbf{x}_n, \mathbf{x}_n')$ and camera matrices K, K'
 - Constraints: $n \times \text{epipolar constraint } (\mathbf{x}'^{\mathsf{T}} \mathbf{F} \mathbf{x} = 0 \text{ or } \hat{\mathbf{x}}'^{\mathsf{T}} \mathbf{E} \hat{\mathbf{x}} = 0)$
 - Solutions) Fundamental matrix: 7/8-point algorithm (7 DOF; intrinsic+extrinsic)
 - **Properties**: det(F) = 0 (or rank(F) = 2)
 - **Estimation**: cv.findFundamentalMat() → 1 solution
 - Conversion to E: $E = K'^T F K$
 - **Degenerate cases**: No translation, correspondence from a single plane
 - Solutions) Essential matrix: 5-point algorithm (5 DOF; extrinsic)
 - **Properties**: det(E) = 0 and $2EE^{T}E tr(EE^{T})E = 0$
 - **Estimation**: cv.findEssentialMat() $\rightarrow k$ solutions
 - Decomposition: cv.decomposeEssentialMat() → 4 solutions "relative pose ambiguity"
 - **Decomposition with positive-depth check**: cv.recoverPose() → 1 solution
 - Degenerate cases: No translation (: $E = [t]_{\times}R$)

```
R, t

\updownarrow (E = [t]_{\times}R)

E

\updownarrow (E = K'^{\mathsf{T}}FK)

F
```

Relative camera pose estimation

- Solutions) Fundamental matrix: 7/8-point algorithm (7 DOF; intrinsic+extrinsic)
 - **Estimation**: cv.findFundamentalMat() → 1 solution
 - Conversion to E: $E = K'^T F K$
 - **Degenerate cases**: No translation, correspondence from a single plane
- Solutions) Essential matrix: 5-point algorithm (5 DOF; extrinsic)
 - **Estimation**: cv.findEssentialMat() $\rightarrow k$ solutions
 - **Decomposition**: cv.decomposeEssentialMat() → 4 solutions "relative pose ambiguity"
 - **Decomposition with positive-depth check**: cv.recoverPose() → 1 solution
 - Degenerate cases: No translation (: $E = [t]_{\times}R$)
- Solutions) Planar homography: 4-point algorithm (8 DOF; up-to scale "scale ambiguity")
 - **Estimation**: cv.findHomography() → 1 solutions
 - Conversion to calibrated H: $\hat{H} = K'^{-1}HK$
 - **Decomposition**: cv.decomposeHomographyMat() → 4 solutions "relative pose ambiguity"
 - **Degenerate cases**: Correspondence **not** from a single plane

- Relative camera pose estimation
 - Solutions) Planar homography: 4-point algorithm (8 DOF; up-to scale "scale ambiguity")
 - **Estimation**: cv.findHomography() → 1 solutions
 - Derivation

$$\lambda' \hat{\mathbf{x}}' = \lambda R \hat{\mathbf{x}} + \mathbf{t}$$

$$\downarrow \frac{1}{d} \mathbf{n}^{\mathsf{T}} \hat{\mathbf{x}} = 1 \quad (\because n_{x} \hat{\mathbf{x}} + n_{y} \hat{\mathbf{y}} + n_{z} \hat{\mathbf{z}} - d = 0)$$

$$\hat{\mathbf{x}}' = \lambda'' \left(R + \frac{1}{d} \mathbf{t} \mathbf{n}^{\mathsf{T}} \right) \hat{\mathbf{x}}$$

$$\downarrow \hat{\mathbf{H}} = R + \frac{1}{d} \mathbf{t} \mathbf{n}^{\mathsf{T}}$$

$$\hat{\mathbf{x}}' = \hat{\mathbf{H}} \hat{\mathbf{x}}$$

$$\downarrow \hat{\mathbf{x}} = K^{-1} \mathbf{x} \quad \text{and} \quad \hat{\mathbf{x}}' = K'^{-1} \mathbf{x}'$$

$$\mathbf{x}' = \mathbf{H} \mathbf{x}$$

- Conversion to calibrated H: $\widehat{H} = K'^{-1}HK$
- **Decomposition**: cv.decomposeHomographyMat() → 4 solutions "relative pose ambiguity"
- Degenerate cases: Correspondence not from a single plane

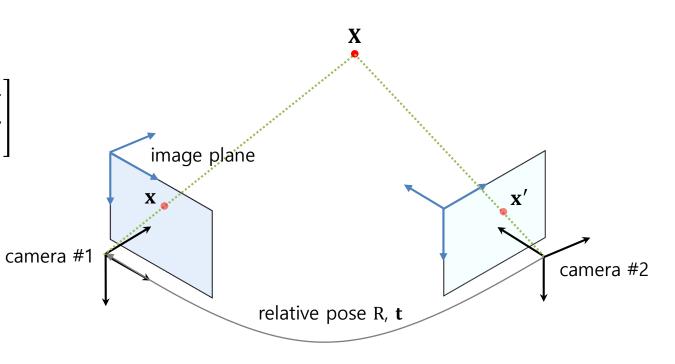
Relative Camera Pose Estimation: Overview

| | Items | General 2D-2D Geometry | Planar 2D-2D Geometry | |
|------------------|--------------------------|--|--|--|
| On Image Planes | Model | Fundamental Matrix (7 DOF) | Planar Homography (8 DOF) | |
| | Formulation | $F = K'^{-1}EK^{-1}$ $E = K'^{T}FK$ | $H = K'\widehat{H}K^{-1}$ $\widehat{H} = K'^{-1}HK$ | |
| | Estimation | - 7-point algorithm $(n \ge 7) \rightarrow k$ solution - (normalized) 8-point algorithm → 1 solution - cv.findFundamentalMat() | - 4-point algorithm (n ≥ 4) → 1 solution - cv::findHomography() | |
| | Input | - $(\mathbf{x}_i, \mathbf{x}_i')$ [px] on the image plane | - $(\mathbf{x}_i, \mathbf{x}_i')$ [px] on a plane in the image plane | |
| | Degenerate Cases | No translational motionCorrespondence from a single plane | - Correspondence <u>not</u> from a single plane | |
| | Decomposition to R and t | - Convert to an essential matrix and decompose it | - cv.decomposeHomographyMat() | |
| les | Model | Essentials Matrix (5 DOF) | (Calibrated) Planar Homgraphy (8 DOF) | |
| Je Planes | Formulation | $E = [\mathbf{t}]_{\times} R$ | $\widehat{\mathbf{H}} = \mathbf{R} + \frac{1}{d} \mathbf{t} \mathbf{n}^{T}$ | |
| d Imaç | Estimation | - 5-point algorithm $(n \ge 5)$ → k solution - cv.findEssentialMat() | 4-point algorithm (n ≥ 4) → 1 solution cv::findHomography() | |
| Normalized Image | Input | - $(\hat{\mathbf{x}}_i, \hat{\mathbf{x}}_i')$ [m] on the normalized image plane | - $(\hat{\mathbf{x}}_i, \hat{\mathbf{x}}_i')$ [m] on a plane in the normalized image plane | |
| | Degenerate Cases | - No translational motion | - Correspondence not from a single plane | |
| On N | Decomposition to R and t | <pre>- cv.decomposeEssentialMat() - cv.recoverPose()</pre> | - cv.decomposeHomographyMat() with $K=I_{3\times 3}$ | |

Triangulation

- Triangulation (point localization)
 - Unknown: Position of a 3D point X (3 DOF)
 - Given: Point correspondence (x, x'), camera matrices (K, K'), and relative pose (R, t)
 - Constraints: $\mathbf{x} = K[I \mid \mathbf{0}] \mathbf{X}, \mathbf{x}' = K'[R \mid \mathbf{t}] \mathbf{X}$
 - Solution
 - OpenCV cv.triangulatePoints()
 - Special case) Stereo cameras

$$R = I_{3\times 3}, \mathbf{t} = \begin{bmatrix} -b \\ 0 \\ 0 \end{bmatrix}, \text{ and } K = K' = \begin{bmatrix} f & 0 & c_x \\ 0 & f & c_y \\ 0 & 0 & 1 \end{bmatrix}$$
$$\therefore Z = \frac{f}{x-x'}b$$

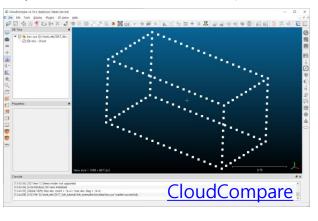


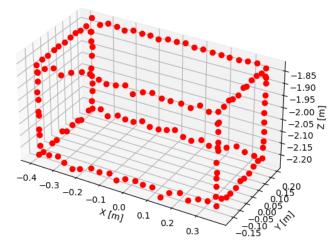
Triangulation

Example) Triangulation

```
f, cx, cy = 1000., 320., 240.
pts0 = np.loadtxt('../data/image_formation0.xyz')[:,:2]
pts1 = np.loadtxt('../data/image_formation1.xyz')[:,:2]
output file = '../data/triangulation.xyz'
# Estimate relative pose of two view
F, _ = cv.findFundamentalMat(pts0, pts1, cv.FM_8POINT)
K = np.array([[f, 0, cx], [0, f, cy], [0, 0, 1]])
E = K.T @ F @ K
_, R, t, _ = <a href="mailto:cv.recoverPose">cv.recoverPose</a>(E, pts0, pts1)
# Reconstruct 3D points (triangulation)
P0 = K @ np.eye(3, 4, dtype=np.float32)
Rt = np.hstack((R, t))
P1 = K @ Rt
X = \frac{\text{cv.triangulatePoints}}{\text{(P0, P1, pts0.T, pts1.}} K[I]0]X
X /= X[3]
                                                     \mathbf{x}' = \mathbf{K}' [\mathbf{R} \mid \mathbf{t}] \mathbf{X}
X = X.T
# Write the reconstructed 3D points
np.savetxt(output file, X)
```

A point cloud: data/box.xyz





output_file: data/triangulation.xyz

Summary

- Planar Homography: $\mathbf{x}_i' = \mathbf{H}\mathbf{x}_i$
 - Example) Perspective distortion correction
 - Example) Planar image stitching
 - Example) 2D video stabilization
- **Epipolar Geometry**: $\mathbf{x'}^\mathsf{T} \mathbf{F} \mathbf{x} = \mathbf{0}$ (on the image plane), $\hat{\mathbf{x}'}^\mathsf{T} \mathbf{E} \hat{\mathbf{x}} = \mathbf{0}$ (on the *normalized* image plane)
 - Example) Epipolar line visualization
- Relative Camera Pose Estimation: Finding R and t (5 DOF)
 - Solutions) Fundamental matrix: 7/8-point algorithm (7 DOF)
 - Solutions) Essential matrix: 5-point algorithm (5 DOF)
 - Solutions) Planar homography: 4-point algorithm (8 DOF)
 - Example) Fundamental matrix estimation
 - Example) Monocular Visual Odometry (Epipolar Version)
- Triangulation: Finding X (3 DOF)
 - Example) Triangulation

$$R, \mathbf{t}$$

$$\updownarrow (E = [\mathbf{t}]_{\times}R)$$
 E

$$\updownarrow (E = K'^{\mathsf{T}}FK)$$
 F