

Computer Vision HW3 Report

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Part 1: Homography estimation



Part 2: Marker-Based Planar AR

1. solve_homography(u, v)

```
def solve_homography(u, v):  
    """  
    This function should return a 3-by-3 homography matrix,  
    u, v are N-by-2 matrices, representing N corresponding points for  $v = T(u)$   
    :param u: N-by-2 source pixel location matrices  
    :param v: N-by-2 destination pixel location matrices  
    :return:  
    """  
    N = u.shape[0]  
    H = None  
    if v.shape[0] is not N:  
        print('u and v should have the same size')  
        return None  
    if N < 4:  
        print('At least 4 points should be given')  
    # TODO: 1. forming A  
    A_for_vx = np.concatenate((u, np.ones((N, 1)), np.zeros((N, 3))),  
                               (-1 * np.multiply(u[:, 0], v[:, 0])).reshape((N, 1)), # -ux * vx  
                               (-1 * np.multiply(u[:, 1], v[:, 0])).reshape((N, 1)), # -uy * vx  
                               (-1 * v[:, 0]).reshape((N, 1))), axis=1) # -vx  
    A_for_vy = np.concatenate((np.zeros((N, 3)), u, np.ones((N, 1))),  
                               (-1 * np.multiply(u[:, 0], v[:, 1])).reshape((N, 1)), # -ux * vy  
                               (-1 * np.multiply(u[:, 1], v[:, 1])).reshape((N, 1)), # -uy * vy  
                               (-1 * v[:, 1]).reshape((N, 1))), axis=1) # -vy  
    A = np.concatenate((A_for_vx, A_for_vy), axis=0)  
    # TODO: 2. solve H with A  
    VT = np.linalg.svd(A)  
    H = VT.T[:3, -1]  
    H = H / H[-1]  
    H = H.reshape(3, 3)  
    return H
```

2. warping()

- The interpolation method I use is “nearest neighbor” (np.rint() in the code).

Code:

```
def warping(src, dst, H, ymin, ymax, xmin, xmax, direction='b'):
    """
    Perform forward/backward warping without for loops. i.e.
    for all pixels in src(xmin~xmax, ymin~ymax), warp to destination
    (xmin=0,ymin=0) source destination
    """
    # forward warp
    # (xmin=0,ymin=0) source destination
    # (xmax=w,ymax=h)
    # backward warp
    # (xmin=0,ymin=0) source destination
    # (xmax=w,ymax=h)

    :param src: source image
    :param dst: destination output image
    :param H:
    :param ymin: lower vertical bound of the destination(source, if forward warp) pixel coordinate
    :param ymax: upper vertical bound of the destination(source, if forward warp) pixel coordinate
    :param xmin: lower horizontal bound of the destination(source, if forward warp) pixel coordinate
    :param xmax: upper horizontal bound of the destination(source, if forward warp) pixel coordinate
    :param direction: indicates backward warping or forward warping
    :return: destination output image
    """

    h_src, w_src, ch = src.shape
    h_dst, w_dst, ch = dst.shape
    H_inv = np.linalg.inv(H)

    # TODO: 1.meshgrid the (x,y) coordinate pairs
    x, y = np.meshgrid(range(xmin, xmax), range(ymin, ymax))

    # TODO: 2.reshape the destination pixels as N x 3 homogeneous coordinate
    N = (xmax - xmin) * (ymax - ymin)
    U = np.concatenate((x.reshape((1, N)), y.reshape((1, N)), np.ones((1, N))), axis=0)

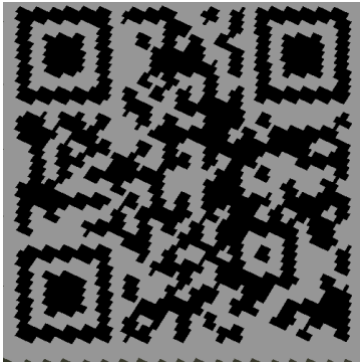
    if direction == 'b':
        # TODO: 3.apply H_inv to the destination pixels and retrieve (u,v) pixels, then reshape to (ymax-ymin),(xmax-xmin)
        V = np.dot(H_inv, U)
        V = np.divide(V, V[-1, :])
        Vx = np rint(V[0, :].reshape((ymax-ymin),(xmax-xmin)))
        Vy = np rint(V[1, :].reshape((ymax-ymin),(xmax-xmin)))
        # TODO: 4.calculate the mask of the transformed coordinate (should not exceed the boundaries of source image)
        x_mask = np.bitwise_and((Vx >= 0), (Vx < w_src))
        y_mask = np.bitwise_and((Vy >= 0), (Vy < h_src))
        mask = np.bitwise_and(x_mask, y_mask)
        # TODO: 5.sample the source image with the masked and reshaped transformed coordinates
        # TODO: 6. assign to destination image with proper masking
        valid_Vx = Vx[mask].astype(int)
        valid_Vy = Vy[mask].astype(int)
        dst[y[mask], x[mask]] = src[valid_Vy, valid_Vx]
        # pass

    elif direction == 'f':
        # TODO: 3.apply H to the source pixels and retrieve (u,v) pixels, then reshape to (ymax-ymin),(xmax-xmin)
        V = np.dot(H, U)
        V = np.divide(V, V[-1, :])
        Vx = np rint(V[0, :].reshape((ymax-ymin),(xmax-xmin)))
        Vy = np rint(V[1, :].reshape((ymax-ymin),(xmax-xmin)))
        # TODO: 4.calculate the mask of the transformed coordinate (should not exceed the boundaries of destination image)
        x_mask = np.bitwise_and((Vx >= 0), (Vx < w_dst))
        y_mask = np.bitwise_and((Vy >= 0), (Vy < h_dst))
        mask = np.bitwise_and(x_mask, y_mask)
        # TODO: 5.filter the valid coordinates using previous obtained mask
        # TODO: 6. assign to destination image using advanced array indexing
        valid_Vx = Vx[mask].astype(int)
        valid_Vy = Vy[mask].astype(int)
        dst[valid_Vy, valid_Vx, :] = src[mask]
        # pass

    return dst
```

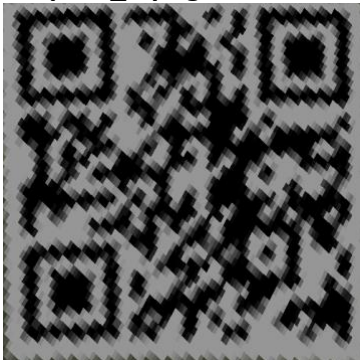
Part 3: Unwarp the secret

1. output3_1.png



<http://media.ee.ntu.edu.tw/courses/cv/21S/>

2. output3_2.png



<http://media.ee.ntu.edu.tw/courses/cv/21S/>

兩張 source images 產生的結果中，第一張比第二張清楚，第二張明顯比較模糊，但兩者都能掃出來同樣的網址。比較原圖，可以觀察到第一張的 QR Code 還算方正，看起來像單純旋轉的正方形，不過因為 pixel 被轉成斜的，所以轉回來的圖沒法像正常的 QR Code 一樣都是直線；第二張原圖可以看出 QR Code 的形狀有變，被拉成接近長方形，邊界也有點弧度，要轉回正方形的 QR Code 就可能會相較困難，pixels 要找對應位置會比較容易偏移，在黑灰邊界就容易有模糊現象。

不過雖然第二張比較不清楚，但黑灰分別的圖案還是有保留下來，所以即使邊界模糊，還是能掃出一樣的網址。

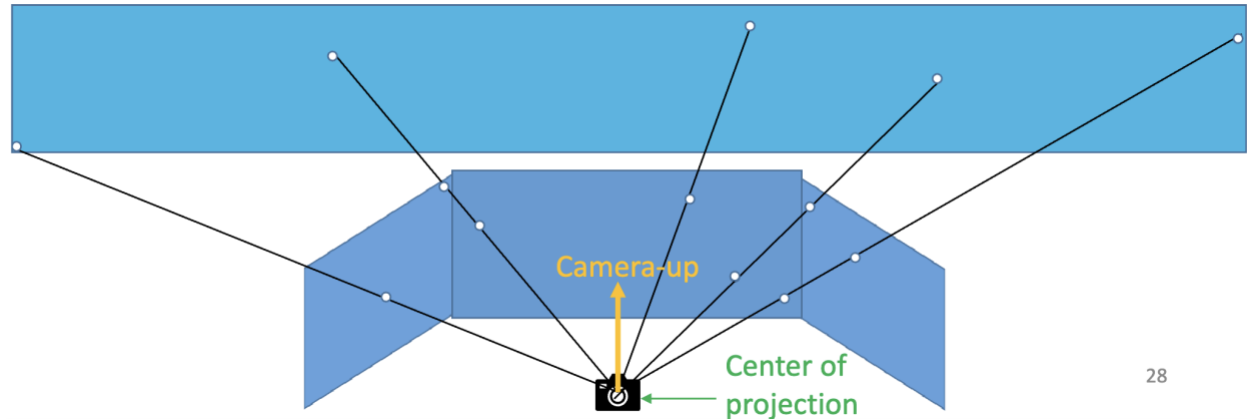
Part 4: Panorama



Can all consecutive images be stitched into a panorama?

不行。以這次作業使用的方式，投射到平面，那圖片的視角不能旋轉超過 180 度。

從下圖（擷取自作業投影片）中可看到，假設圖片上方為 0 度，那正負 90 度的地方就會投影到無限遠處，超過這總共 180 度的範圍，就會往圖片下方投影，無法與圖片上的點共平面。



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實際比較例子可看作業得到的 `output4.png` 和下圖。`output4.png` 中視角在範圍內，可以看到除了接面處因為不同圖片亮度不同，所以看得出接線外，圖形都有連在一起，形成一張不錯的全景圖。另外下面前四張圖片是素材，第五張是全景圖結果，可以看到如果正常接起來，最右邊應該會有鏡子、架子，但或許因為視角差太大，導致無法完整產生全景圖。

