# Initialization

Run the following code to import the modules you'll need. After your finish the assignment, **remember to run all cells** and save the note book to your local machine as a PDF for gradescope submission.

```
import os
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.patches as patches
```

### Download data

In this section we will download the data and setup the paths.

```
In []: # Download the data (Commented out because downloaded onto local machine)
    # if not os.path.exists('/content/carseq.npy'):
    # !wget https://www.cs.cmu.edu/~deva/data/carseq.npy -0 /content/carseq.npy
    # if not os.path.exists('/content/girlseq.npy'):
    # !wget https://www.cs.cmu.edu/~deva/data/girlseq.npy -0 /content/girlseq.npy
```

# Q2.1: Theory Questions (5 points)

Please refer to the handout for the detailed questions.

# Q2.1.1: What is $\frac{\partial \mathbf{W}(\mathbf{x}; \mathbf{p})}{\partial \mathbf{p}^T}$ ? (**Hint**: It should be a 2x2 matrix)

==== your answer here! =====

$$W(x;p) = x + p = \left[egin{array}{c} x + p_x \ y + p_y \end{array}
ight]$$

$$rac{\partial \mathbf{W}(\mathbf{x};\mathbf{p})}{\partial \mathbf{p}^T} = egin{bmatrix} 1 & 0 \ 0 & 1 \end{bmatrix}$$

==== end of your answer =====

## Q2.1.2: What is $\mathbf{A}$ and $\mathbf{b}$ ?

```
===== your answer here! =====
A = \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} b = T(X) - I(W(x;p)) \mathbf{x} \text{ is } \Delta p. \mathbf{A}\mathbf{x} = \mathbf{b} ===== \text{end of your answer} ======
```

# Q2.1.3 What conditions must $\mathbf{A}^T \mathbf{A}$ meet so that a unique solution to $\Delta \mathbf{p}$ can be found?

```
==== your answer here! ===== A^TA must be positive semi-definite. ===== end of your answer =====
```

# Q2.2: Lucas-Kanade (20 points)

Make sure to comment your code and use proper names for your variables.

```
In [ ]: from scipy.interpolate import RectBivariateSpline
        from numpy.linalg import lstsq
        def LucasKanade(It, It1, rect, threshold, num_iters, p0):
            :param[np.array(H, W)] It : Grayscale image at time t [float]
            :param[np.array(H, W)] It1 : Grayscale image at time t+1 [float]
            :param[np.array(4, 1)] rect : [x1 y1 x2 y2] coordinates of the rectangular temp
                                          where [x1, y1] is the top-left, and [x2, y2] is t
                                          [floats] that maybe fractional.
            :param[float] threshold
                                      : If change in parameters is less than thresh, term
                                       : Maximum number of optimization iterations
            :param[int] num_iters
            :param[np.array(2, 1)] p0 : Initial translation parameters [p_x0, p_y0] to ad
            :return[np.array(2, 1)] p : Final translation parameters [p_x, p_y]
            # ===== your code here! =====
            # Hint: Iterate over num_iters and for each iteration, construct a linear syste
            # Construct [A] by computing image gradients at (possibly fractional) pixel loc
            # We suggest using RectBivariateSpline from scipy.interpolate to interpolate pi
            # We suggest using lstsq from numpy.linalg to solve the linear system
            # Once you solve for [delta_p], add it to [p] (and move on to next iteration)
            # HINT/WARNING:
```

```
# RectBivariateSpline and Meshgrid use inconsistent defaults with respect to 'x
# https://docs.scipy.org/doc/scipy/reference/generated/scipy.interpolate.RectBi
# https://numpy.org/doc/stable/reference/generated/numpy.meshgrid.html
p = p0
px = p[0]
py = p[1]
i = 0
delta_p_norm_square = threshold
x1 = rect[0]
y1 = rect[1]
x2 = rect[2]
y2 = rect[3]
\# T_t = It[y1:y2+1,x1:x2+1]
# T_t_flat = np.reshape(T_t, (T_t.size,1))
dWdp = np.eye(2)
height = It.shape[0]
width = It.shape[1]
height_arr = np.arange(height)
width_arr = np.arange(width)
rbs_T_t = RectBivariateSpline(height_arr,width_arr,It)
iX = np.arange(x1+px,x2+px+0.5,1)
iY = np.arange(y1+py,y2+py+0.5,1)
xgrid, ygrid = np.meshgrid(iX,iY,indexing="ij")
T_t = rbs_T_t.ev(ygrid,xgrid).T
T_t_flat = np.reshape(T_t, (T_t.size,1))
rbs_IW = RectBivariateSpline(height_arr, width_arr, It1)
gradientAllX = np.gradient(It1,axis=1)
gradientAllY = np.gradient(It1,axis=0)
rbs gx = RectBivariateSpline(height arr,width arr,gradientAllX)
rbs_gy = RectBivariateSpline(height_arr, width_arr, gradientAllY)
while (i < num_iters) and (delta_p_norm_square >= threshold):
    px = p[0]
    py = p[1]
    iX = np.arange(x1+px,x2+px+0.5,1)
    iY = np.arange(y1+py,y2+py+0.5,1)
    xgrid, ygrid = np.meshgrid(iX,iY,indexing="ij")
    IW = rbs_IW.ev(ygrid,xgrid).T
    IW_flat = np.reshape(IW, (IW.size,1))
    gradientVecX = np.ravel(rbs_gx.ev(ygrid,xgrid).T)
    gradientVecY = np.ravel(rbs_gy.ev(ygrid,xgrid).T)
    gradientMatrix = np.vstack((gradientVecX,gradientVecY)).T # (pixPatch,2) gr
    A = gradientMatrix@dWdp
    b = T t flat-IW flat
```

```
#delta_p = np.linalg.lstsq(A.T@A,A.T@b,rcond=None)[0]
delta_p = np.linalg.lstsq(A,b,rcond=None)[0]

p = p + delta_p

delta_p_norm_square = np.square(np.linalg.norm(delta_p,ord=2,keepdims=True)
i = i + 1

# plt.figure(2)
# plt.imshow(T_t)
# plt.figure(3)
# plt.imshow(IW)
# ===== End of code =====
return p
```

# Debug Q2.2

A few tips to debug your implementation:

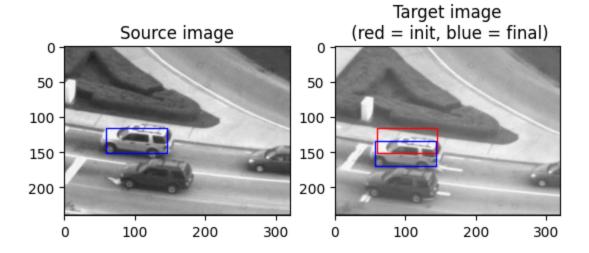
- Feel free to use and modify the following snippet to debug your implementation. The snippet simply visualizes the translation resulting from running LK on a single frame. You should be able to see a slight shift in the template.
- You may also want to visualize the image gradients you compute within your LK implementation
- Plot iterations vs the norm of delta p

```
In [ ]: def draw_rect(rect,color):
            w = rect[2] - rect[0]
            h = rect[3] - rect[1]
            plt.gca().add_patch(patches.Rectangle((rect[0],rect[1]), w, h, linewidth=1, edg
In [ ]: num_iters = 100
        threshold = 0.01
        seq = np.load("data\\carseq.npy")
        rect = np.array([59, 116, 145, 151])
        rect = np.reshape(rect, (rect.size,1))
        It = seq[:,:,0]
        # Source frame
        plt.figure()
        plt.subplot(1,2,1)
        plt.imshow(It, cmap='gray')
        plt.title('Source image')
        draw_rect(rect, 'b')
        # Target frame + LK
        It1 = seq[:,:, 20]
        plt.subplot(1,2,2)
```

```
plt.imshow(It1, cmap='gray')
plt.title('Target image\n (red = init, blue = final)')
p = LucasKanade(It, It1, rect, threshold, num_iters, np.zeros((2,1)))

rect_t1 = (rect + np.concatenate((p,p)))

draw_rect(rect,'r')
draw_rect(rect_t1,'b')
```



# Q2.3: Tracking with template update (15 points)

```
In [ ]: def TrackSequence(seq, rect, num_iters, threshold):
                              : (H, W, T), sequence of frames
            :param seq
                             : (4, 1), coordinates of template in the initial frame. top-le
            :param num_iters : int, number of iterations for running the optimization
            :param threshold : float, threshold for terminating the LK optimization
            :return: rects : (T, 4) tracked rectangles for each frame
            0.00
            H, W, N = seq.shape
            rects = np.zeros((N,4))
            It = seq[:,:,0]
            pNow = np.zeros((2,1))
            \#rect = np.reshape(rect, (4,1))
            # Iterate over the car sequence and track the car
            for i in range(seq.shape[2]):
                # ===== your code here! =====
                # TODO: add your code track the object of interest in the sequence
                if (i == seq.shape[2]-1):
                    It = seq[:,:,i]
                    It1 = seq[:,:,i]
                else:
                    It = seq[:,:,i]
                    It1 = seq[:,:,i+1]
                pNow = LucasKanade(It,It1,rect,threshold,num_iters,pNow)
                rects[i,:] = (np.copy(rect) + np.concatenate((pNow,pNow))).T
```

```
# ==== End of code =====

#rects = np.array(rects)
assert rects.shape == (N, 4), f"Your output sequence {rects.shape} is not ({N}x
return rects
```

#### Q2.3 (a) - Track Car Sequence

Run the following snippets. If you have implemented LucasKanade and TrackSequence function correctly, you should see the box tracking the car accurately. Please note that the tracking might drift slightly towards the end, and that is entirely normal.

Feel free to play with these snippets of code by playing with the parameters.

```
In [ ]:
    def visualize_track(seq,rects,frames,rect):
        # Visualize tracks on an image sequence for a select number of frames
        plt.figure(figsize=(15,15))
        for i in range(len(frames)):
            idx = frames[i]
            frame = seq[:, :, idx]
            plt.subplot(1,len(frames),i+1)
            plt.imshow(frame, cmap='gray')
            plt.axis('off')
            draw_rect(rects[idx],'b')
```

```
In []: seq = np.load("data\\carseq.npy")
    rect = np.array([59, 116, 145, 151])
    rect = np.reshape(rect, (rect.size,1))

# NOTE: feel free to play with these parameters
    num_iters = 10000
    threshold = 0.01

rects = TrackSequence(seq, rect, num_iters, threshold)

visualize_track(seq,rects,[0, 79, 159, 279, 409],rect)
#scale = 20
#visualize_track(seq,rects,[0*scale, 1*scale, 2*scale, 3*scale, 4*scale],rect) # Ch
```











#### Q2.3 (b) - Track Girl Sequence

Same as the car sequence.

```
In [ ]: # Loads the squence
    seq = np.load("data\\girlseq.npy")
    rect = np.array([280, 152, 330, 318])
```

```
rect = np.reshape(rect, (rect.size,1))

# NOTE: feel free to play with these parameters
num_iters = 10000
threshold = 0.01

rects = TrackSequence(seq, rect, num_iters, threshold)

visualize_track(seq,rects,[0, 14, 34, 64, 84],rect)
```









