

Advanced Computer Vision – HW3

Optical Flow Estimation

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1. Description

- Implementing *Horn-Schunck* optical flow estimation
- Synthetically translate lena.bmp one pixel to the right and downward
- Try λ of 0.1, 1, 10.

2. Method

- The *Horn-Schunck* method of estimating optical flow is a method that enforce two constraints, which are brightness constancy and smooth flow field.
- Use the image transformation method to translate lena.bmp to lena_shifted.bmp by moving one pixel to the right and downward.

```
def translate(img, dx, dy):  
    T = np.float32([[1, 0, dx], [0, 1, dy]])  
    shifted = cv2.warpAffine(img, T, (img.shape[1], img.shape[0]))  
    return shifted
```

- Use *Horn-Schunck* algorithm as below to estimate the optical flow. More detail is referenced from [1].
 1. Precompute image gradients I_x, I_y .
 2. Precompute temporal gradients I_t .
 3. Initialize flow field $u = 0, v = 0$.
 4. While not converged, compute flow field updates for each pixel:

$$\hat{u}_{kl} = \bar{u}_{kl} - \frac{I_x \bar{u}_{kl} + I_y \bar{v}_{kl} + I_t}{\lambda^{-1} + I_x^2 + I_y^2} I_x$$
$$\hat{v}_{kl} = \bar{v}_{kl} - \frac{I_x \bar{u}_{kl} + I_y \bar{v}_{kl} + I_t}{\lambda^{-1} + I_x^2 + I_y^2} I_y$$

```

def HornSchunck(img1, img2, lam=1, N_iter=8):
    # Laplacian kernel
    L_kernel = np.array([[1/12, 1/6, 1/12],
                          [1/6, 0, 1/6],
                          [1/12, 1/6, 1/12]], float)

    img1 = img1.astype(np.float32)
    img2 = img2.astype(np.float32)

    # set up initial velocities
    uInitial = np.zeros([img1.shape[0], img1.shape[1]])
    vInitial = np.zeros([img1.shape[0], img1.shape[1]])

    # set initial value for the flow vectors
    u = uInitial
    v = vInitial

    # calculate derivatives
    [fx, fy, ft] = calDerivatives(img1, img2)

    for _ in range(N_iter):
        # calculate the average of neighborhood velocity
        u_Avg = filter2D(u, L_kernel)
        v_Avg = filter2D(v, L_kernel)
        der = (fx*u_Avg + fy*v_Avg + ft) / (1 + lam*(fx**2 + fy**2))
        u = u_Avg - fx * der
        v = v_Avg - fy * der

    return u, v

```

3. Result

- Iteration = 1:



Figure 1. result for Lambda = 0.1



Figure 2. result for $\Lambda = 1$

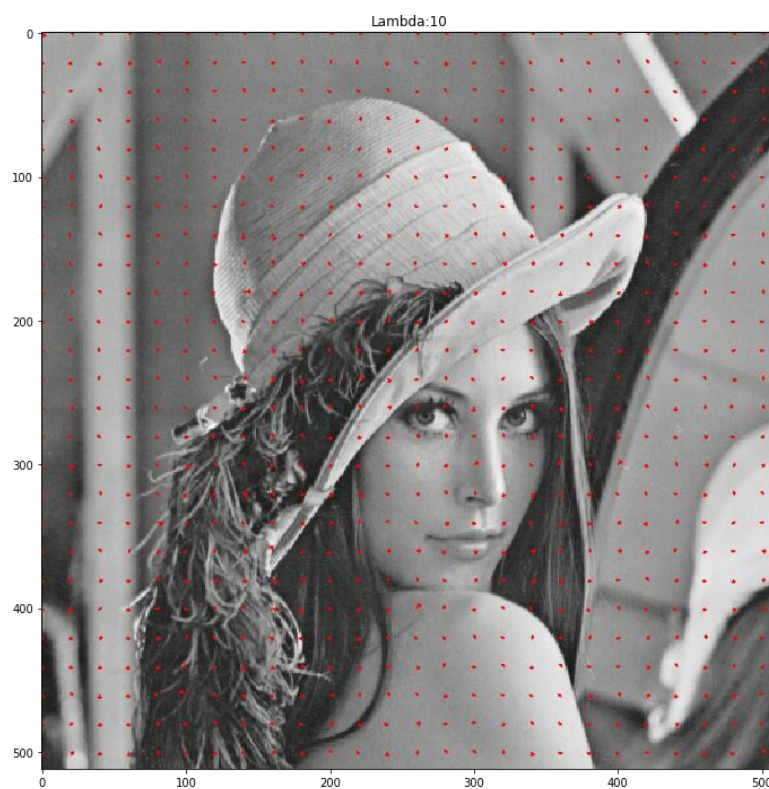


Figure 3. result for $\Lambda = 10$

- Iteration = 4:



Figure 4. result for $\text{Lambda} = 0.1$



Figure 5. result for $\text{Lambda} = 1$

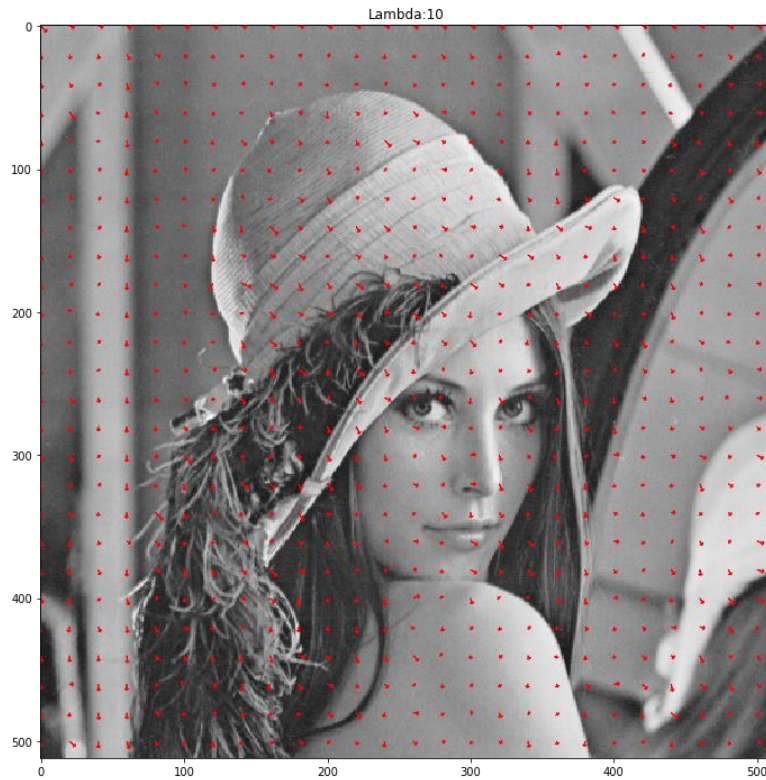


Figure 6. result for $\text{Lambda} = 10$

- Iteration = 16:

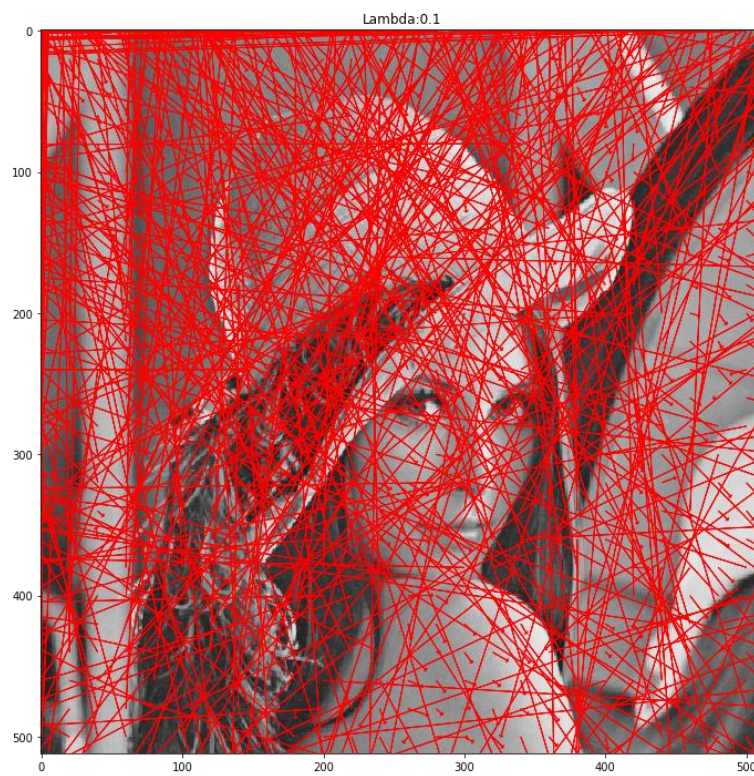


Figure 7. result for $\text{Lambda} = 0.1$



Figure 8. result for $\Lambda = 1$



Figure 9. result for $\Lambda = 10$

- Iteration = 64:

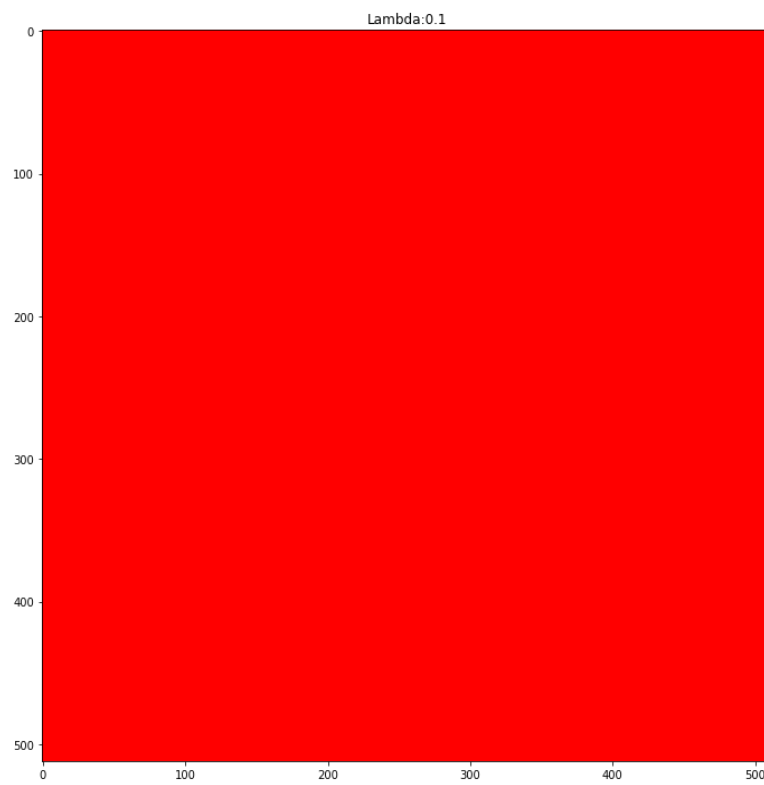


Figure 10. result for Lambda = 0.1



Figure 11. result for Lambda = 1



Figure 11. result for $\Lambda = 10$

- Reference

[1] <http://datahacker.rs/013-optical-flow-using-horn-and-schunck-method/>