Foundations of Data Science Image Filtering and Object Identification

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1 Image Filtering

1.1 Question 1.d

The effect of applying a filter can be studied by observing its *impulse response*. Executing the following snippet we created a test image (Figure 1) in which only the central pixel has a non-zero value:

```
img_imp = np.zeros([27,27])
img_imp[13, 13] = 1.0
plt.figure(1), plt.imshow(img_imp, cmap='gray')
```

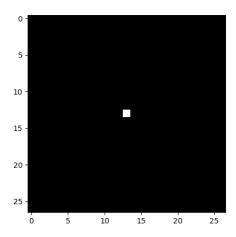


Figure 1: Test Image

Executing the following snippet we created 1D Gaussian and Gaussian derivative kernels, Gx and Dx respectively.

```
sigma = 7.0
[Gx, x] = gauss_module.gauss(sigma)
[Dx, x] = gauss_module.gaussdx(sigma)
```

We applied the following filter combinations:

- 1. First Gx, then Gx^T
- 2. First Gx, then Dx^T
- 3. First Dx^T , then Gx
- 4. First Dx, then Dx^T
- 5. First Dx, then Gx^T
- 6. First Gx^T , then Dx

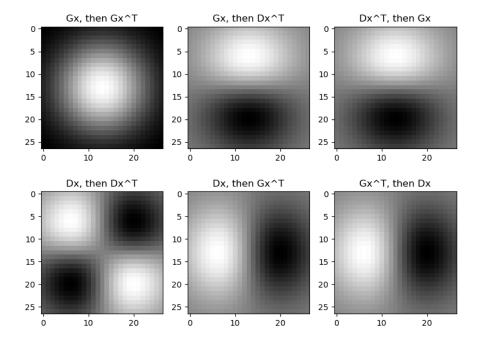


Figure 2: Applying filter combinations

As we can see in Figure 2, the first filter combination [...] **TODO** What happens when you apply the following filter combinations?

1.2 Question 1.e

We implemented a gaussderiv method that takes an input image and generates two copies of it, smoothed according to a standard deviation σ and derived in the directions x and y respectively.

The results of applying gaussderiv, with $\sigma = 7.0$, to the provided example images (graf.png and gantrycrane.png) are shown in Figures 5 and 6.



Figure 3: graf.png



Figure 4: gantrycrane.png

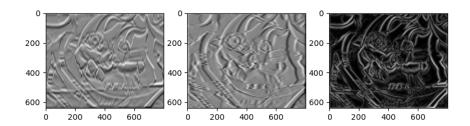


Figure 5: Results of applying gaussderiv on graf.png

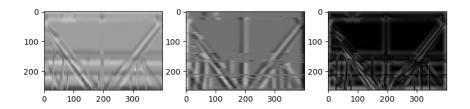


Figure 6: Results of applying gaussderiv on gantrycrane.png

TODO Comment on the output in your report.
TODO Consider also why smoothing an image is important before applying the derivative filter.