

```
In [1]: %matplotlib inline
from __future__ import print_function
#import ganymede
#ganymede.configure('uav.beaver.works')
import matplotlib.pyplot as plt
import numpy as np
import cv2
import os
```

```
In [ ]: def check(p): pass
check(0)
```

## Note

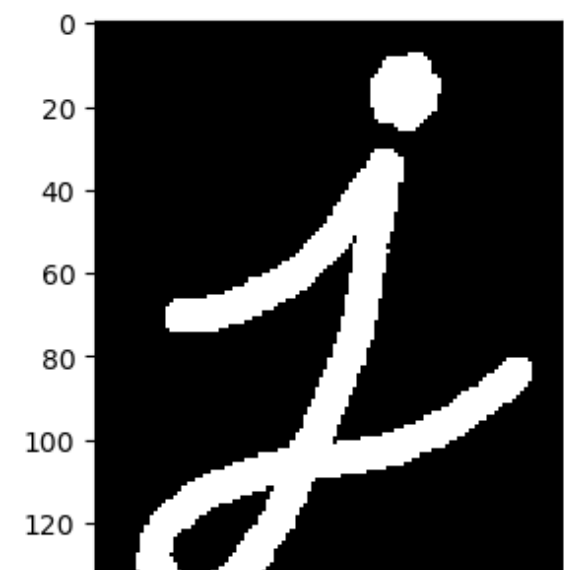
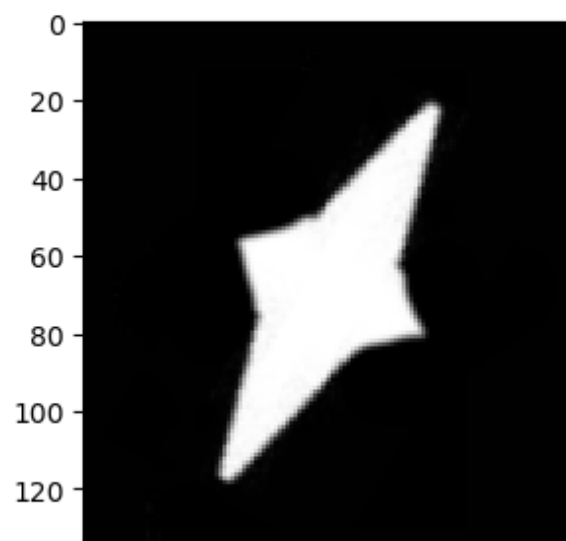
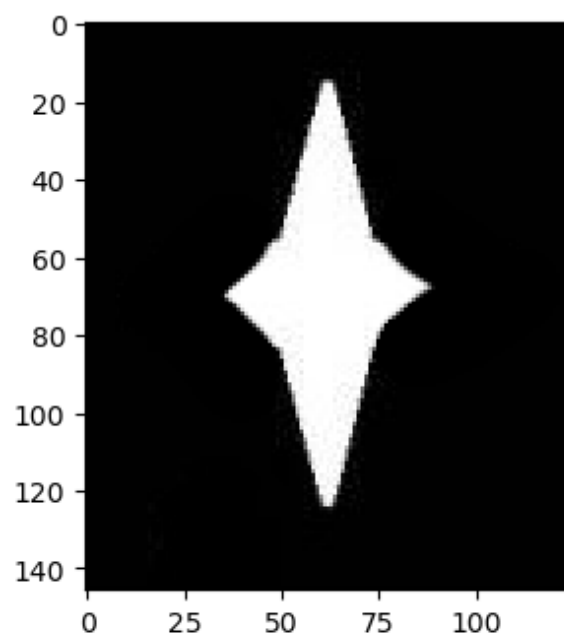
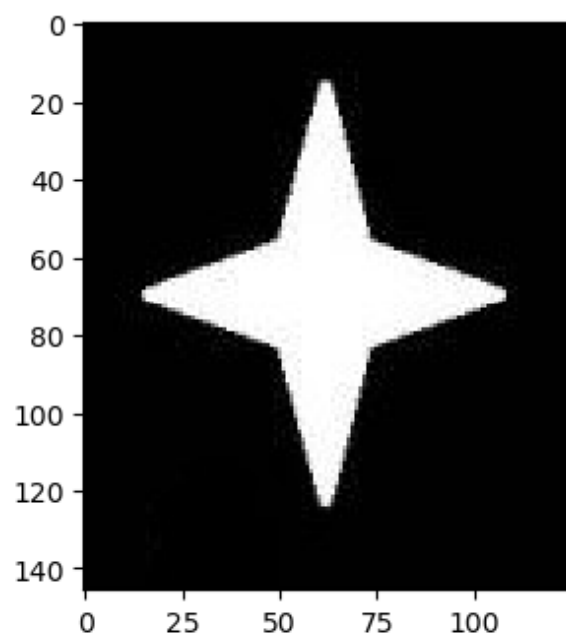
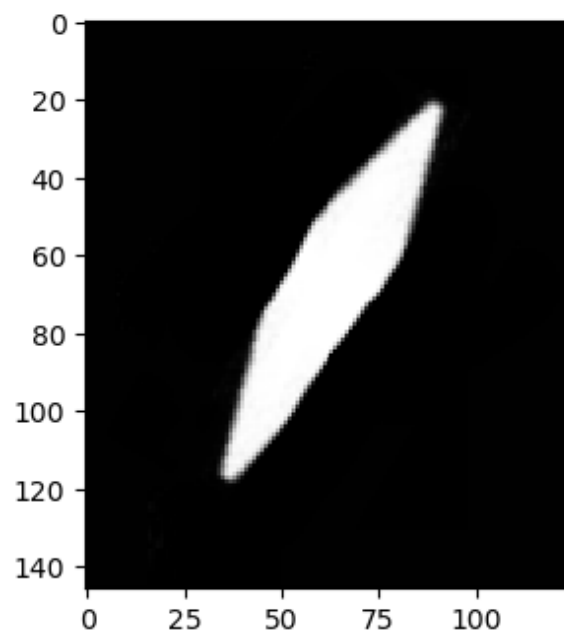
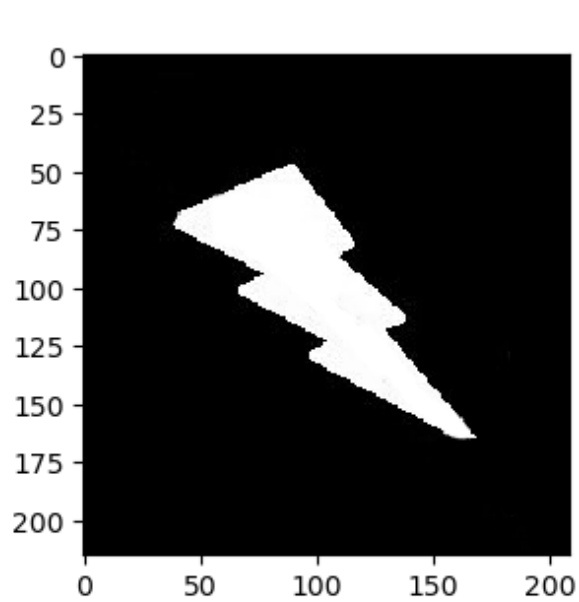
`cv2.imshow()` will not work in a notebook, even though the OpenCV tutorials use it. Instead, use `plt.imshow` and family to visualize your results.

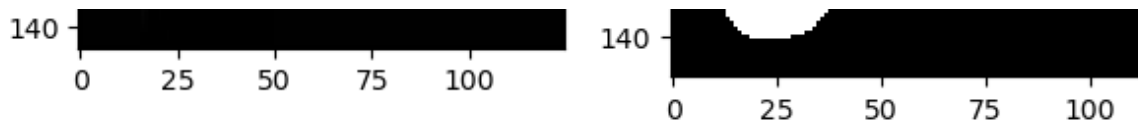
```
In [17]: #Charlie Lai

lightningbolt      = cv2.imread('shapes/lightningbolt.png', cv2.IMREAD_GRAYSCALE)
blob               = cv2.imread('shapes/blob.png', cv2.IMREAD_GRAYSCALE)
star               = cv2.imread('shapes/star.png', cv2.IMREAD_GRAYSCALE)
squishedstar       = cv2.imread('shapes/squishedstar.png', cv2.IMREAD_GRAYSCALE)
squishedturnedstar = cv2.imread('shapes/squishedturnedstar.png', cv2.IMREAD_GRAYSCALE)
letterj            = cv2.imread('shapes/letterj.png', cv2.IMREAD_GRAYSCALE)

images = [lightningbolt, blob, star, squishedstar, squishedturnedstar, letterj]

fig, ax = plt.subplots(nrows=3, ncols=2)
for a, i in zip(ax.flatten(), images):
    a.imshow(i, cmap='gray', interpolation='none');
fig.set_size_inches(7, 14);
```





```
In [18]: intensity_values = set(lightningbolt.flatten())
print(len(intensity_values))
```

75

## Question:

What would you expect the value to be, visually? What explains the actual value?

```
In [ ]: # TODO
# I would expect the value to be 2 because it seems like there are only 2
```

## Thresholding

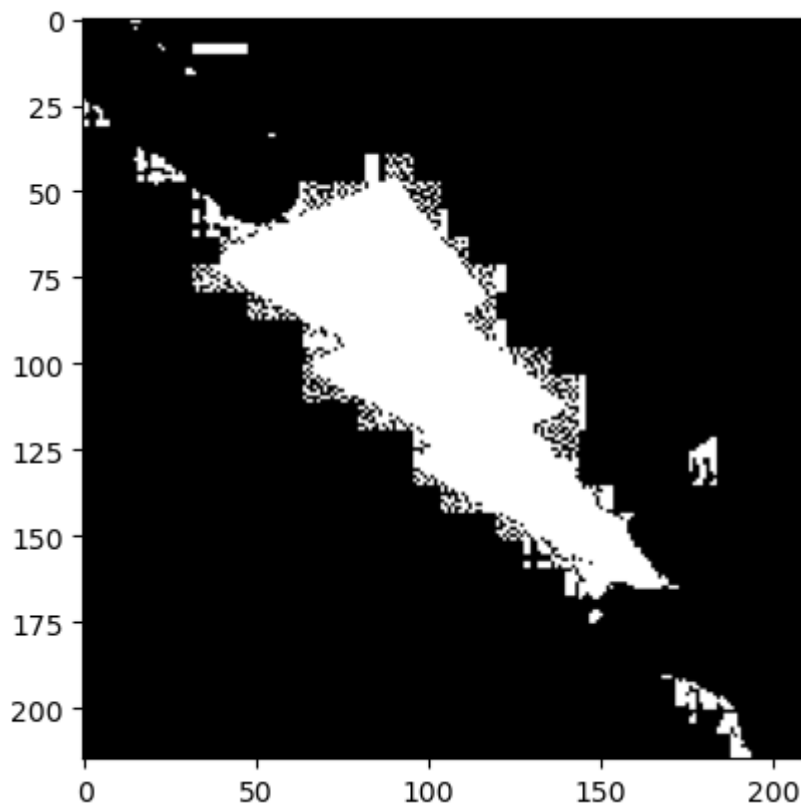
[https://docs.opencv.org/3.4.1/d7/d4d/tutorial\\_py\\_thresholding.html](https://docs.opencv.org/3.4.1/d7/d4d/tutorial_py_thresholding.html)

```
In [4]: _, lightningbolt = cv2.threshold(lightningbolt, 0, 255, cv2.THRESH_BINARY)

intensity_values = set(lightningbolt.flatten())
print(len(intensity_values))

plt.imshow(lightningbolt, cmap='gray');
```

2



## Question

What happens when the above values are used for thresholding? What is a "good" value for thresholding the above images? Why?

```
In [ ]: ## TODO
        ## It sets every value that isn't black to white. A more strict value, a
```

## Exercises

### Steps

1. Read each tutorial
  - Skim all parts of each tutorial to understand what each operation does
  - Focus on the part you will need for the requested transformation
2. Apply the transformation and visualize it

## 1. Blend lightningbolt and blob together

[https://docs.opencv.org/3.4.1/d0/d86/tutorial\\_py\\_image\\_arithmetics.html](https://docs.opencv.org/3.4.1/d0/d86/tutorial_py_image_arithmetics.html)

*Remember:* Don't use `imshow` from OpenCV, use `imshow` from `matplotlib`

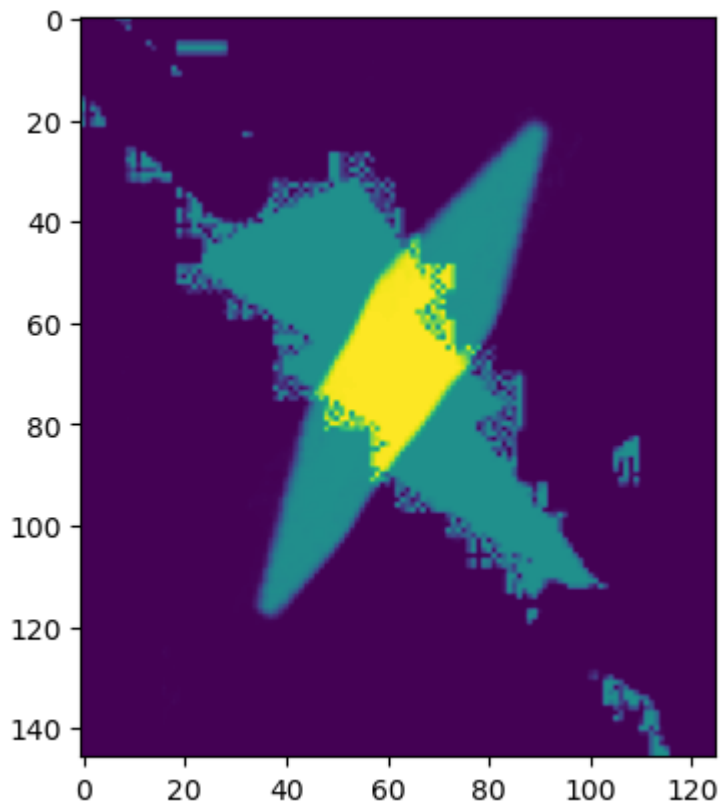
```
In [5]: # 1. Blend
        # TODO
        #
        height, width = blob.shape[:2]
        bolt_resized = cv2.resize(lightningbolt, (width, height), interpolation=c
        print(width)
        print(height)

        img = cv2.addWeighted(bolt_resized, .5, blob, .5, 0)
        plt.imshow(img)
```

125

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Out[5]: <matplotlib.image.AxesImage at 0xf1173583afe0>



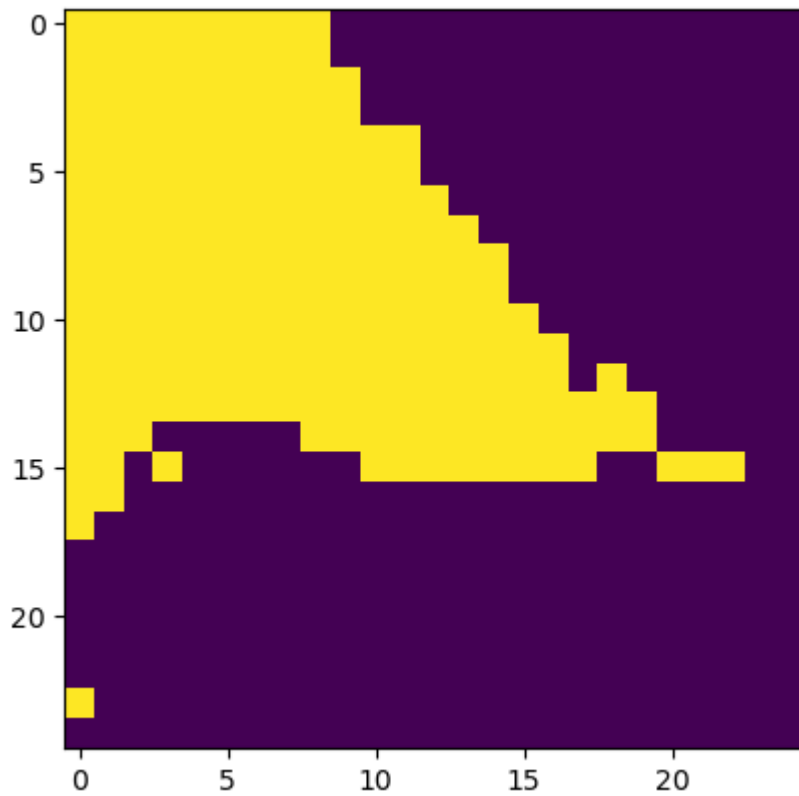
## 2. Find a ROI which contains the point of the lightning bolt

[https://docs.opencv.org/3.4.1/d3/df2/tutorial\\_py\\_basic\\_ops.html](https://docs.opencv.org/3.4.1/d3/df2/tutorial_py_basic_ops.html)

```
In [6]: # 2. ROI
# TODO

point = lightningbolt[150:175, 150:175]
plt.imshow(point)
```

```
Out[6]: <matplotlib.image.AxesImage at 0xf11735b85570>
```

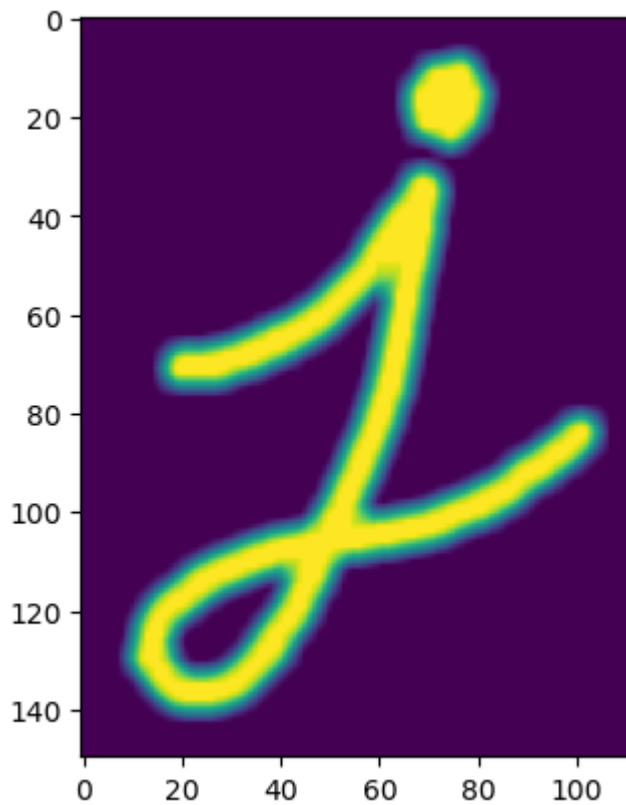


### 3. Use an averaging kernel on the letter j

[https://docs.opencv.org/3.4.1/d4/d13/tutorial\\_py\\_filtering.html](https://docs.opencv.org/3.4.1/d4/d13/tutorial_py_filtering.html)

```
In [7]: # 3.  
# TODO  
img = cv2.blur(letterj, (5,5))  
plt.imshow(img)
```

```
Out[7]: <matplotlib.image.AxesImage at 0xf11735bcf5e0>
```



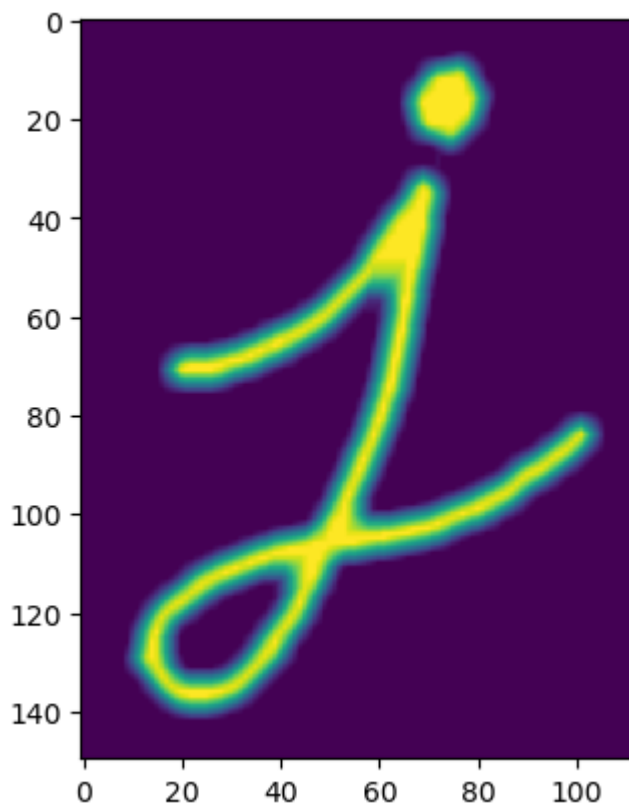
## Morphology

[https://docs.opencv.org/3.4.1/d9/d61/tutorial\\_py\\_morphological\\_ops.html](https://docs.opencv.org/3.4.1/d9/d61/tutorial_py_morphological_ops.html)

### 4. Perform erosion on j with a 3x3 kernel

```
In [8]: # 4
# TODO
kernel = np.ones((3, 3), np.uint8)
erodedJ = cv2.erode(img, kernel, iterations = 1)
plt.imshow(erodedJ)
```

Out[8]: <matplotlib.image.AxesImage at 0xf11735c432b0>

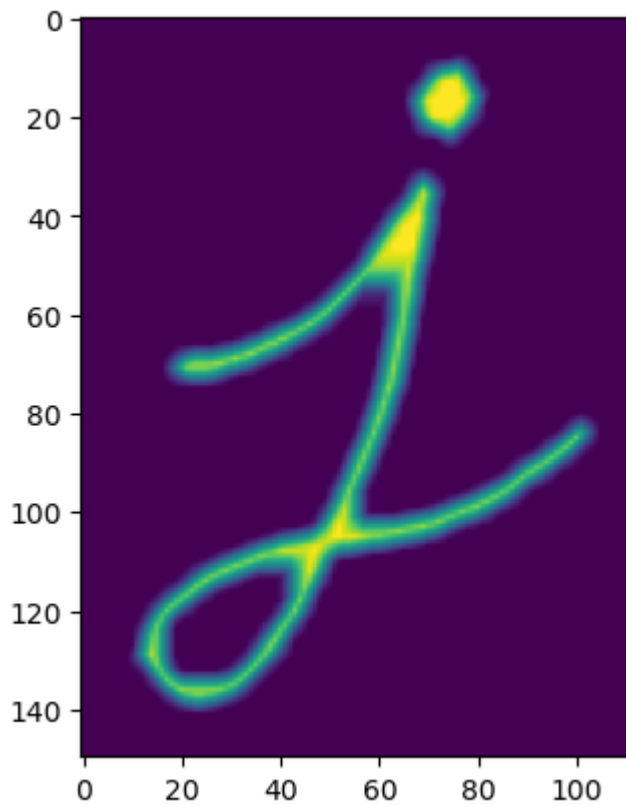


## 5. Perform erosion on j with a 5x5 kernel

```
In [9]: # 5
# TODO
kernel = np.ones((5, 5), np.uint8)
erodedJ = cv2.erode(img, kernel, iterations = 1)
plt.imshow(erodedJ)
```

Out[9]: <matplotlib.image.AxesImage at 0xf11735aa5de0>





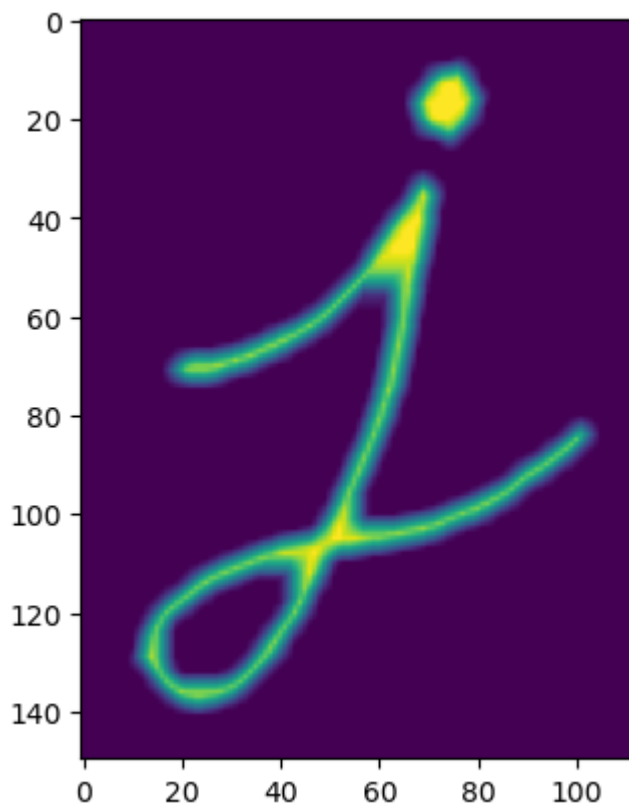
## 6. Perform erosion on j with two iterations, using a kernel size of your choice

Hint: look at the OpenCV API documentation. It is possible to perform two iterations of erosion in one line of Python!

[https://docs.opencv.org/3.4.1/d4/d86/group\\_imgproc\\_filter.html#gaeb1e0c1033e3f6b891a25d0511362aeb](https://docs.opencv.org/3.4.1/d4/d86/group_imgproc_filter.html#gaeb1e0c1033e3f6b891a25d0511362aeb)

```
In [10]: # 6
# TODO
kernel = np.ones((3, 3), np.uint8)
erodedJ = cv2.erode(img, kernel, iterations = 2)
plt.imshow(erodedJ)
```

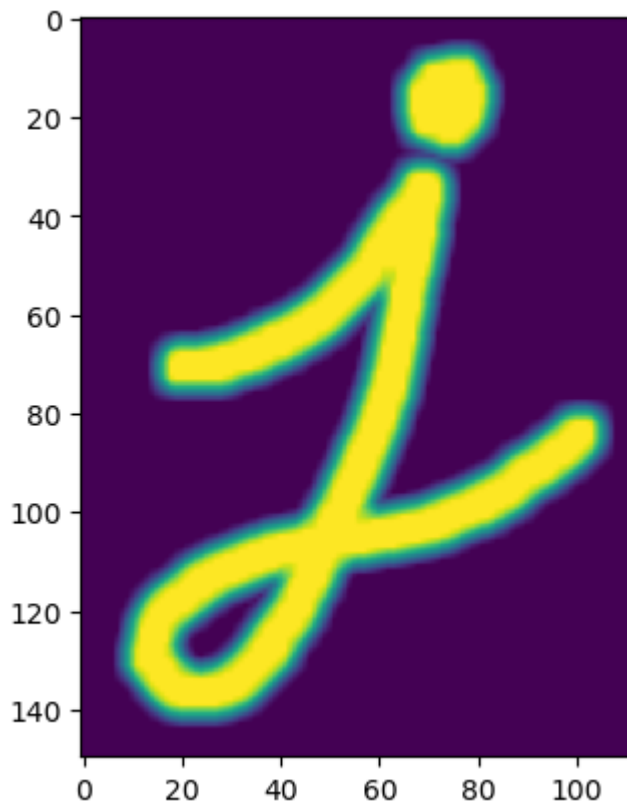
```
Out[10]: <matplotlib.image.AxesImage at 0xf11735b11720>
```



## 7. Perform dilation on j with a 3x3 kernel

```
In [11]: # 7
# TODO
kernel = np.ones((3, 3), np.uint8)
erodedJ = cv2.dilate(img, kernel, iterations = 1)
plt.imshow(erodedJ)
```

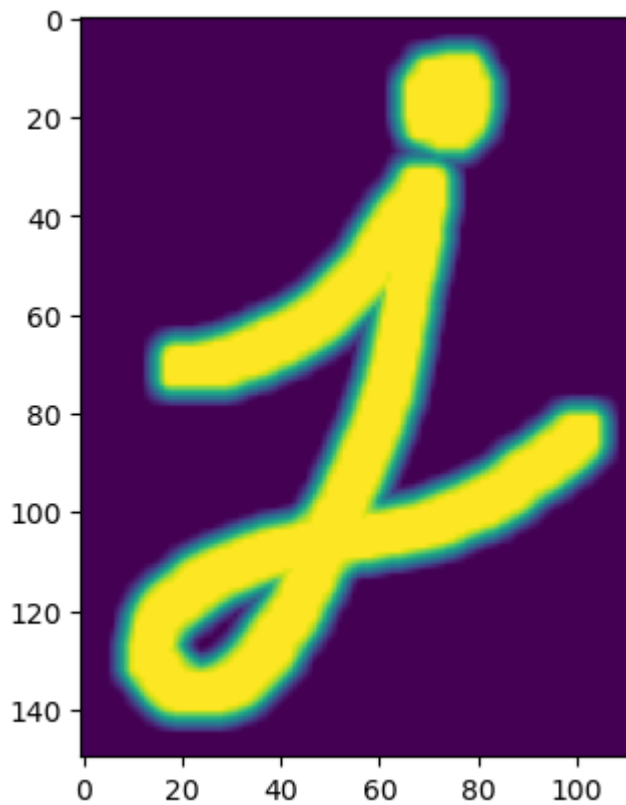
```
Out[11]: <matplotlib.image.AxesImage at 0xf1173597d150>
```



## 8. Perform dilation on j with a 5x5 kernel

```
In [12]: # 8
# TODO
kernel = np.ones((5, 5), np.uint8)
erodedJ = cv2.dilate(img, kernel, iterations = 1)
plt.imshow(erodedJ)
```

```
Out[12]: <matplotlib.image.AxesImage at 0xf117359f0a60>
```



## 9. What is the effect of kernel size on morphology operations?

```
In [13]: # 9
# TODO
# The larger the kernel size, the greater the effect of the morphology op
```

## 10. What is the difference between repeated iterations of a morphology operation with a small kernel, versus a single iteration with a large kernel?

```
In [14]: # 10
# TODO
# Repeated iterations with a small kernel would tend to produce gradual a
```

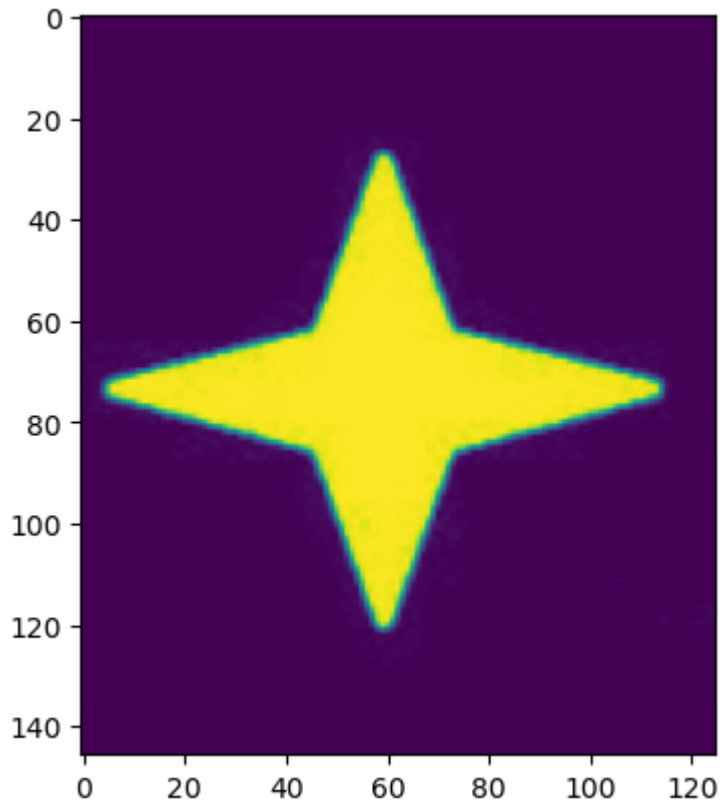
## 11. Rotate the lightningbolt and star by 90 degrees

[https://docs.opencv.org/3.4.1/da/d6e/tutorial\\_py\\_geometric\\_transformations.html](https://docs.opencv.org/3.4.1/da/d6e/tutorial_py_geometric_transformations.html)

```
In [15]: # 11
# TODO
rows2,cols2 = star.shape
M2 = cv2.getRotationMatrix2D((cols2/2,rows2/2),90,1)
rotatedstar = cv2.warpAffine(star,M2,(cols2,rows2))
```

```
plt.imshow(rotatedstar)
```

Out[15]: <matplotlib.image.AxesImage at 0xf11735a5d600>

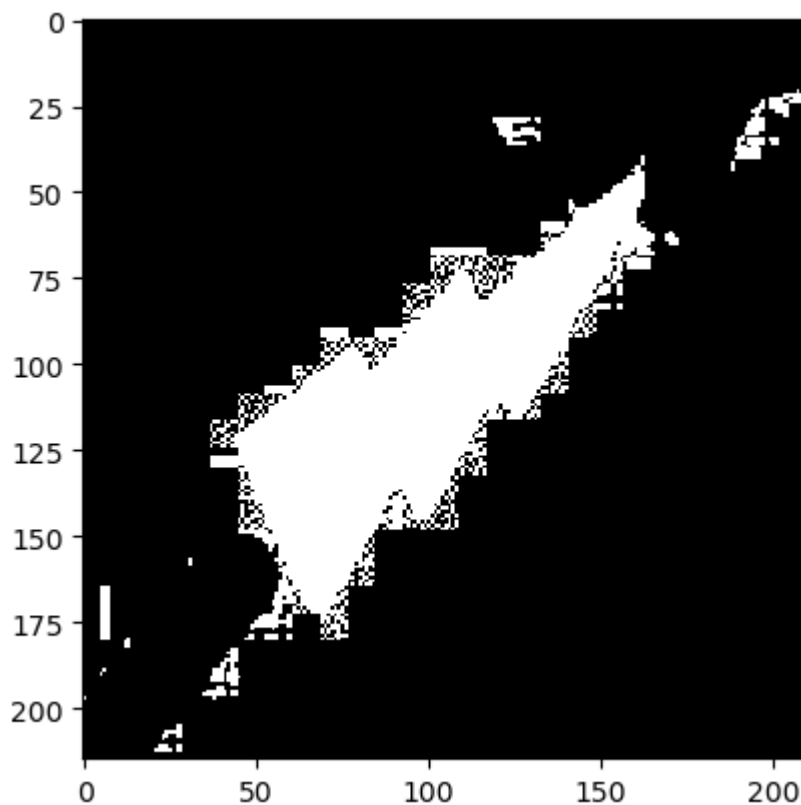


```
In [16]: rows, cols = lightningbolt.shape

# Create a rotation matrix
M = cv2.getRotationMatrix2D((cols / 2, rows / 2), 90, 1)

# Apply the affine transformation with cubic interpolation
rotatedbolt = cv2.warpAffine(lightningbolt, M, (cols, rows), flags=cv2.IN

# Display the rotated image
fig, ax = plt.subplots()
ax.imshow(rotatedbolt, cmap='gray', interpolation='none')
plt.show()
```



## 12. STRETCH GOAL:

Visualize the result of Laplacian, Sobel X, and Sobel Y on all of the images. Also, produce a combined image of both Sobel X and Sobel Y for each image. Is Exercise 1 the best way to do this? Are there other options?

You should have 4 outputs (Laplacian, SobelX, SobelY, and the combination) for each input image visualized at the end.

[https://docs.opencv.org/3.4.1/d5/d0f/tutorial\\_py\\_gradients.html](https://docs.opencv.org/3.4.1/d5/d0f/tutorial_py_gradients.html)

## When you are done:

You should have one or more images for each exercise.

1. Double-check that you filled in your name at the top of the notebook!
2. Click File -> Export Notebook As -> PDF
3. Email the PDF to YOURTEAMNAME@beaver.works