

```
In [2]: from __future__ import print_function
%matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import cv2
from IPython.display import HTML, YouTubeVideo
import matplotlib.patches as patches
from matplotlib.lines import Line2D
#import ganymede
#ganymede.configure('uav.beaver.works')
```

Enter your name below and run the cell:Joy Deng

Individual cells can be run with Ctrl + Enter

```
In [3]: #ganymede.name('YOUR NAME HERE')
#def check(p):
#    ganymede.update(p,True)
#check(0)
```

<https://www.khanacademy.org/math/statistics-probability/describing-relationships-quantitative-data/more-on-regression/v/squared-error-of-regression-line>

Note: All Khan Academy content is available for free at [khanacademy.org](https://www.khanacademy.org)

```
In [4]: YouTubeVideo('60vhLPS7rj4', width=560, height=315)
```

Out[4]:

Squared error of regression line | Regression | Probability a...



```
In [5]: YouTubeVideo('mIx20j5y9Q8', width=560, height=315)
```

Out[5]:

Proof (part 1) minimizing squared error to regression line | ...



```
In [6]: YouTubeVideo('f60noxctvUk', width=560, height=315)
```

Out[6]:

Proof (part 2) minimizing squared error to regression line | ...



```
In [7]: YouTubeVideo('u1HhUB3NP8g', width=560, height=315)
```

Out[7]:

Proof (part 3) minimizing squared error to regression line | ...

In [8]: `YouTubeVideo('8RSTQl0bQuw', width=560, height=315)`

Out[8]:

Proof (part 4) minimizing squared error to regression line | ...

In [9]: `YouTubeVideo('GAmzwIkGFgE', width=560, height=315)`

Out[9]:

Regression line example | Regression | Probability and Stati...



The last video is optional

```
In [10]: YouTubeVideo('ww_yT9ckPWw', width=560, height=315)
```

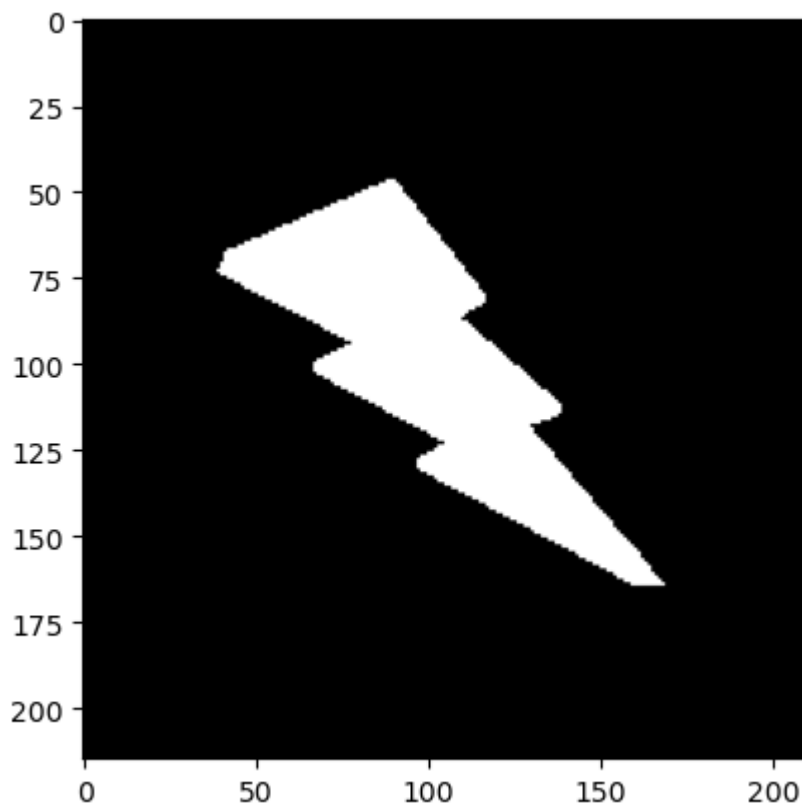
Out[10]:

Second regression example | Regression | Probability and S...



```
In [11]: lightningbolt = cv2.imread('shapes/lightningbolt.png', cv2.IMREAD_GRAYSCA
_, lightningbolt = cv2.threshold(lightningbolt, 150, 255, cv2.THRESH_BINARY)
print(lightningbolt.shape)
fig, ax = plt.subplots()
ax.imshow(lightningbolt, cmap='gray');
```

(215, 209)



```
In [12]: np.argwhere?
```

Signature: `np.argwhere(a)`

Docstring:

Find the indices of array elements that are non-zero, grouped by element.

Parameters

a : array_like
Input data.

Returns

index_array : (N, a.ndim) ndarray
Indices of elements that are non-zero. Indices are grouped by element.
This array will have shape ``(N, a.ndim)`` where ``N`` is the number of
non-zero items.

See Also

`where`, `nonzero`

Notes

``np.argwhere(a)`` is almost the same as ``np.transpose(np.nonzero(a))``, but produces a result of the correct shape for a 0D array.

The output of ``argwhere`` is not suitable for indexing arrays.
For this purpose use ``nonzero(a)`` instead.

Examples

```
>>> x = np.arange(6).reshape(2,3)
```

```
>>> x
```

```
array([[0, 1, 2],
       [3, 4, 5]])
```

```
>>> np.argwhere(x>1)
```

```
array([[0, 2],
       [1, 0],
       [1, 1],
       [1, 2]])
```

File: /usr/lib/python3/dist-packages/numpy/core/numeric.py

Type: function

```
In [13]: bolt = np.argwhere(lightningbolt)
```

Linear Regression

$$m = \frac{\bar{x}\bar{y} - \overline{xy}}{(\bar{x})^2 - \overline{x^2}}$$

$$b = \bar{y} - m\bar{x}$$

Question: how can we extract the xs and ys separately from the result of argwhere?

Hint: review numpy slicing by columns and rows

```
In [14]: # slice the individual
# Your answer here
```

Question: Why would we want to convert x and y points from int values to floats?

```
In [15]: # so they are easy to manipulate and take the mean of when finding m and
```

```
In [16]: import numpy as np
def calculate_regression(points): # input is the result of np.argwhere
    points = points.astype(float)
    xs = points[:,0] # slice along the vertical columns
    ys = points[:,1]
    x_mean = np.mean(xs)
    y_mean = np.mean(ys)

    xy_mean = np.mean(xs*ys)

    x_squared_mean = np.mean(xs**2)

    m = (x_mean * y_mean - xy_mean)/(x_mean**2 - x_squared_mean)

    b = y_mean - m*x_mean

    return (m,b)
#cv.morphopen
```

```
In [17]: import numpy as np
a = np.array([[ 0.14022471,  0.96360618], #random
              [ 0.37601032,  0.25528411], #random
              [ 0.49313049,  0.94909878]]) #random
print(a.shape)
```

(3, 2)

The intercept we calculated, b , may be outside of the pixel space of the image, so we must find two points inside of pixel space, (x_1, y_1) and (x_2, y_2) which will allow us to plot our regression line on the image. It may be best to choose points on the regression line which also occur on the boundaries/extrema of the image.

```
In [18]: def find_inliers(m, b, shape):
# find intersection of the box
y = float(shape[0])
x = float(shape[1])
xs=[0,shape[0],-b/m,(shape[1]-b)/m]

if m != 0:
    xs.sort()
    x_1=xs[1]
    x_2=xs[2]
    y_1=m*x_1+b
    y_2=m*x_2+b
    return ([x_1,x_2],[y_1,y_2])
```

```
m,b = calculate_regression(np.argwhere(star))
print (m,b)
_ = find_inliers(m,b, star.shape)
print(find_inliers(m,b, star.shape))
```

NameError

Traceback (most recent call last)

Cell In[18], line 15

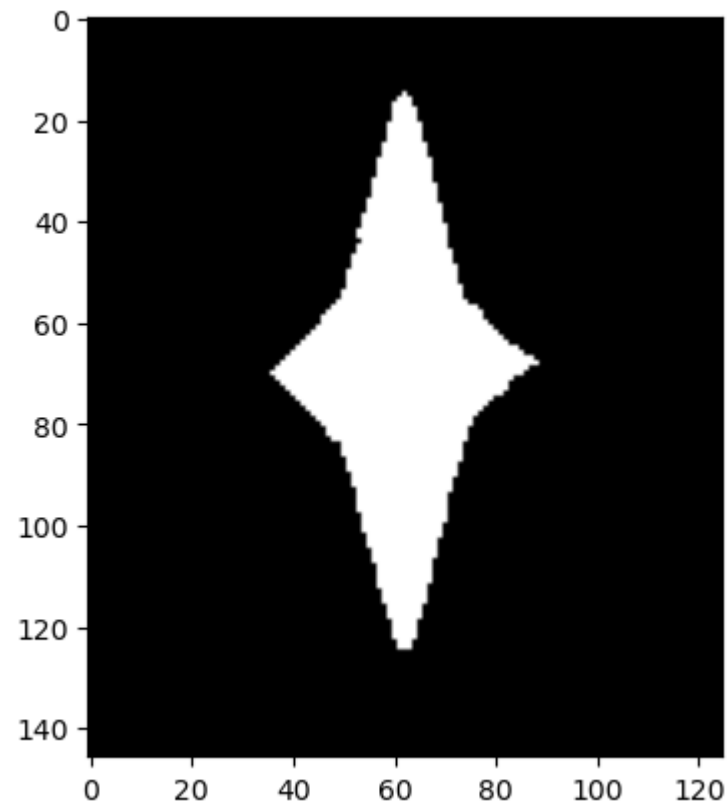
```
12         y_2=m*x_2+b
13         return([x_1,x_2],[y_1,y_2])
--> 15 m,b = calculate_regression(np.argwhere(star))
16 print (m,b)
17 _ = find_inliers(m,b, star.shape)
```

NameError: name 'star' is not defined

```
In [19]: import cv2
star = cv2.imread('shapes/squishedstar.png', cv2.IMREAD_GRAYSCALE)
print(star.shape)

_, star = cv2.threshold(star,125,255,cv2.THRESH_BINARY)
fig,ax = plt.subplots()
ax.imshow(star, cmap='gray');
```

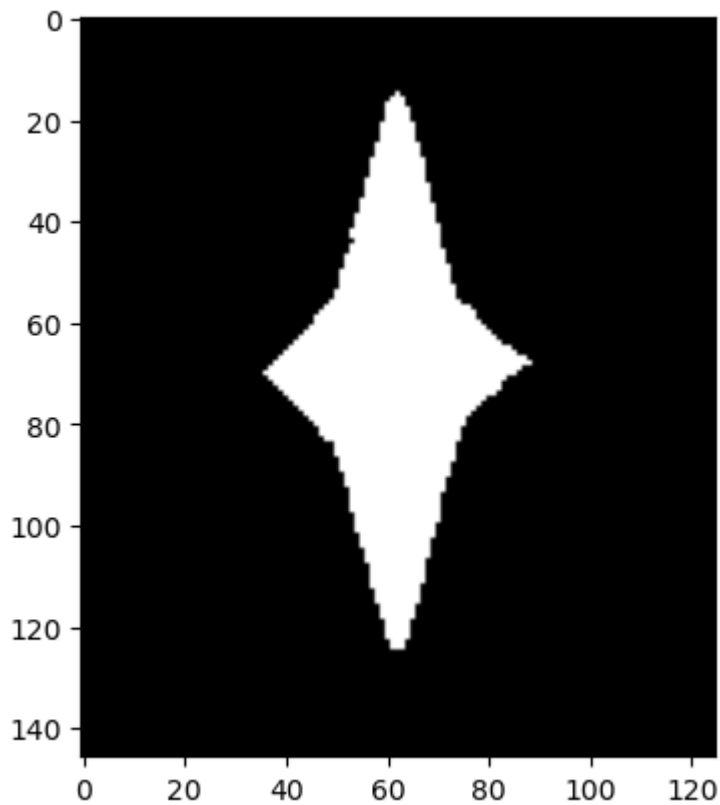
(146, 125)



```
In [20]: import cv2
star = cv2.imread('shapes/squishedstar.png', cv2.IMREAD_GRAYSCALE)
print(star.shape)

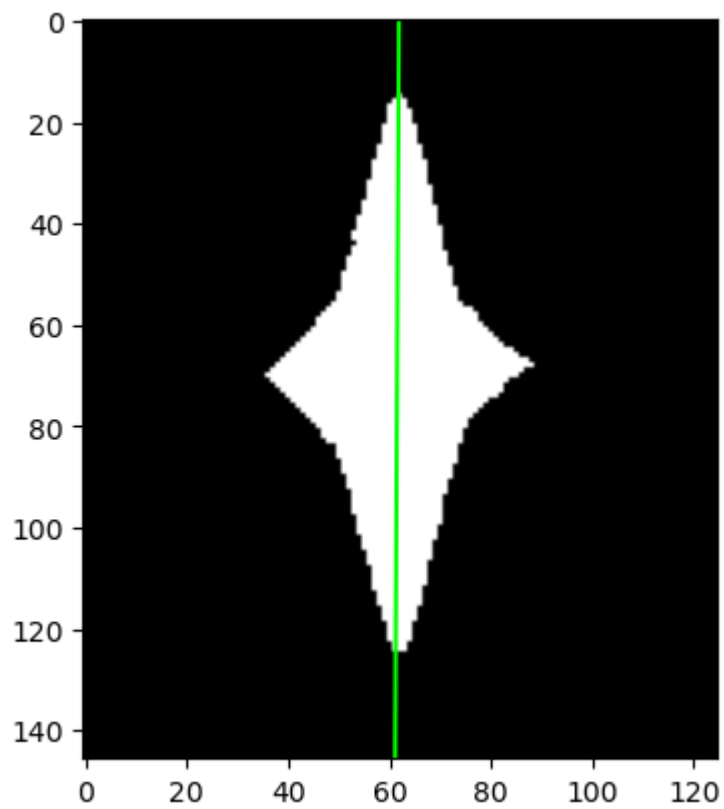
_, star = cv2.threshold(star,125,255,cv2.THRESH_BINARY)
fig,ax = plt.subplots()
ax.imshow(star, cmap='gray');
```


(146, 125)



```
In [89]: m,b = calculate_regression(np.argwhere(star))
         _ = find_inliers(m,b, star.shape)
```

```
In [99]: # below is an example of how to draw a random line from (10,25) to (10,55)
         # TODO: replace this with the result of find_inliers
         # -- pay attention to the directions of the x and y axes
         #    in image space, row-column space, and cartesian space
         # Look at the help function for Line2D below
         #((None, 146.0), (125.0, 66.81778197976166), (None, 0), (59.6795206617331
fig,ax = plt.subplots()
ax.imshow(star, cmap='gray');
mask=np.argwhere(star)
m,b=calculate_regression(mask)
pts=find_inliers(m,b, star.shape)
regression = Line2D(pts[1],pts[0],color='lime')
ax.add_line(regression);
```



In [91]: `Line2D?`

Init signature:

```

Line2D(
    xdata,
    ydata,
    linewidth=None,
    linestyle=None,
    color=None,
    marker=None,
    markersize=None,
    markeredgewidth=None,
    markeredgecolor=None,
    markerfacecolor=None,
    markerfacecoloralt='none',
    fillstyle=None,
    antialiased=None,
    dash_capstyle=None,
    solid_capstyle=None,
    dash_joinstyle=None,
    solid_joinstyle=None,
    pickradius=5,
    drawstyle=None,
    markevery=None,
    **kwargs,
)

```

Docstring:

A line - the line can have both a solid linestyle connecting all the vertices, and a marker at each vertex. Additionally, the drawing of the solid line is influenced by the drawstyle, e.g., one can create "stepped" lines in various styles.

Init docstring:

Create a `Line2D` instance with `*x*` and `*y*` data in sequences of `*xdata*`, `*ydata*`.

Additional keyword arguments are `Line2D` properties:

Properties:

`agg_filter`: a filter function, which takes a (m, n, 3) float array and a dpi value, and returns a (m, n, 3) array

`alpha`: scalar or None

`animated`: bool

`antialiased` or `aa`: bool

`clip_box`: `Bbox`

`clip_on`: bool

`clip_path`: Patch or (Path, Transform) or None

`color` or `c`: color

`dash_capstyle`: `CapStyle` or {'butt', 'projecting', 'round'}

`dash_joinstyle`: `JoinStyle` or {'miter', 'round', 'bevel'}

`dashes`: sequence of floats (on/off ink in points) or (None, None)

`data`: (2, N) array or two 1D arrays

`drawstyle` or `ds`: {'default', 'steps', 'steps-pre', 'steps-mid', 'steps-post'}, default: 'default'

`figure`: `Figure`

`fillstyle`: {'full', 'left', 'right', 'bottom', 'top', 'none'}

`gid`: str

`in_layout`: bool

`label`: object

`linestyle` or `ls`: {'-', '--', '-.', ':', '', (offset, on-off-seq), ...}

`linewidth` or `lw`: float

`marker`: marker style string, `~.path.Path` or `~.markers.MarkerStyle`

`markeredgecolor` or `mec`: color

```

    markeredgewidth or mew: float
    markerfacecolor or mfc: color
    markerfacecoloralt or mfcalt: color
    markersize or ms: float
    markevery: None or int or (int, int) or slice or list[int] or float or
(float, float) or list[bool]
    path_effects: `.AbstractPathEffect`
    picker: float or callable[[Artist, Event], tuple[bool, dict]]
    pickradius: float
    rasterized: bool
    sketch_params: (scale: float, length: float, randomness: float)
    snap: bool or None
    solid_capstyle: `.CapStyle` or {'butt', 'projecting', 'round'}
    solid_joinstyle: `.JoinStyle` or {'miter', 'round', 'bevel'}
    transform: unknown
    url: str
    visible: bool
    xdata: 1D array
    ydata: 1D array
    zorder: float

```

See :meth:`set_linestyle` for a description of the line styles,
:meth:`set_marker` for a description of the markers, and
:meth:`set_drawstyle` for a description of the draw styles.

```

File:          /usr/lib/python3/dist-packages/matplotlib/lines.py
Type:          type
Subclasses:    _AxiLine, Line3D

```

TODO

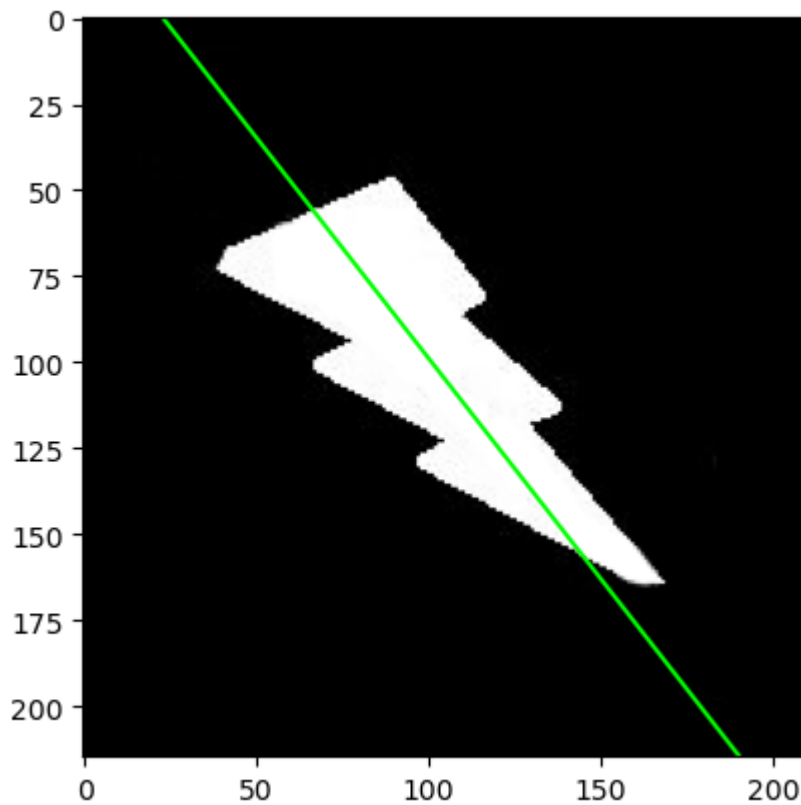
1. Run your linear regression algorithm on the following images.
2. Plot each of the results.
3. Include each result in your submitted PDF.

```

In [100... lightningbolt      = cv2.imread('shapes/lightningbolt.png', cv2.IMREAD_GR

fig,ax = plt.subplots()
ax.imshow(lightningbolt, cmap='gray');
mask=np.argwhere(lightningbolt)
m,b=calculate_regression(mask)
pts=find_inliers(m,b, lightningbolt.shape)
regression = Line2D(pts[1],pts[0],color='lime')
ax.add_line(regression);

```

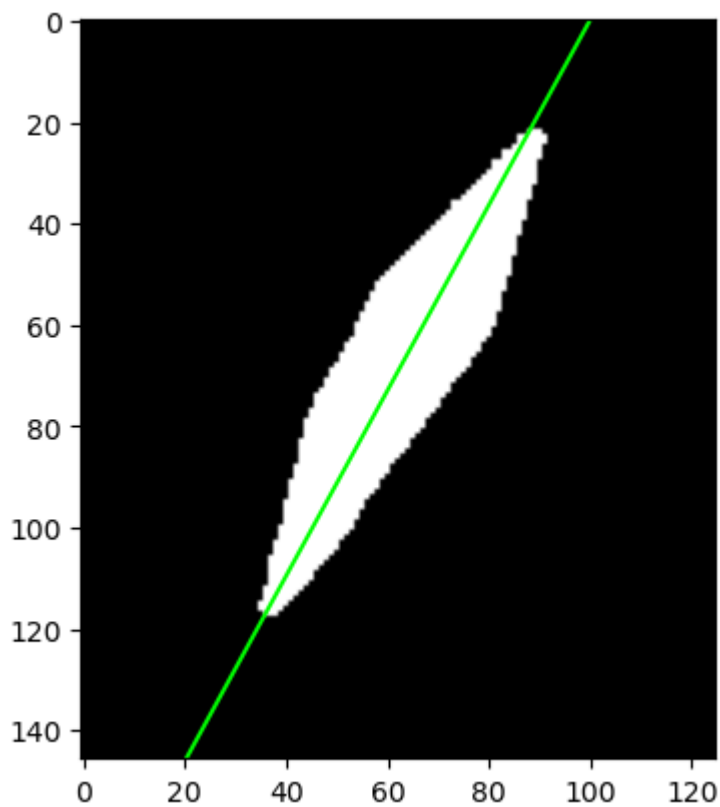


```
In [21]: blob = cv2.imread('shapes/blob.png', cv2.IMREAD_GRAYSCALE)
_, blob = cv2.threshold(blob, 125, 255, cv2.THRESH_BINARY)
fig, ax = plt.subplots()
ax.imshow(blob, cmap='gray');
mask = np.argwhere(blob)
m, b = calculate_regression(mask)
pts = find_inliers(m, b, blob.shape)
regression = Line2D(pts[1], pts[0], color='lime')
ax.add_line(regression);

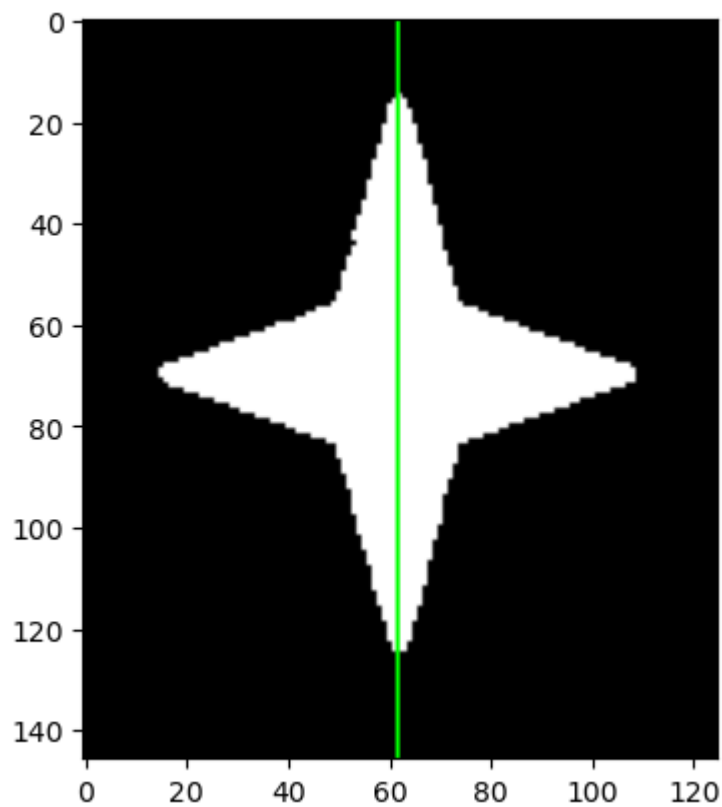
images = [lightningbolt, blob, star, squishedstar, squishedturnedstar, le
```

```
-----
-
NameError                                Traceback (most recent call las
t)
Cell In[21], line 11
      8 regression = Line2D(pts[1], pts[0], color='lime')
      9 ax.add_line(regression);
--> 11 images = [lightningbolt, blob, star, squishedstar, squishedturnedst
ar, letterj]

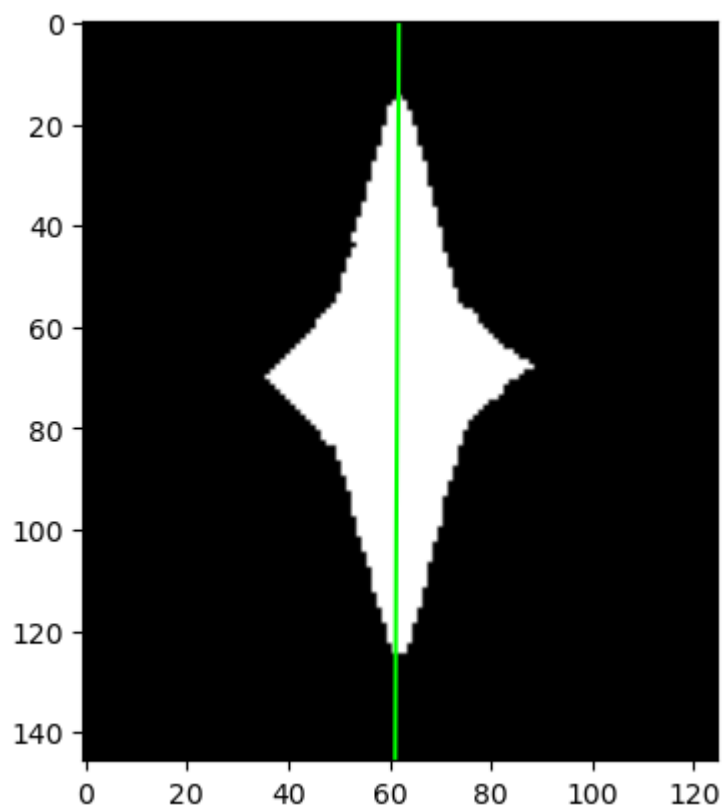
NameError: name 'squishedstar' is not defined
```



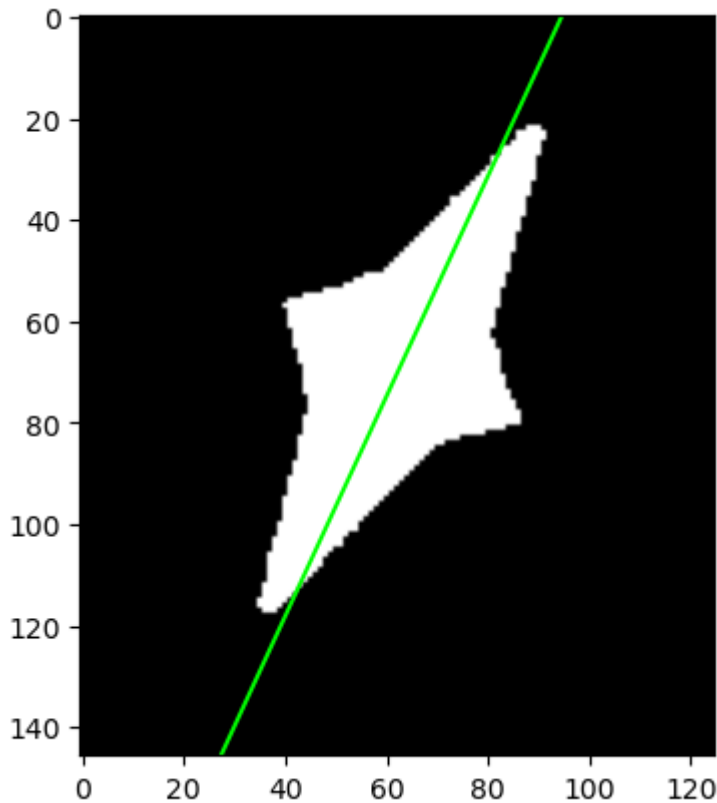
```
In [26]: star = cv2.imread('shapes/star.png', cv2.IMREAD_GRAYSCALE)
_, star = cv2.threshold(star, 125, 255, cv2.THRESH_BINARY)
fig, ax = plt.subplots()
ax.imshow(star, cmap='gray');
mask = np.argwhere(star)
m, b = calculate_regression(mask)
pts = find_inliers(m, b, star.shape)
regression = Line2D(pts[1], pts[0], color='lime')
ax.add_line(regression);
```



```
In [25]: squishedstar = cv2.imread('shapes/squishedstar.png', cv2.IMREAD_GRAYSCALE)
squishedstar = cv2.threshold(squishedstar, 125, 255, cv2.THRESH_BINARY)
fig, ax = plt.subplots()
ax.imshow(squishedstar, cmap='gray');
mask = np.argwhere(squishedstar)
m, b = calculate_regression(mask)
pts = find_inliers(m, b, squishedstar.shape)
regression = Line2D(pts[1], pts[0], color='lime')
ax.add_line(regression);
```



```
In [27]: import cv2
squishedturnedstar = cv2.imread('shapes/squishedturnedstar.png', cv2.IMRE
_, squishedturnedstar = cv2.threshold(squishedturnedstar,125,255,cv2.THRE
fig,ax = plt.subplots()
ax.imshow(squishedturnedstar, cmap='gray');
mask=np.argwhere(squishedturnedstar)
m,b=calculate_regression(mask)
pts=find_inliers(m,b, squishedturnedstar.shape)
regression = Line2D(pts[1],pts[0],color='lime')
ax.add_line(regression);
```



```
In [74]: _, squishedturnedstar = cv2.threshold(squishedturnedstar,125,255,cv2.THRE
fig,ax = plt.subplots()
ax.imshow(squishedturnedstar, cmap='gray');
mask=np.argwhere(squishedturnedstar)
m,b=calculate_regression(mask)
pts=find_inliers(m,b, squishedturnedstar.shape)
regression = Line2D(pts[1],pts[0],color='lime')
ax.add_line(regression);
images = [lightningbolt, blob, star, squishedstar, squishedturnedstar, le
```

When you are done:

You should have six images with regression lines plotted on top of them.

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

Stretch goal

Implement a machine learning algorithm!

Random Sample Consensus, commonly referred to as *RANSAC*, is one of the most widely used machine learning algorithms. In essence, it is a 'guess and check' algorithm. Take a small random sample of your data - two points in this case. Next, define a line through those two points. After doing so, count the number of *inliers*, or points closest to that line (euclidean distance is one way to do this).

https://en.wikipedia.org/wiki/Random_sample_consensus

Implement RANSAC for linear regression, and run it on all of your images.