

Q3

June 1, 2020

```
[1]: %%javascript
      IPython.OutputArea.prototype._should_scroll = function(lines) {
          return false;
      }
```

<IPython.core.display.Javascript object>

```
[2]: import numpy as np
      import matplotlib.pyplot as plt
      import seaborn as sns
      import cv2
      from IPython.core.display import display, HTML
      display(HTML("<style>.container { width:100% !important; }</style>"))
```

<IPython.core.display.HTML object>

0.0.1 1.) Image Resize to 10, 30, 50%

```
[3]: img = plt.imread('Quiz.png')
```

```
[4]: def resize(img, percent):
      h = int(img.shape[0] * percent)
      w = int(img.shape[1] * percent)

      return cv2.resize(img, (w,h))
```

```
[5]: img10 = resize(img, 0.1)
      img30 = resize(img, 0.3)
      img50 = resize(img, 0.5)
```

```
[6]: plt.subplots(nrows=2,ncols=2, figsize=(20,15))

      plt.subplot(221)
```

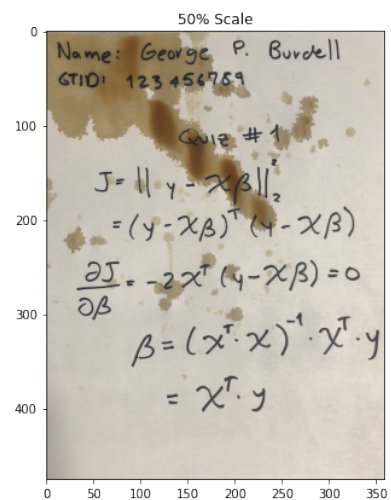
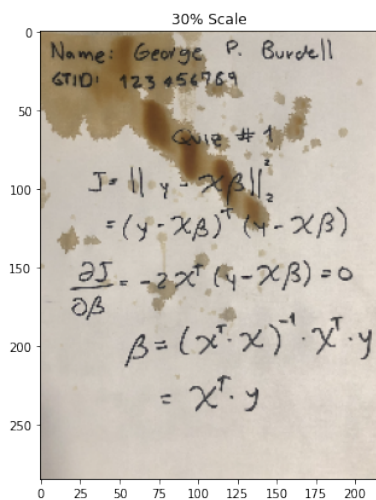
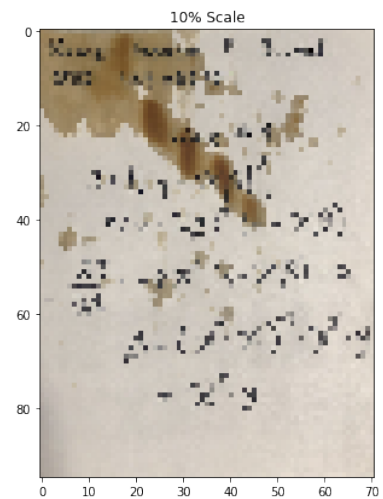
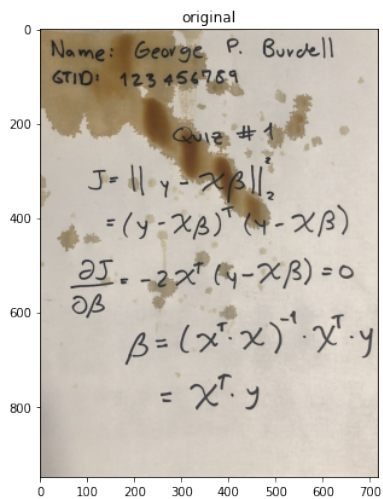
```
plt.imshow(img)
plt.title('original')

plt.subplot(222)
plt.imshow(img10)
plt.title('10% Scale')

plt.subplot(223)
plt.imshow(img30)
plt.title('30% Scale')

plt.subplot(224)
plt.imshow(img50)
plt.title('50% Scale')

plt.show()
```



0.0.2 2.) Convert to grayscale

```
[7]: gray = (cv2.cvtColor(img, cv2.COLOR_RGB2GRAY) * 255).astype('int32')
gray10 = (cv2.cvtColor(img10, cv2.COLOR_RGB2GRAY) * 255).astype('int32')
gray30 = (cv2.cvtColor(img30, cv2.COLOR_RGB2GRAY) * 255).astype('int32')
gray50 = (cv2.cvtColor(img50, cv2.COLOR_RGB2GRAY) * 255).astype('int32')
```

```
[8]: plt.subplots(nrows=4,ncols=2, figsize=(30,50))

plt.subplot(421)
plt.imshow(gray, cmap='gray', vmin=0, vmax=255)
plt.title('original')

plt.subplot(422)
sns.distplot(gray.reshape(-1),kde=False,hist_kws=dict(edgecolor="k",
↳linewidth=2))

plt.subplot(423)
plt.imshow(gray50, cmap='gray', vmin=0, vmax=255)
plt.title('50% Scale')

plt.subplot(424)
sns.distplot(gray50.reshape(-1),kde=False,hist_kws=dict(edgecolor="k",
↳linewidth=2))

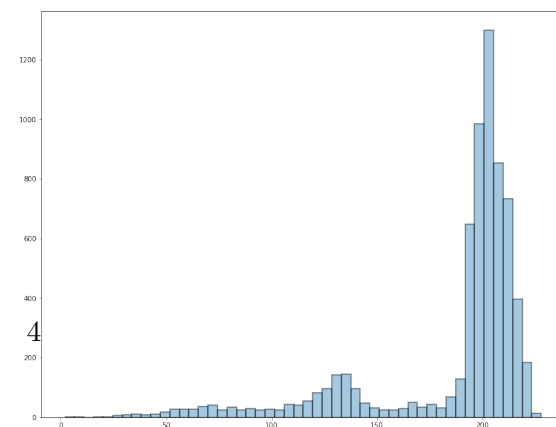
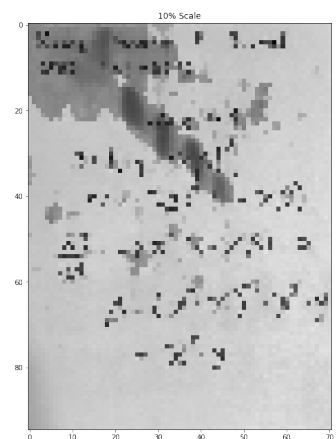
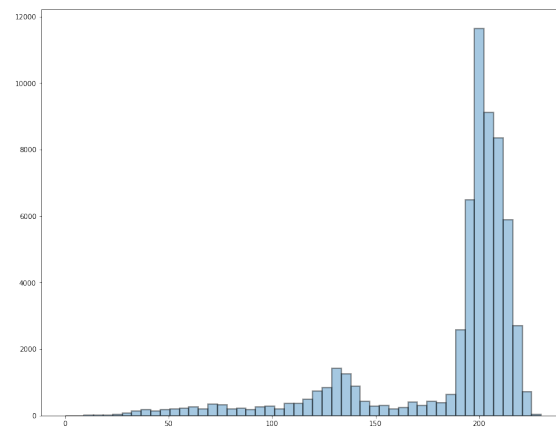
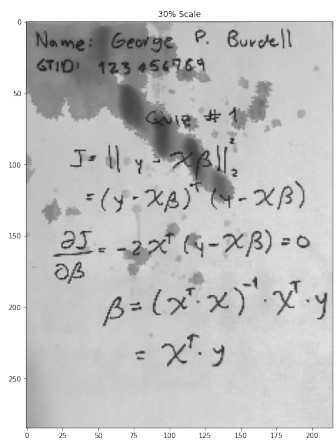
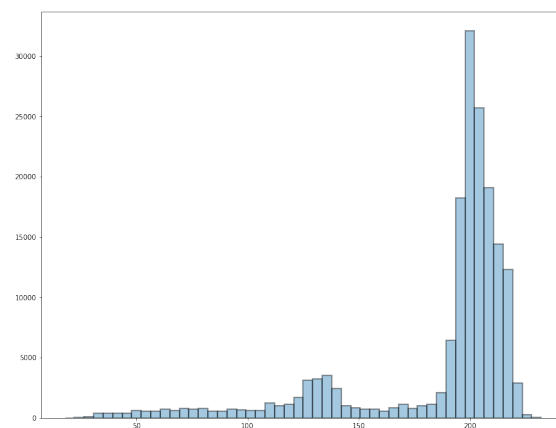
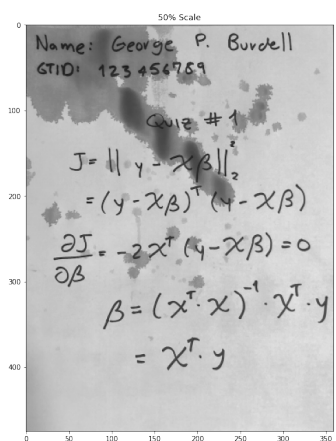
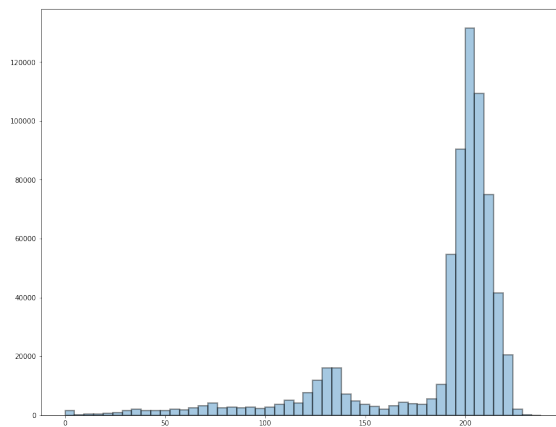
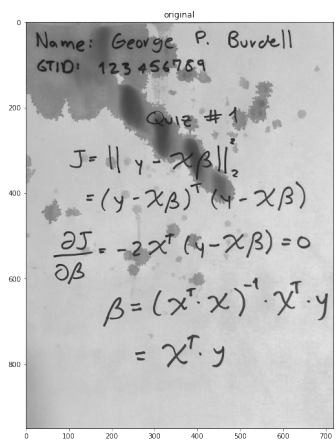
plt.subplot(425)
plt.imshow(gray30, cmap='gray', vmin=0, vmax=255)
plt.title('30% Scale')

plt.subplot(426)
sns.distplot(gray30.reshape(-1),kde=False, hist_kws=dict(edgecolor="k",
↳linewidth=2))

plt.subplot(427)
plt.imshow(gray10, cmap='gray', vmin=0, vmax=255)
plt.title('10% Scale')

plt.subplot(428)
sns.distplot(gray10.reshape(-1),kde=False,hist_kws=dict(edgecolor="k",
↳linewidth=2))

plt.show()
```



0.0.3 Image Selection

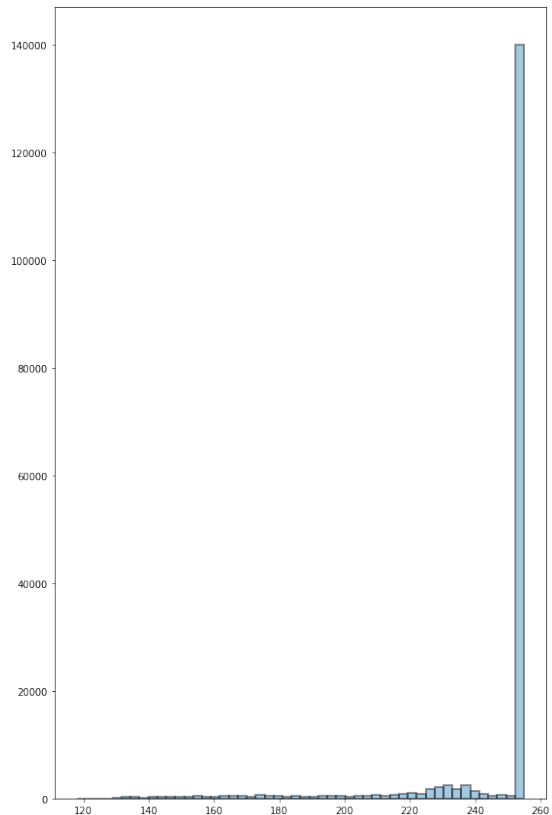
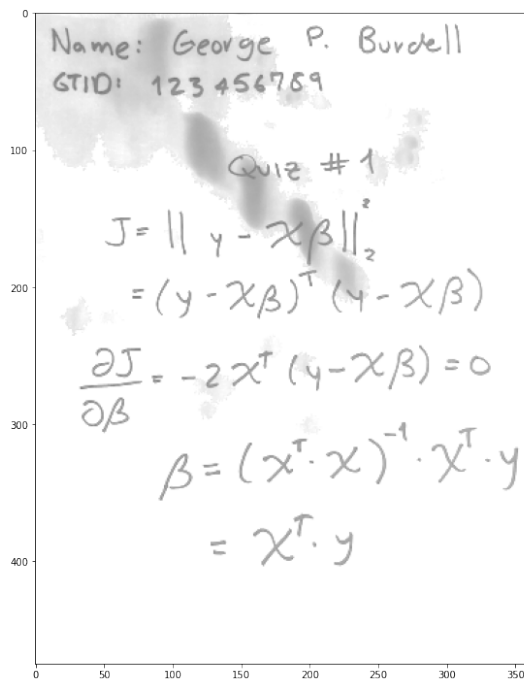
The 50% image looks to be the best candidate. It reduces the original image from > 6 million pixels to approx 1.7 million pixels while maintaining sufficient resolution. The 30% and 10% appear to be too grainy, and probably will begin to lose informational structure as transformations are later applied

0.0.4 3.) Pixel Shifting

```
[9]: def pixelshift(image, s):  
      img = image.copy()  
      img += s  
      img[img>255] = 255  
      img[img<0] = 0  
      return img
```

```
[10]: shift = pixelshift(gray50, 100)
```

```
[11]: plt.subplots(nrows=1, ncols=2, figsize=(20,15))  
      plt.subplot(121)  
      plt.imshow(shift, cmap='gray', vmin=0, vmax=255)  
  
      plt.subplot(122)  
      sns.distplot(shift.reshape(-1), kde=False, hist_kws=dict(edgecolor="k",  
      ↪ linewidth=2))  
  
      plt.show()
```



0.0.5 4.) Histogram Stretching

```
[12]: def stretch(image):

    img = image.copy()
    pmin = np.min(img)

    #max pixel intensity
    pmax = np.max(img)

    #lambda
    lam = 255

    #hist stretching function
    s = lambda x: ((x - pmin) / (pmax - pmin)) * lam

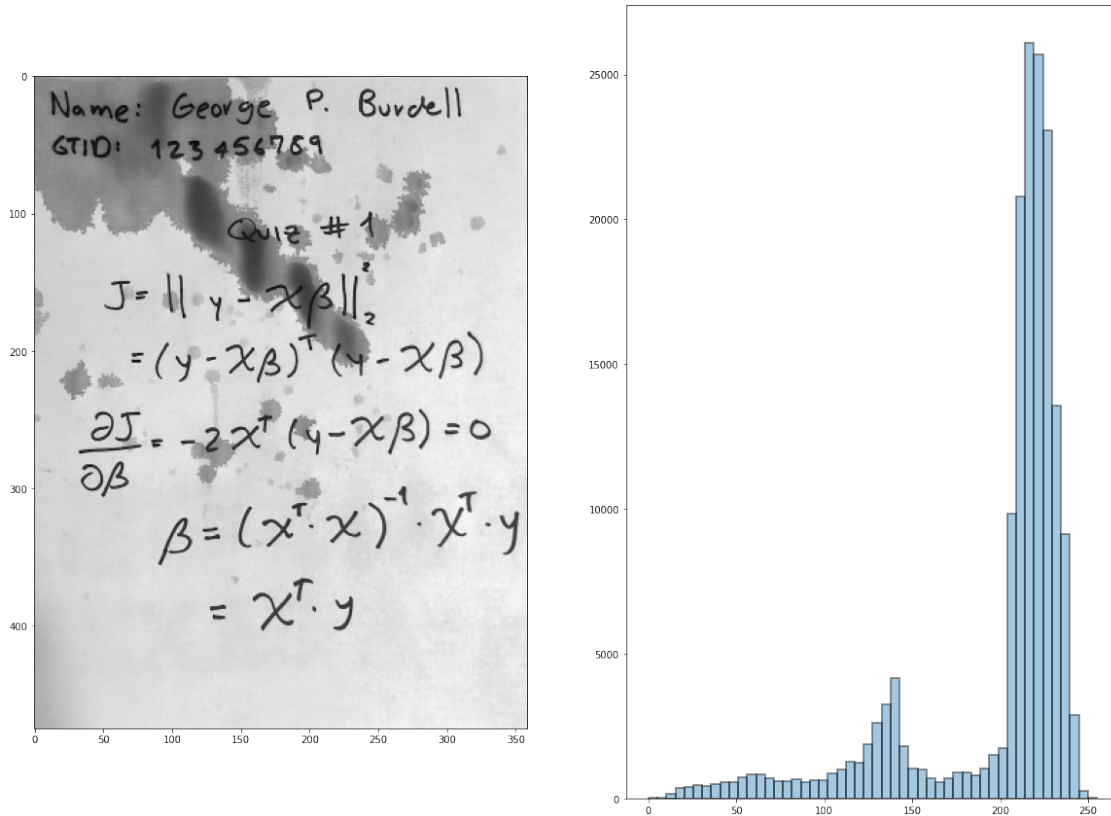
    return s(img).astype(int)
```

```
[13]: img_stretch = stretch(gray50)
```

```
[14]: plt.subplots(nrows=1, ncols=2, figsize=(20,15))
plt.subplot(121)
plt.imshow(img_stretch, cmap='gray', vmin=0, vmax=255)

plt.subplot(122)
sns.distplot(img_stretch.reshape(-1), kde=False, hist_kws=dict(edgecolor="k",
↪ linewidth=2))

plt.show()
```



0.0.6 5.) Linear (negative image)

```
[15]: def linear(image):
img = image.copy()
lin = lambda x: (256 - 1) - x

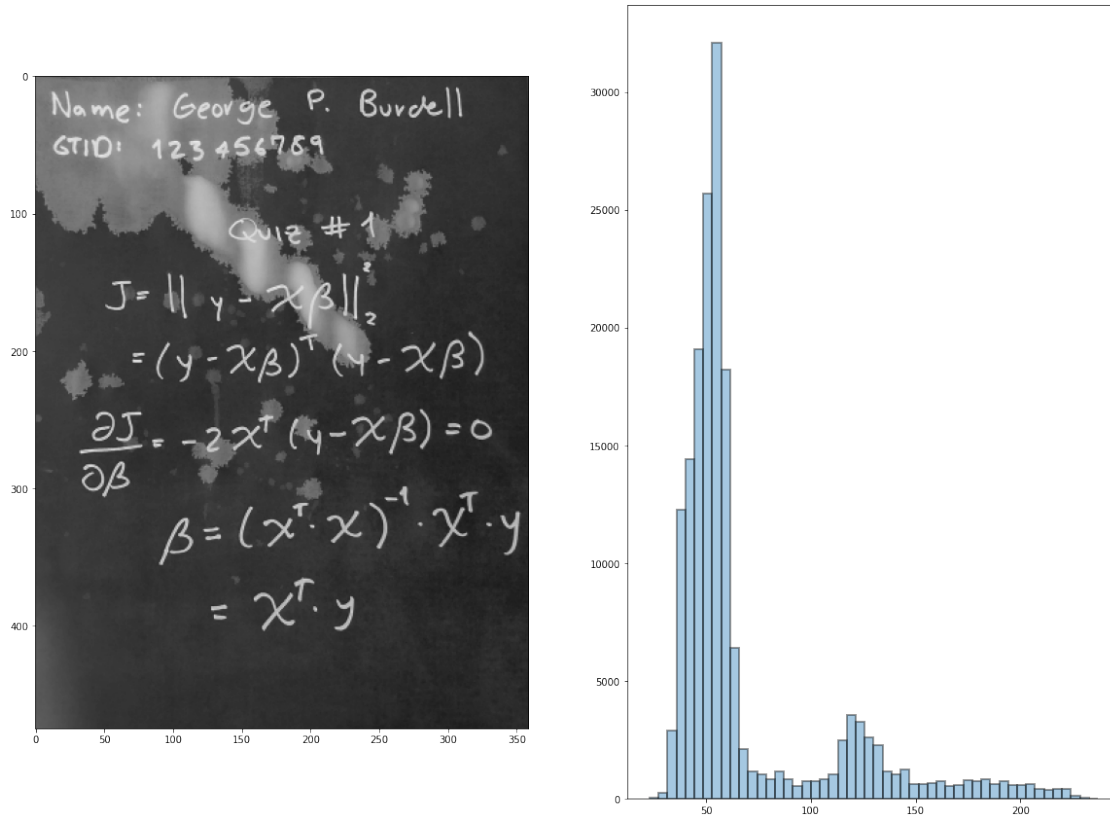
return lin(img).astype(int)
```

```
[16]: lin_img = linear(gray50)
```

```
[17]: plt.subplots(nrows=1, ncols=2, figsize=(20,15))
plt.subplot(121)
plt.imshow(lin_img, cmap='gray', vmin=0, vmax=255)

plt.subplot(122)
sns.distplot(lin_img.reshape(-1), kde=False, hist_kws=dict(edgecolor="k",
↪ linewidth=2))

plt.show()
```



0.0.7 6.) logarithmic transformation

```
[18]: def log(image):
    img = image.copy()
    c=45

    logpix = lambda x: c * np.log(x + 1)

    return logpix(img).astype(int)
```

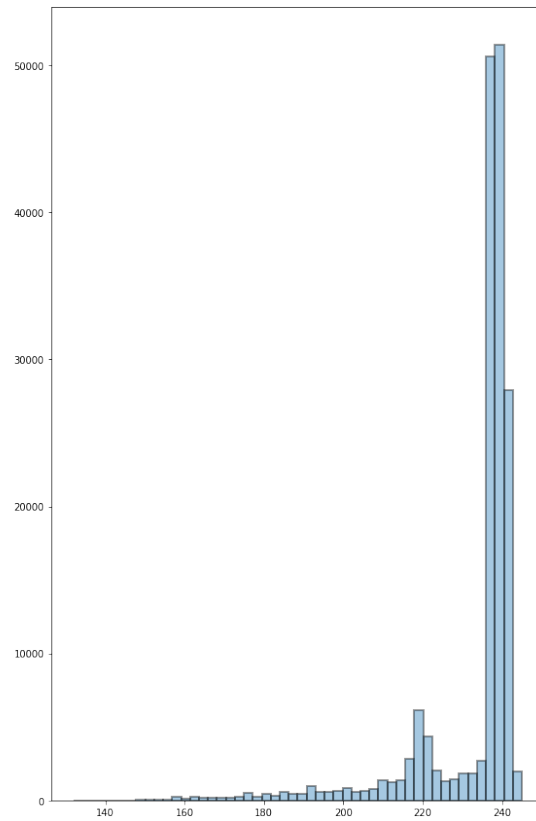
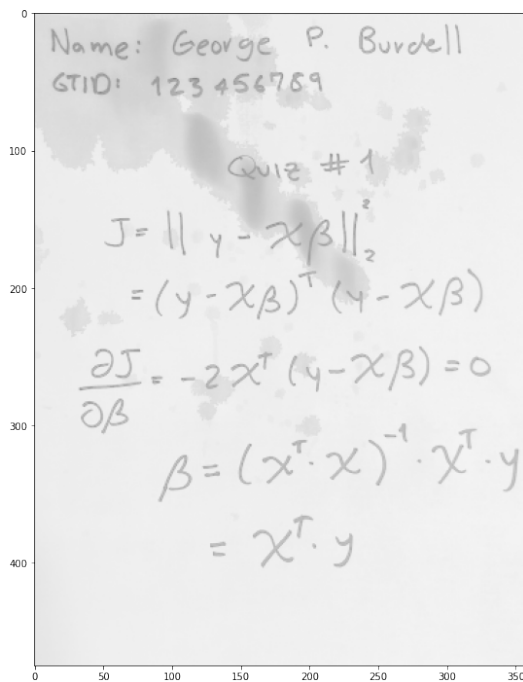


```
[19]: log_img = log(gray50)
```

```
[20]: plt.subplots(nrows=1, ncols=2, figsize=(20,15))
plt.subplot(121)
plt.imshow(log_img, cmap='gray', vmin=0, vmax=255)

plt.subplot(122)
sns.distplot(log_img.reshape(-1), kde=False, hist_kws=dict(edgecolor="k",
    linewidth=2))

plt.show()
```



0.0.8 7.) Power Function Transformation

```
[21]: def power(image):
    img = image.copy()
    c=0.1
    gam = 1.5

    powpix = lambda x: c * x**gam
```

```

img = powpix(img).astype(int)

img[img > 255] = 255

return img

```

```
[22]: pow_img = power(gray50)
```

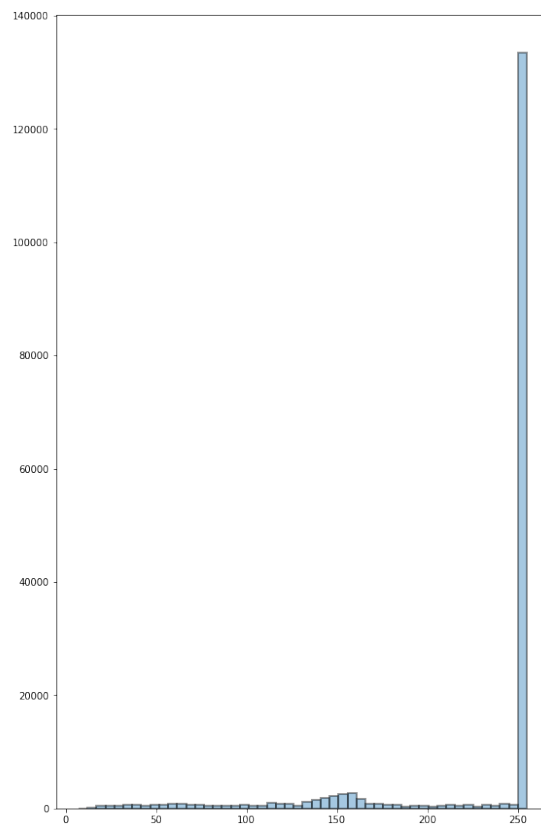
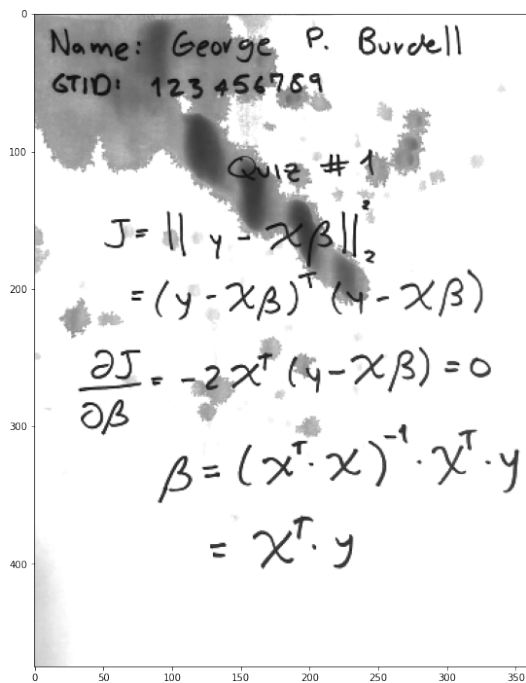
```

[23]: plt.subplots(nrows=1, ncols=2, figsize=(20,15))
plt.subplot(121)
plt.imshow(pow_img, cmap='gray', vmin=0, vmax=255)

plt.subplot(122)
sns.distplot(pow_img.reshape(-1), kde=False, hist_kws=dict(edgecolor="k",
↪ linewidth=2))

plt.show()

```



0.0.9 8.) Black and White

```
[24]: def bnw(image, thresh):  
        img = image.copy()  
        img[img <= thresh] = 0  
        img[img > thresh] = 1  
  
        return img
```

0.0.10 linear image to black and white

```
[25]: lin_bw = bnw(lin_img, 183)
```

0.0.11 power image to black and white

```
[26]: pow_bw = bnw(pow_img, 61)
```

0.0.12 logarithmic image to black and white

```
[27]: log_bw = bnw(log_img, 193)
```

0.0.13 histogram stretch to black and white

```
[28]: stretch_bw = bnw(img_stretch, 65)
```

0.0.14 Pixel shift to black and white

```
[29]: shift_bw = bnw(shift, 173)
```

0.0.15 All results

```
[30]: results = [lin_bw, pow_bw, log_bw, stretch_bw, shift_bw]  
titles = ['linear transform', 'power transform', 'log transform', 'histogram_  
→stretch', 'pixel shift']
```

```
[31]: for i in range(len(results)):  
        plt.figure(figsize=(15,8))  
        plt.imshow(results[i], cmap='gray')  
        plt.axis('off')
```

```
plt.title(titles[i])  
plt.show()
```

linear transform

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Quiz #1

$$J = \|y - X\beta\|_2^2$$
$$= (y - X\beta)^T (y - X\beta)$$
$$\frac{\partial J}{\partial \beta} = -2X^T (y - X\beta) = 0$$
$$\beta = (X^T X)^{-1} X^T y$$
$$= X^T y$$

power transform

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! Quiz #1

$$J = \|y - X\beta\|_2^2 \\ = (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial J}{\partial \beta} = -2X^T (y - X\beta) = 0$$

$$\beta = (X^T X)^{-1} X^T y \\ = X^+ y$$

log transform

Name: George P. Burdell

GTID: 123456789

Quiz #1

$$J = \|y - X\beta\|_2^2 \\ = (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial J}{\partial \beta} = -2X^T (y - X\beta) = 0$$

$$\beta = (X^T X)^{-1} \cdot X^T \cdot y \\ = X^T \cdot y$$

histogram stretch

Name: George P. Burdell

GTID: 123456789

Quiz #1

$$J = \|y - X\beta\|_2^2 \\ = (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial J}{\partial \beta} = -2X^T (y - X\beta) = 0$$

$$\beta = (X^T X)^{-1} X^T y \\ = X^T y$$

pixel shift

Name: George P. Burdell

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Quiz #1

$$J = \|y - X\beta\|_2^2 \\ = (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial J}{\partial \beta} = -2X^T (y - X\beta) = 0$$

$$\beta = (X^T X)^{-1} X^T y \\ = X^+ y$$

1 Winner Winner Chicken Dinner

```
[32]: plt.figure(figsize=(30,20))  
plt.imshow(lin_bw, cmap='gray')  
plt.axis('off')  
plt.show()
```


Name: George P. Burdell

GTID: 123456789

Quiz #1

$$J = \|y - X\beta\|_2^2$$

$$= (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial J}{\partial \beta} = -2X^T(y - X\beta) = 0$$

$$\begin{aligned}\beta &= (X^T X)^{-1} X^T y \\ &= X^T y\end{aligned}$$

[]: