

## Artificial Intelligence and Machine Learning

**Applications** 





- Linear Regression Tutorial & Playground
- Image Compression
- Logistic Regression Interactive Demo
- Neural Network Regression Tutorial & Playground
- Auto-Encoders



## Linear Regression Tutorial

https://mlu-explain.github.io/linear-regression/

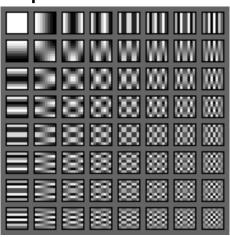


## Linear Regression Playground

https://observablehq.com/@yizhe-ang/interactive-visualization-of-linear-regression

# Image Compression with Linear Regression

- An image can be transformed into the frequency domain and represented as a combination of some basic components.
- Cosine Basis Functions and Discrete Cosine Transforms (DCT) can enable this.
- The human eye is most sensitive to low frequencies. Therefore, most of the high frequencies can be ignored.
- Principal behind JPEG Image Compression.

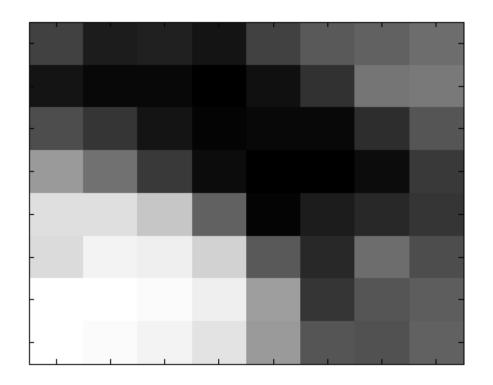


# Image Compression with Linear Regression

- Using linear regression, we can learn the weight of each of these cosine basis.
- We only need to store the weights of the basis frequencies.
- The image is reconstructed using these weights.



8 x 8 Pixels



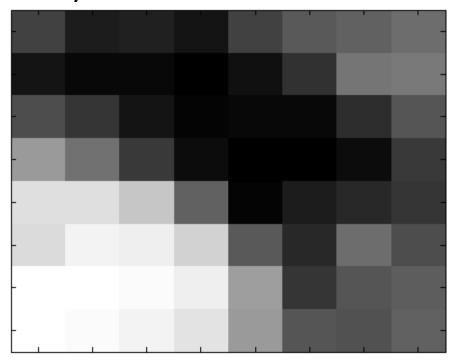
**Image** 





- Gray-Scale Example:
- Value Range 0 (black) --- 255 (white)

```
63 33 36 28 63 81 86 98
27 18 17 11 22 48 104 108
72 52 28 15 17 16 47 77
132 100 56 19 10 9 21 55
187 186 166 88 13 34 43 51
184 203 199 177 82 44 97 73
211 214 208 198 134 52 78 83
211 210 203 191 133 79 74 86
```





#### 2D-DCT of matrix

Numbers are coefficients of polynomial

```
-304 210 104 -69 10 20 -12 7

-327 -260 67 70 -10 -15 21 8

93 -84 -66 16 24 -2 -5 9

89 33 -19 -20 -26 21 -3 0

-9 42 18 27 -7 -17 29 -7

-5 15 -10 17 32 -15 -4 7

10 3 -12 -1 2 3 -2 -3

12 30 0 -3 -3 -6 12 -1
```





Cut the least significant components

```
      -304
      210
      104
      -69
      10
      20
      -12
      0

      -327
      -260
      67
      70
      -10
      -15
      0
      0

      93
      -84
      -66
      16
      24
      0
      0
      0

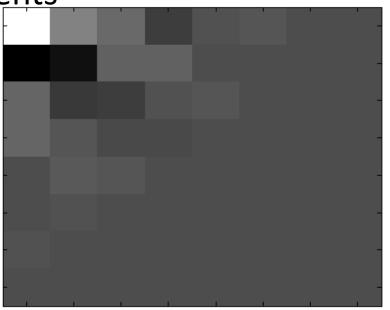
      89
      33
      -19
      -20
      0
      0
      0
      0

      -9
      42
      18
      0
      0
      0
      0
      0

      -5
      15
      0
      0
      0
      0
      0
      0

      10
      0
      0
      0
      0
      0
      0

      0
      0
      0
      0
      0
      0
```



As you can see, we save a little over half the original memory.



## Reconstructing the Image

New Matrix and Compressed Image

```
      55
      41
      27
      39
      56 69 92 106

      35
      22
      7
      16
      35 59 88 101

      65
      49
      21
      5
      6 28 62 73

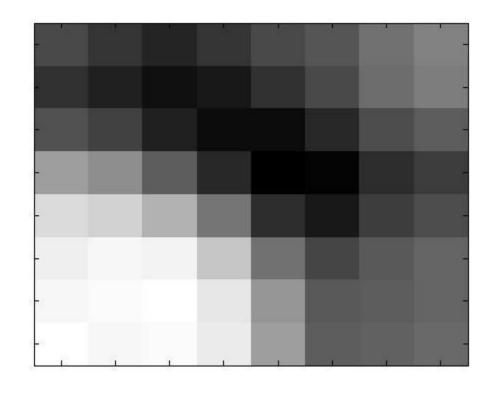
      130 114
      75
      28
      -7 -1 33 46

      180 175 148
      95
      33 16 45 59

      200 206 203 165
      92 55 71 82

      205 207 214 193 121 70 75 83

      214 205 209 196 129 75 78 85
```

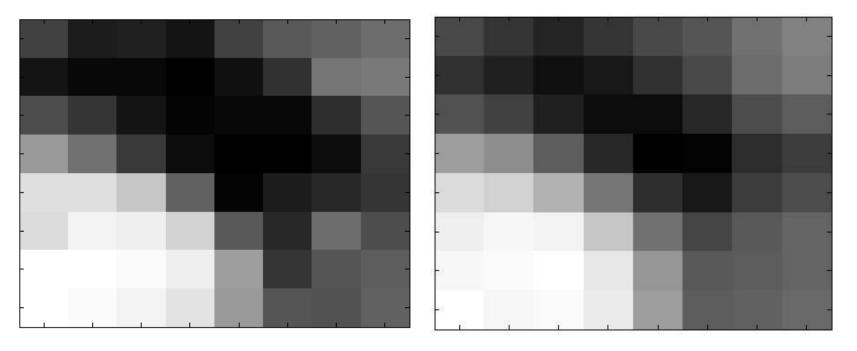




### Can You Tell the Difference?

#### Original

#### Compressed





### Original



### Compressed





## Logistic Regression Interactive Tutorial

https://mlu-explain.github.io/logistic-regression/



#### Neural Networks Tutorial

https://mlu-explain.github.io/neural-networks/



## Neural Networks Playground

https://playground.tensorflow.org/



## Try it Yourself – Digit Classification

https://trekhleb.dev/machine-learningexperiments/#/experiments/DigitsRecognitionMLP



## Try it Yourself – Sketch Recognition

https://trekhleb.dev/machine-learningexperiments/#/experiments/SketchRecognitionMLP

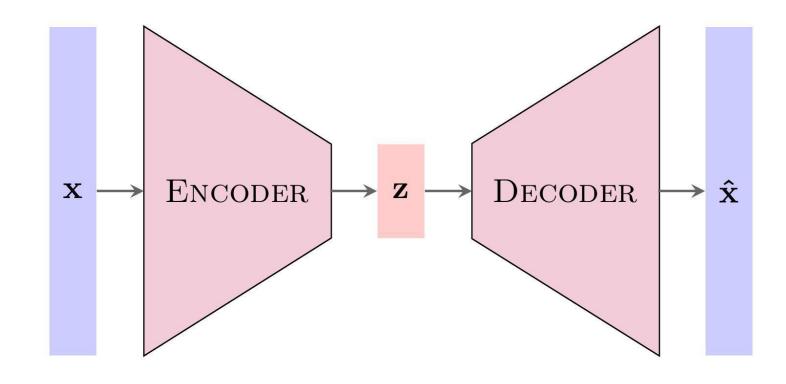


## Applications – AutoEncoders

- Autoencoders are a type of neural networks where the input is also the output.
- They come under unsupervised learning and there are no labels involved.
- An autoencoder consists of two parts: encoder and decoder.
- The idea here is that you take a higher dimensional input, project it into a lower dimensional space and then project it back into the input space.



## Applications – AutoEncoders





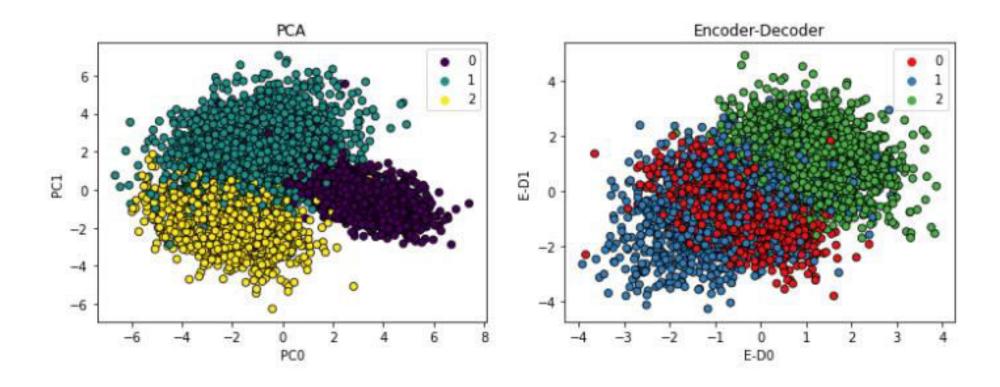
## Applications – AutoEncoders

- The autoencoder model tries to minimize the reconstruction error (RE).
- Typically, mean squared is used as the loss function for autoencoders.
- The objective is to minimize the following:

$$L(x,\hat{x}) = rac{1}{N} \sum_{i=1}^{N} ||x_i - \hat{x}_i||^2$$

# AutoEncoders for Dimensionality Reduction

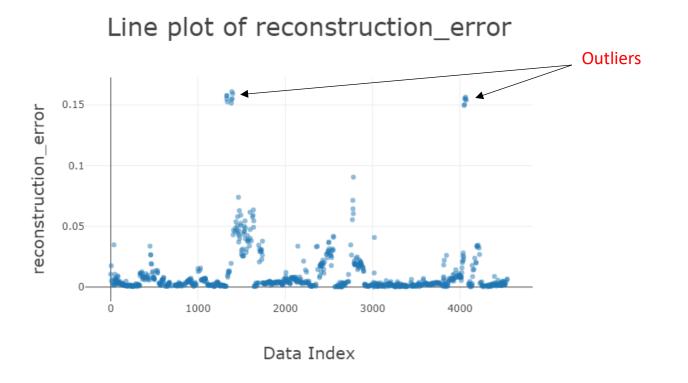
The encoder output can be used for dimensionality reduction





#### AutoEncoders for Outlier Detection

• The reconstruction error can be used to detect outliers. Out of distribution samples will have high Reconstruction error.





## AutoEncoders for Image Completion

