

## Outlines

- (1) Installation
- (2) Download dataset
- (3) Visualize image
- (4) train classifier
- (5) Evaluate
- (6) visualize insights

~~Docker Setup~~: How to install TF using Docker  
Starting by ensuring Docker itself is installed.

## Installation

### • Docker Setup:

- Install Docker, open terminal, note the Docker
- Run Tensorflow Docker image from Docker Hub<sup>TF</sup>
- Containers Start Jupyter Notebooks on port 8888 - access via browser.

### • Pre-Installed Packages include:

- Tensorflow with tf.contrib.learn
- matplotlib for plotting
- TF.learn high-level Estimators.

## Download dataset

→ Load MNIST via a one-liner

~~mnist~~ `mnist = learn.datasets.load_dataset('mnist')`

- Comes with 55k training, 10k test 28x28 grayscale images.

→ Access via: `mnist.train.images, labels`  
`mnist.test.images, labels`

→ Can you subset for faster iteration



## Visualize images

- Plot Sample digits with a helper function.  
`image.reshape((28, 28))`
- Images are 784-length 1D-arrays - reshape to 2D for plot
- Displays a variety of clear to ambiguous handwriting samples!

## Train Linear classifier

- problem: multi class classification (digits 0-9)
- features: raw pixel values, 784 per image
- flatten image  $\rightarrow$  Linear Classification image

feature columns = real-valued columns ("1", dimension)  
Classifier = Linear Classifier (n-classes=10, feature  
columns = feature-columns)

### Conceptually:

- 784 Input nodes, 10 output nodes, fully connected via weights.
- Each pixel contributes evidence (weight  $\times$  pixel value) to each class.

### Train using

- 784 inputs nodes, 10 output nodes, fully connected via weights.
- Each pixel contributes evidence (weight  $\times$

Classifier: `fit(data, labels, batch_size=100, steps=1000)`



## Evaluate the models

- Evaluate Performance:  
Classifier. evaluate (test\_data, test\_labels) (accuracy)
- Approximate Accuracy: ~90%.
- Use Predict () to see correct and incorrect example predictions.

Embeddings  
Loading  
Class  
Bucket

## Visualize Learned Weights

- Access learned weights for each class (784 dims)
- Reshape to  $28 \times 28$  and plot with a red/blue (seismic) color map.
- Red = positive weight (evidence for class)
- Blue = -ve weight (evidence against)

## Interpretation:

- middle pixel strong +ve for '1'.
- Same pixel strong -ve for '0'.

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