

# Learning from zero...

Some issues you might face



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# Steps to take

1. Get as many examples of shoes as possible
2. Train using these examples
3. Profit!





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# Steps to take

1. Get as many examples of shoes as possible
2. Train using these examples
3. Profit!

Training accuracy: .920

Training accuracy: .935

Training accuracy: .947

Training accuracy: .961

Training accuracy: .977

Training accuracy: .995

Training accuracy: 1.00

# Steps to take

1. ~~Get as many examples of shoes as possible~~
2. ~~Train using these examples~~
3. Profit?



# Data

The network 'sees' everything. Has no context for measuring how well it does with data it has never previously been exposed to.

## Data

## Validation Data

The network 'sees' a subset of your data. You can use the rest to measure its performance against previously unseen data.



**Data**

**Validation Data**

**Test Data**

The network 'sees' a subset of your data. You can use an unseen subset to measure its accuracy while training (validation), and then another subset to measure its accuracy after it's finished training (test).

**Data**

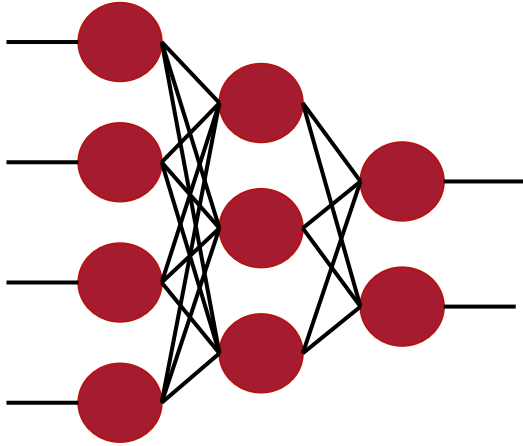
**Validation Data**

**Test Data**

**Accuracy: 0.999**

**Accuracy: 0.920**

**Accuracy: 0.800**



**Data**

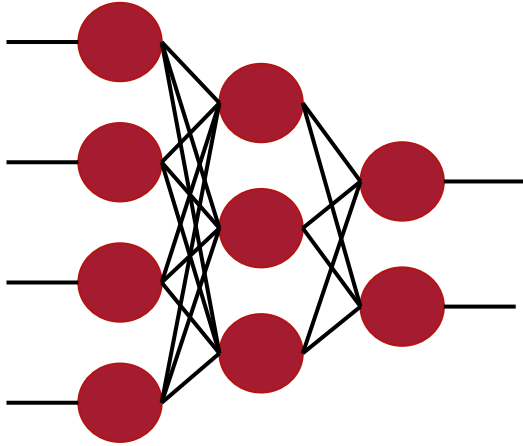
**Validation Data**

**Test Data**

**Accuracy: 0.999**

**Accuracy: 0.920**

**Accuracy: 0.800**



**Data**

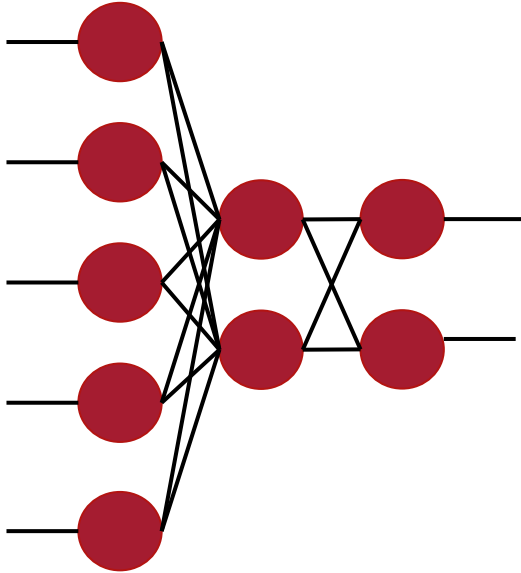
**Validation Data**

**Test Data**

**Accuracy: 0.942**

**Accuracy: 0.930**

**Accuracy: 0.925**



```
import tensorflow as tf

data = tf.keras.datasets.mnist
(training_images, training_labels), (val_images, val_labels) = data.load_data()

training_images = training_images / 255.0
val_images = val_images / 255.0

model = tf.keras.models.Sequential(
    [tf.keras.layers.Flatten(input_shape=(28,28)),
     tf.keras.layers.Dense(20, activation=tf.nn.relu),
     tf.keras.layers.Dense(10, activation=tf.nn.softmax)])
```



```
model.fit(training_images, training_labels, epochs=20)
```

```
Epoch 16/20
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.0931 - accuracy: 0.9729
```

```
Epoch 17/20
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.0907 - accuracy: 0.9731
```

```
Epoch 18/20
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.0892 - accuracy: 0.9735
```

```
Epoch 19/20
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.0868 - accuracy: 0.9740
```

```
Epoch 20/20
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.0844 - accuracy: 0.9750
```

```
model.fit(training_images, training_labels,  
          validation_data=(val_images, val_labels),  
          epochs=20)
```

```
model.fit(training_images, training_labels,  
          validation_data=(val_images, val_labels),  
          epochs=20)
```

```
Epoch 14/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0960 - accuracy: 0.9713 - val_loss: 0.1410 - val_accuracy: 0.9588  
Epoch 15/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0933 - accuracy: 0.9722 - val_loss: 0.1353 - val_accuracy: 0.9626  
Epoch 16/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0910 - accuracy: 0.9729 - val_loss: 0.1356 - val_accuracy: 0.9622  
Epoch 17/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0880 - accuracy: 0.9734 - val_loss: 0.1339 - val_accuracy: 0.9625  
Epoch 18/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0855 - accuracy: 0.9740 - val_loss: 0.1349 - val_accuracy: 0.9619  
Epoch 19/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0832 - accuracy: 0.9754 - val_loss: 0.1341 - val_accuracy: 0.9624  
Epoch 20/20  
1875/1875 [=====] - 4s 2ms/step - loss: 0.0808 - accuracy: 0.9759 - val_loss: 0.1340 - val_accuracy: 0.9626
```

```
model.evaluate(test_images, test_labels)
```

```
313/313 [=====] - 1s 2ms/step - loss: 0.1495 - accuracy: 0.9569
```

Quiz!