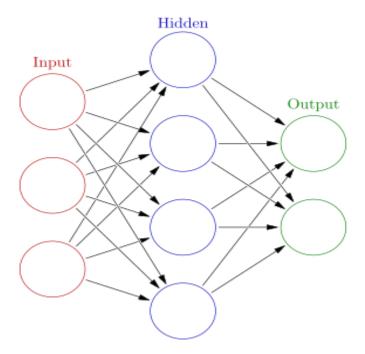
Implementing a one-layer Neural Network

We will illustrate how to create a one hidden layer NN

We will use the iris data for this exercise

We will build a one-hidden layer neural network to predict the fourth attribute, Petal Width from the other three (Sepal length, Sepal width, Petal length).



```
In [1]:
"""

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"""

import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf1
import tensorflow.compat.v1 as tf
from sklearn import datasets
```

```
tf.disable v2 behavior()
from tensorflow.python.framework import ops
ops.reset default graph()
In [2]:
ops.reset default graph()
In [3]:
iris = datasets.load iris()
x \text{ vals} = \text{np.array}([x[0:3] \text{ for } x \text{ in } iris.data])
y_vals = np.array([x[3] for x in iris.data])
In [4]:
# Create graph session
sess = tf.Session()
In [5]:
# make results reproducible
seed = 2
tf.set_random_seed(seed)
np.random.seed(seed)
In [6]:
# Split data into train/test = 80%/20%
train indices = np.random.choice(len(x vals), round(len(x vals)*0.8), repla
ce=False)
test indices = np.array(list(set(range(len(x vals))) - set(train indices)))
x_vals_train = x_vals[train_indices]
x vals test = x vals[test indices]
y_vals_train = y_vals[train_indices]
y vals test = y vals[test indices]
In [7]:
# Normalize by column (min-max norm)
def normalize cols(m):
    col max = m.max(axis=0)
    col min = m.min(axis=0)
    return (m-col min) / (col max - col min)
x vals train = np.nan to num(normalize cols(x vals train))
x vals test = np.nan to num(normalize cols(x vals test))
```

```
In [8]:
# Declare batch size
batch size = 50
# Initialize placeholders
x data = tf.placeholder(shape=[None, 3], dtype=tf.float32)
y target = tf.placeholder(shape=[None, 1], dtype=tf.float32)
In [9]:
# Create variables for both NN layers
hidden layer nodes = 10
A1 = tf.Variable(tf.random normal(shape=[3,hidden layer nodes])) # inputs -
> hidden nodes
b1 = tf.Variable(tf.random normal(shape=[hidden layer nodes])) # one bias
es for each hidden node
A2 = tf.Variable(tf.random normal(shape=[hidden layer nodes,1])) # hidden i
nputs -> 1 output
b2 = tf.Variable(tf.random normal(shape=[1]))  # 1 bias for the output
# Declare model operations
hidden output = tf.nn.relu(tf.add(tf.matmul(x data, A1), b1))
final output = tf.nn.relu(tf.add(tf.matmul(hidden output, A2), b2))
# Declare loss function (MSE)
loss = tf.reduce mean(tf.square(y target - final output))
# Declare optimizer
my opt = tf.train.GradientDescentOptimizer(0.005)
train step = my opt.minimize(loss)
In [10]:
# Initialize variables
init = tf.global variables initializer()
sess.run(init)
# Training loop
loss_vec = []
test loss = []
for i in range (500):
    rand index = np.random.choice(len(x vals train), size=batch size)
    rand x = x vals train[rand index]
    rand_y = np.transpose([y_vals_train[rand_index]])
    sess.run(train step, feed dict={x data: rand x, y target: rand y})
```

```
temp_loss = sess.run(loss, feed_dict={x_data: rand_x, y_target: rand_y})

loss_vec.append(np.sqrt(temp_loss))

test_temp_loss = sess.run(loss, feed_dict={x_data: x_vals_test, y_targe});

test_loss.append(np.sqrt(test_temp_loss))

if (i+1)%50==0:
    print('Generation: ' + str(i+1) + '. Loss = ' + str(temp_loss))
```