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```
# Worksheet 23

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### Topics

- Tuning Neural Networks

## Tuning Neural Networks

Nothing to do in this worksheet except follow along in lecture to better understand Neural Networks.
```

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Topics

· Tuning Neural Networks

Tuning Neural Networks

Nothing to do in this worksheet except follow along in lecture / use this code to better understand Neural Networks.

```
import math as m
import numpy as np
import matplotlib.pyplot as plt
import sklearn.datasets as datasets
from tensorflow import keras, math, random, stack
from tensorflow.keras import layers, initializers
from tensorflow.keras.activations import relu
# Set random seed for reproducibility
np. random. seed (1)
random.set_seed(1)
# Data generation - don't modify
centers = [[0, 0]]
t, _ = datasets.make_blobs(n_samples=200, centers=centers, cluster_std=1,
                                                               random_state=1)
colors = np.array([x for x in 'bgrcmyk'])
# CURVE
def generate_curve_data(t):
        # create some space between the classes
         \text{X = np.array(list(filter(lambda \ x \ : \ m.cos(4*x[0]) \ - \ x[1] \ < \ -.5 \ \ \text{or} \ \ m.cos(4*x[0]) \ \ - \ x[1] \ > \ .5, \ \ t)))} 
         \label{eq:conditional} Y = \text{np.array}([1 \text{ if } \textbf{m.}\cos(4*x[0]) - x[1] >= 0 \text{ else } 0 \text{ for } x \text{ in } X]) 
        return X, Y
# The model - modify this
model = keras. models. Sequential()
\verb| model.add(layers.Dense(3, input\_dim=2, activation="sigmoid"))|\\
model.add(layers.Dense(1, activation="sigmoid"))
model.compile(loss="binary_crossentropy")
X, Y = generate_curve_data(t)
# plot the data
plt. \ scatter(X[:,0],X[:,1],color=colors[Y].\ tolist(), \quad s=100, \quad alpha=.9)
plt.show()
history = model.fit(X, Y, batch_size=50, epochs=200)
# Plot the decision boundary
# create a mesh to plot in
h = .02 # step size in the mesh
x_{min}, x_{max} = X[:, 0].min() - .5, X[:, 0].max() + 1
y_min, y_max = X[:, 1].min() - .5, X[:, 1].max() + 1
xx, yy = np. meshgrid(np. arange(x_min, x_max, h),
                                          np.arange(y_min, y_max, h))
meshData = np.c_[xx.ravel(), yy.ravel()]
fig, ax = plt.subplots()
Z = model.predict(meshData)
  = np.array([0 if x < .5 else 1 for x = Z])
Z = Z.reshape(xx.shape)
ax. contourf(xx, yy, Z, alpha=.3, cmap=plt.cm.Paired)
ax.axis('off')
# Plot also the training points
ax.\,scatter\,(\texttt{X[:, 0], X[:, 1], color=colors[Y].tolist(), s=100, alpha=.9)}\\
plt.title("Decision Boundary")
```

plt.show()