

```
In [ ]: # Import our dependencies
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, OneHotEncoder, MinMaxScaler
import pandas as pd
import tensorflow as tf
import numpy as np

# Import our input dataset
df = pd.read_csv('../neural-network/pitcher_salaries_cleaned.csv')
df.head()
```

```
In [ ]: # create log transformed column for salary
df['sal-log'] = np.log10(df['Salary'])
df
```

## Reduce down to top features

---

```
In [ ]: df = df.drop(["Full Name", "Team", "League", "Age", "Earned Runs", "Home Runs", "Wins", "Losses", "Weight", "Height"])
df.head()
```

## Split Features/Target & Training/Testing Sets

---

Split into features and target

- **y variable:** Our target variable, Salary
- **X variable:** Our features; just drop Salary and Full Name

```
In [ ]: # Split our preprocessed data into our features and target arrays
y = df["sal-log"].values
X = df.drop(["sal-log"], 1).values

# Split the preprocessed data into a training and testing dataset
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=1)
```

## Build and Instantiate StandardScaler object, then standardize numerical features

---

```
In [ ]: # Create a StandardScaler instance
scaler = MinMaxScaler()

# Fit the StandardScaler
X_scaler = scaler.fit(X_train)

# Scale the data
X_train_scaled = X_scaler.transform(X_train)
X_test_scaled = X_scaler.transform(X_test)
```

```
In [ ]: # see if data scaled properly
scaled_data = pd.DataFrame(X_train_scaled)
scaled_data.head()
```

```
In [ ]: # see if data scaled properly
scaled_y = pd.DataFrame(y_train_scaled)
scaled_y.head()
```

## Build Neural Net Framework

```
In [56]: # Define the model - deep neural net
number_input_features = len(X_train[0])
hidden_nodes_layer1 = 45
hidden_nodes_layer2 = 45
hidden_nodes_layer3 = 20
hidden_nodes_layer4 = 12
hidden_nodes_layer5 = 11

nn = tf.keras.models.Sequential()

# First hidden layer
nn.add(
    tf.keras.layers.Dense(units=hidden_nodes_layer1, input_dim=number_input_features, activation="elu")
)

# Second hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer2, activation="elu"))

# Third hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer3, activation="elu"))

# Fourth hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer4, activation="elu"))

# Fifth hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer5, activation="elu"))

# Output layer
nn.add(tf.keras.layers.Dense(units=10, activation="elu"))

# Check the structure of the model
nn.summary()
```

Model: "sequential\_9"

Layer (type)	Output Shape	Param #
=====		
dense_46 (Dense)	(None, 45)	360
dense_47 (Dense)	(None, 45)	2070
dense_48 (Dense)	(None, 20)	920
dense_49 (Dense)	(None, 12)	252
dense_50 (Dense)	(None, 11)	143
dense_51 (Dense)	(None, 10)	120
=====		
Total params: 3,865		
Trainable params: 3,865		
Non-trainable params: 0		

## Compile the Model

```
In [57]: # Compile the model
nn.compile(loss="mean_squared_logarithmic_error", optimizer="RMSprop", metrics=["accuracy"])
```

## Train the model

In [58]:

```
# Train the model
fit_model = nn.fit(X_train,y_train,epochs=200)
```

```
Epoch 1/200
116/116 [=====] - 1s 2ms/step - loss: 1.7872 - accuracy: 0.0049
Epoch 2/200
116/116 [=====] - 0s 1ms/step - loss: 1.2043 - accuracy: 0.0032
Epoch 3/200
116/116 [=====] - 0s 1ms/step - loss: 1.1706 - accuracy: 0.0022
Epoch 4/200
116/116 [=====] - 0s 2ms/step - loss: 1.1604 - accuracy: 0.0024
Epoch 5/200
116/116 [=====] - 0s 2ms/step - loss: 1.1560 - accuracy: 0.0024
Epoch 6/200
116/116 [=====] - 0s 1ms/step - loss: 1.1533 - accuracy: 0.0032
Epoch 7/200
116/116 [=====] - 0s 1ms/step - loss: 1.1515 - accuracy: 0.0030
Epoch 8/200
116/116 [=====] - 0s 1ms/step - loss: 1.1505 - accuracy: 0.0024
Epoch 9/200
116/116 [=====] - 0s 1ms/step - loss: 1.1499 - accuracy: 0.0027
Epoch 10/200
116/116 [=====] - 0s 2ms/step - loss: 1.1493 - accuracy: 0.0035
Epoch 11/200
116/116 [=====] - 0s 2ms/step - loss: 1.1490 - accuracy: 0.0022
Epoch 12/200
116/116 [=====] - 0s 2ms/step - loss: 1.1488 - accuracy: 0.0024
Epoch 13/200
116/116 [=====] - 0s 1ms/step - loss: 1.1485 - accuracy: 0.0027
Epoch 14/200
116/116 [=====] - 0s 1ms/step - loss: 1.1484 - accuracy: 0.0016
Epoch 15/200
116/116 [=====] - 0s 3ms/step - loss: 1.1484 - accuracy: 0.0027
Epoch 16/200
116/116 [=====] - 0s 2ms/step - loss: 1.1483 - accuracy: 0.0027
Epoch 17/200
116/116 [=====] - 0s 2ms/step - loss: 1.1482 - accuracy: 0.0022
Epoch 18/200
116/116 [=====] - 0s 1ms/step - loss: 1.1480 - accuracy: 0.0024
Epoch 19/200
116/116 [=====] - 0s 1ms/step - loss: 1.1480 - accuracy: 0.0030
Epoch 20/200
116/116 [=====] - 0s 1ms/step - loss: 1.1480 - accuracy: 0.0024
Epoch 21/200
116/116 [=====] - 0s 1ms/step - loss: 1.1480 - accuracy: 0.0027
Epoch 22/200
116/116 [=====] - 0s 1ms/step - loss: 1.1479 - accuracy: 0.0030
Epoch 23/200
116/116 [=====] - 0s 1ms/step - loss: 1.1479 - accuracy: 0.0027
Epoch 24/200
116/116 [=====] - 0s 1ms/step - loss: 1.1478 - accuracy: 0.0014
Epoch 25/200
116/116 [=====] - 0s 1ms/step - loss: 1.1478 - accuracy: 0.0019
Epoch 26/200
116/116 [=====] - 0s 2ms/step - loss: 1.1479 - accuracy: 0.0027
Epoch 27/200
116/116 [=====] - 0s 2ms/step - loss: 1.1478 - accuracy: 0.0024
Epoch 28/200
116/116 [=====] - 0s 2ms/step - loss: 1.1477 - accuracy: 0.0024
Epoch 29/200
116/116 [=====] - 0s 2ms/step - loss: 1.1477 - accuracy: 0.0027
Epoch 30/200
116/116 [=====] - 0s 1ms/step - loss: 1.1477 - accuracy: 0.0027
Epoch 31/200
116/116 [=====] - 0s 2ms/step - loss: 1.1477 - accuracy: 0.0035
Epoch 32/200
116/116 [=====] - 0s 1ms/step - loss: 1.1477 - accuracy: 0.0024
Epoch 33/200
116/116 [=====] - 0s 1ms/step - loss: 1.1476 - accuracy: 0.0014
Epoch 34/200
116/116 [=====] - 0s 2ms/step - loss: 1.1477 - accuracy: 0.0030
Epoch 35/200
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116/116 [=====] - 0s 2ms/step - loss: 1.1476 - accuracy: 0.0027  
Epoch 36/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1476 - accuracy: 0.0030  
Epoch 37/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1476 - accuracy: 0.0016  
Epoch 38/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1476 - accuracy: 0.0014  
Epoch 39/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1476 - accuracy: 0.0024  
Epoch 40/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1476 - accuracy: 0.0030  
Epoch 41/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1476 - accuracy: 0.0011  
Epoch 42/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0024  
Epoch 43/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 44/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 45/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 46/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1476 - accuracy: 0.0011  
Epoch 47/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 5.4025e-04  
Epoch 48/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 49/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 50/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0019  
Epoch 51/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 5.4025e-04  
Epoch 52/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0016  
Epoch 53/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 8.1037e-04  
Epoch 54/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 55/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1475 - accuracy: 0.0016  
Epoch 56/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0014  
Epoch 57/200  
116/116 [=====] - 0s 922us/step - loss: 1.1475 - accuracy: 0.0022  
Epoch 58/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0016  
Epoch 59/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 60/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0014  
Epoch 61/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0011  
Epoch 62/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0022  
Epoch 63/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0022  
Epoch 64/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0014  
Epoch 65/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 66/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 67/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0016  
Epoch 68/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 69/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 70/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 8.1037e-04  
Epoch 71/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 72/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0027  
Epoch 73/200

116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 74/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 5.4025e-04  
Epoch 75/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 76/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1475 - accuracy: 0.0019  
Epoch 77/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 78/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 79/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 80/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 81/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 82/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 83/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 84/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
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116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 86/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 87/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 88/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 89/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
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116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 91/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 92/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 93/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 94/200  
116/116 [=====] - 0s 948us/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 95/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 96/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 97/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 98/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 2.7012e-04  
Epoch 99/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 100/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 101/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0030  
Epoch 102/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 103/200  
116/116 [=====] - 0s 991us/step - loss: 1.1474 - accuracy: 8.1037e-04  
Epoch 104/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 105/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 106/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0030  
Epoch 107/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 108/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 8.1037e-04  
Epoch 109/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 110/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 2.7012e-04  
Epoch 111/200

116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 112/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 113/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0024  
Epoch 114/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 8.1037e-04  
Epoch 115/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 116/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 117/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 118/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 5.4025e-04  
Epoch 119/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 120/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 121/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0022  
Epoch 122/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 123/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0027  
Epoch 124/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 5.4025e-04  
Epoch 125/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 126/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0027  
Epoch 127/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 128/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 8.1037e-04  
Epoch 129/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1473 - accuracy: 0.0011  
Epoch 130/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0014  
Epoch 131/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0014  
Epoch 132/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0019  
Epoch 133/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 134/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 8.1037e-04  
Epoch 135/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0011  
Epoch 136/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1473 - accuracy: 0.0024  
Epoch 137/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0014  
Epoch 138/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 8.1037e-04  
Epoch 139/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0016  
Epoch 140/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0022  
Epoch 141/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0011  
Epoch 142/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 143/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 2.7012e-04  
Epoch 144/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1474 - accuracy: 0.0016  
Epoch 145/200  
116/116 [=====] - 0s 3ms/step - loss: 1.1473 - accuracy: 8.1037e-04  
Epoch 146/200  
116/116 [=====] - 0s 2ms/step - loss: 1.1474 - accuracy: 0.0014  
Epoch 147/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 8.1037e-04  
Epoch 148/200  
116/116 [=====] - 0s 1ms/step - loss: 1.1473 - accuracy: 0.0011  
Epoch 149/200

116/116 [=====] - 0s 2ms/step - loss: 0.6185 - accuracy: 2.7012e-04  
Epoch 150/200  
116/116 [=====] - 0s 1ms/step - loss: 0.1825 - accuracy: 8.1037e-04  
Epoch 151/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0566 - accuracy: 0.0022  
Epoch 152/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0176 - accuracy: 5.4025e-04  
Epoch 153/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0094 - accuracy: 0.0024  
Epoch 154/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0084 - accuracy: 0.0065  
Epoch 155/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0081 - accuracy: 0.0059  
Epoch 156/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0077 - accuracy: 0.0059  
Epoch 157/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0074 - accuracy: 0.0054  
Epoch 158/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0074 - accuracy: 0.0057  
Epoch 159/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0072 - accuracy: 0.0043  
Epoch 160/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0072 - accuracy: 0.0054  
Epoch 161/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0071 - accuracy: 0.0049  
Epoch 162/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0072 - accuracy: 0.0043  
Epoch 163/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0041  
Epoch 164/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0043  
Epoch 165/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0046  
Epoch 166/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0046  
Epoch 167/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0070 - accuracy: 0.0046  
Epoch 168/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0032  
Epoch 169/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0035  
Epoch 170/200  
116/116 [=====] - 0s 896us/step - loss: 0.0070 - accuracy: 0.0041  
Epoch 171/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0027  
Epoch 172/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0024  
Epoch 173/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0070 - accuracy: 0.0022  
Epoch 174/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0022  
Epoch 175/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0016  
Epoch 176/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0011  
Epoch 177/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0016  
Epoch 178/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 5.4025e-04  
Epoch 179/200  
116/116 [=====] - 0s 991us/step - loss: 0.0070 - accuracy: 5.4025e-04  
Epoch 180/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0069 - accuracy: 0.0014  
Epoch 181/200  
116/116 [=====] - 0s 2ms/step - loss: 0.0069 - accuracy: 5.4025e-04  
Epoch 182/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 0.0016  
Epoch 183/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 5.4025e-04  
Epoch 184/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0071 - accuracy: 8.1037e-04  
Epoch 185/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 0.0014  
Epoch 186/200  
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 5.4025e-04  
Epoch 187/200

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116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 5.4025e-04
Epoch 188/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 8.1037e-04
Epoch 189/200
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 8.1037e-04
Epoch 190/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 5.4025e-04
Epoch 191/200
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 5.4025e-04
Epoch 192/200
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 2.7012e-04
Epoch 193/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 2.7012e-04
Epoch 194/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 0.0000e+00
Epoch 195/200
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 5.4025e-04
Epoch 196/200
116/116 [=====] - 0s 1ms/step - loss: 0.0069 - accuracy: 2.7012e-04
Epoch 197/200
116/116 [=====] - 0s 2ms/step - loss: 0.0068 - accuracy: 5.4025e-04
Epoch 198/200
116/116 [=====] - 0s 2ms/step - loss: 0.0069 - accuracy: 2.7012e-04
Epoch 199/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 2.7012e-04
Epoch 200/200
116/116 [=====] - 0s 1ms/step - loss: 0.0068 - accuracy: 0.0011

```

In [59]:

```

# Evaluate the model using the test data
model_loss, model_accuracy = nn.evaluate(X_test_scaled,y_test,verbose=2)
print(f"Loss: {model_loss}, Accuracy: {model_accuracy}")

```

```

39/39 - 0s - loss: 0.0401 - accuracy: 0.0049 - 150ms/epoch - 4ms/step
Loss: 0.04013507068157196, Accuracy: 0.004858299624174833

```

In [60]:

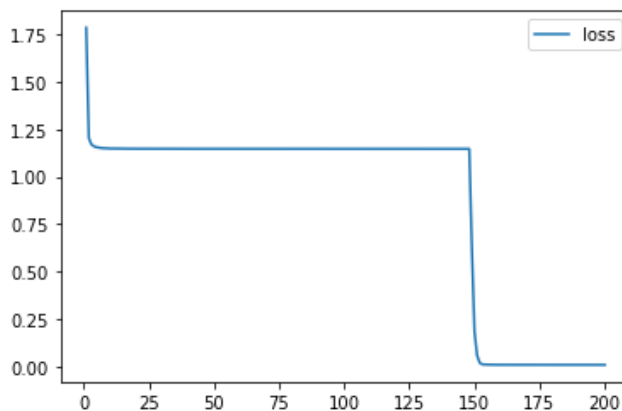
```

# Create a DataFrame containing training history
history_df = pd.DataFrame(fit_model.history, index=range(1,len(fit_model.history["loss"])+1))

# Plot the loss
history_df.plot(y="loss")

```

Out[60]: <AxesSubplot:>



In [61]:

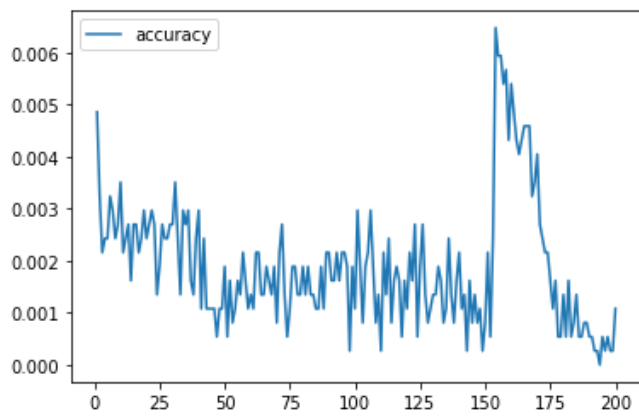
```

# Plot the accuracy
history_df.plot(y="accuracy")

```

Out[61]: <AxesSubplot:>





In [ ]: