

```
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 [54]: # Define the model - deep neural net
number_input_features = len(X_train[0])
hidden_nodes_layer1 = 100
hidden_nodes_layer2 = 45
hidden_nodes_layer3 = 20

nn = tf.keras.models.Sequential()

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

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

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

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

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

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

Model: "sequential_9"

Layer (type)	Output Shape	Param #
=====		
dense_45 (Dense)	(None, 100)	800
dense_46 (Dense)	(None, 45)	4545
dense_47 (Dense)	(None, 45)	2070
dense_48 (Dense)	(None, 45)	2070
dense_49 (Dense)	(None, 10)	460
=====		
Total params: 9,945		
Trainable params: 9,945		
Non-trainable params: 0		

Compile the Model

```
In [55]: # Compile the model
nn.compile(loss="mean_squared_error", optimizer="adam", metrics=["accuracy"])
```

Train the model

```
In [56]: # Train the model
fit_model = nn.fit(X_train,y_train,epochs=200)
```

Epoch 1/200
116/116 [=====] - 0s 835us/step - loss: 14.1331 - accuracy: 0.0014

Epoch 2/200
116/116 [=====] - 0s 817us/step - loss: 11.4165 - accuracy: 0.0016
Epoch 3/200
116/116 [=====] - 0s 849us/step - loss: 11.3996 - accuracy: 0.0024
Epoch 4/200
116/116 [=====] - 0s 979us/step - loss: 11.3922 - accuracy: 0.0043
Epoch 5/200
116/116 [=====] - 0s 870us/step - loss: 11.3791 - accuracy: 0.0070
Epoch 6/200
116/116 [=====] - 0s 800us/step - loss: 11.3804 - accuracy: 0.0062
Epoch 7/200
116/116 [=====] - 0s 974us/step - loss: 11.3803 - accuracy: 0.0057
Epoch 8/200
116/116 [=====] - 0s 1ms/step - loss: 11.3821 - accuracy: 0.0041
Epoch 9/200
116/116 [=====] - 0s 1ms/step - loss: 11.3870 - accuracy: 0.0043
Epoch 10/200
116/116 [=====] - 0s 809us/step - loss: 11.3811 - accuracy: 0.0046
Epoch 11/200
116/116 [=====] - 0s 843us/step - loss: 11.3815 - accuracy: 0.0016
Epoch 12/200
116/116 [=====] - 0s 774us/step - loss: 11.3780 - accuracy: 0.0038
Epoch 13/200
116/116 [=====] - 0s 852us/step - loss: 11.3822 - accuracy: 0.0032
Epoch 14/200
116/116 [=====] - 0s 1ms/step - loss: 11.3808 - accuracy: 0.0027
Epoch 15/200
116/116 [=====] - 0s 791us/step - loss: 11.3745 - accuracy: 0.0038
Epoch 16/200
116/116 [=====] - 0s 817us/step - loss: 11.3756 - accuracy: 0.0035
Epoch 17/200
116/116 [=====] - 0s 2ms/step - loss: 11.3798 - accuracy: 0.0030
Epoch 18/200
116/116 [=====] - 0s 2ms/step - loss: 11.3770 - accuracy: 0.0024
Epoch 19/200
116/116 [=====] - 0s 2ms/step - loss: 11.3810 - accuracy: 0.0027
Epoch 20/200
116/116 [=====] - 0s 913us/step - loss: 11.3835 - accuracy: 0.0032
Epoch 21/200
116/116 [=====] - 0s 1ms/step - loss: 11.3799 - accuracy: 0.0022
Epoch 22/200
116/116 [=====] - 0s 939us/step - loss: 11.3785 - accuracy: 0.0022
Epoch 23/200
116/116 [=====] - 0s 904us/step - loss: 11.3762 - accuracy: 0.0022
Epoch 24/200
116/116 [=====] - 0s 790us/step - loss: 11.3723 - accuracy: 0.0024
Epoch 25/200
116/116 [=====] - 0s 922us/step - loss: 11.3831 - accuracy: 0.0014
Epoch 26/200
116/116 [=====] - 0s 826us/step - loss: 11.3742 - accuracy: 0.0046
Epoch 27/200
116/116 [=====] - 0s 826us/step - loss: 11.3783 - accuracy: 0.0027
Epoch 28/200
116/116 [=====] - 0s 809us/step - loss: 11.3766 - accuracy: 0.0030
Epoch 29/200
116/116 [=====] - 0s 861us/step - loss: 11.3754 - accuracy: 0.0024
Epoch 30/200
116/116 [=====] - 0s 879us/step - loss: 11.3798 - accuracy: 0.0041
Epoch 31/200
116/116 [=====] - 0s 809us/step - loss: 11.3731 - accuracy: 0.0024
Epoch 32/200
116/116 [=====] - 0s 783us/step - loss: 11.3752 - accuracy: 0.0019
Epoch 33/200
116/116 [=====] - 0s 826us/step - loss: 11.3718 - accuracy: 0.0041
Epoch 34/200
116/116 [=====] - 0s 790us/step - loss: 11.3771 - accuracy: 0.0022
Epoch 35/200
116/116 [=====] - 0s 906us/step - loss: 11.3748 - accuracy: 0.0035
Epoch 36/200
116/116 [=====] - 0s 1ms/step - loss: 11.3827 - accuracy: 0.0022
Epoch 37/200
116/116 [=====] - 0s 809us/step - loss: 11.3774 - accuracy: 0.0027
Epoch 38/200
116/116 [=====] - 0s 1ms/step - loss: 11.3780 - accuracy: 0.0014
Epoch 39/200
116/116 [=====] - 0s 2ms/step - loss: 11.3726 - accuracy: 0.0027

Epoch 40/200
116/116 [=====] - 0s 1ms/step - loss: 11.3728 - accuracy: 0.0011
Epoch 41/200
116/116 [=====] - 0s 1ms/step - loss: 11.3801 - accuracy: 0.0027
Epoch 42/200
116/116 [=====] - 0s 817us/step - loss: 11.3754 - accuracy: 0.0035
Epoch 43/200
116/116 [=====] - 0s 861us/step - loss: 11.3736 - accuracy: 0.0022
Epoch 44/200
116/116 [=====] - 0s 807us/step - loss: 11.3793 - accuracy: 0.0027
Epoch 45/200
116/116 [=====] - 0s 861us/step - loss: 11.3746 - accuracy: 0.0030
Epoch 46/200
116/116 [=====] - 0s 818us/step - loss: 11.3797 - accuracy: 0.0011
Epoch 47/200
116/116 [=====] - 0s 1ms/step - loss: 11.3755 - accuracy: 0.0030
Epoch 48/200
116/116 [=====] - 0s 1ms/step - loss: 11.3783 - accuracy: 0.0030
Epoch 49/200
116/116 [=====] - 0s 965us/step - loss: 11.3821 - accuracy: 0.0016
Epoch 50/200
116/116 [=====] - 0s 860us/step - loss: 11.3777 - accuracy: 0.0014
Epoch 51/200
116/116 [=====] - 0s 817us/step - loss: 11.3778 - accuracy: 8.1037e-04
Epoch 52/200
116/116 [=====] - 0s 852us/step - loss: 11.3722 - accuracy: 0.0000e+00
Epoch 53/200
116/116 [=====] - 0s 853us/step - loss: 11.3757 - accuracy: 0.0011
Epoch 54/200
116/116 [=====] - 0s 843us/step - loss: 11.3739 - accuracy: 0.0054
Epoch 55/200
116/116 [=====] - 0s 843us/step - loss: 11.3756 - accuracy: 0.0014
Epoch 56/200
116/116 [=====] - 0s 817us/step - loss: 11.3783 - accuracy: 0.0030
Epoch 57/200
116/116 [=====] - 0s 913us/step - loss: 11.3738 - accuracy: 0.0019
Epoch 58/200
116/116 [=====] - 0s 817us/step - loss: 11.3727 - accuracy: 0.0024
Epoch 59/200
116/116 [=====] - 0s 861us/step - loss: 11.3787 - accuracy: 0.0011
Epoch 60/200
116/116 [=====] - 0s 973us/step - loss: 11.3731 - accuracy: 0.0016
Epoch 61/200
116/116 [=====] - 0s 2ms/step - loss: 11.3710 - accuracy: 0.0022
Epoch 62/200
116/116 [=====] - 0s 2ms/step - loss: 11.3772 - accuracy: 0.0022
Epoch 63/200
116/116 [=====] - 0s 1ms/step - loss: 11.3720 - accuracy: 0.0022
Epoch 64/200
116/116 [=====] - 0s 957us/step - loss: 11.3718 - accuracy: 0.0014
Epoch 65/200
116/116 [=====] - 0s 2ms/step - loss: 11.3775 - accuracy: 0.0011
Epoch 66/200
116/116 [=====] - 0s 2ms/step - loss: 11.3767 - accuracy: 0.0035
Epoch 67/200
116/116 [=====] - 0s 2ms/step - loss: 11.3719 - accuracy: 0.0019
Epoch 68/200
116/116 [=====] - 0s 2ms/step - loss: 11.3737 - accuracy: 0.0014
Epoch 69/200
116/116 [=====] - 0s 2ms/step - loss: 11.3717 - accuracy: 0.0014
Epoch 70/200
116/116 [=====] - 0s 2ms/step - loss: 11.3704 - accuracy: 0.0030
Epoch 71/200
116/116 [=====] - 0s 2ms/step - loss: 11.3754 - accuracy: 0.0024
Epoch 72/200
116/116 [=====] - 0s 2ms/step - loss: 11.3756 - accuracy: 5.4025e-04
Epoch 73/200
116/116 [=====] - 0s 2ms/step - loss: 11.3701 - accuracy: 0.0022
Epoch 74/200
116/116 [=====] - 0s 2ms/step - loss: 11.3711 - accuracy: 0.0019
Epoch 75/200
116/116 [=====] - 0s 2ms/step - loss: 11.3717 - accuracy: 0.0024
Epoch 76/200
116/116 [=====] - 0s 2ms/step - loss: 11.3722 - accuracy: 0.0016
Epoch 77/200
116/116 [=====] - 0s 2ms/step - loss: 11.3730 - accuracy: 2.7012e-04

Epoch 78/200
116/116 [=====] - 0s 2ms/step - loss: 11.3723 - accuracy: 0.0014
Epoch 79/200
116/116 [=====] - 0s 2ms/step - loss: 11.3739 - accuracy: 5.4025e-04
Epoch 80/200
116/116 [=====] - 0s 2ms/step - loss: 11.3690 - accuracy: 0.0019
Epoch 81/200
116/116 [=====] - 0s 2ms/step - loss: 11.3732 - accuracy: 0.0011
Epoch 82/200
116/116 [=====] - 0s 2ms/step - loss: 11.3704 - accuracy: 0.0016
Epoch 83/200
116/116 [=====] - 0s 2ms/step - loss: 11.3752 - accuracy: 0.0016
Epoch 84/200
116/116 [=====] - 0s 2ms/step - loss: 11.3755 - accuracy: 0.0024
Epoch 85/200
116/116 [=====] - 0s 2ms/step - loss: 11.3772 - accuracy: 0.0011
Epoch 86/200
116/116 [=====] - 0s 2ms/step - loss: 11.3711 - accuracy: 0.0022
Epoch 87/200
116/116 [=====] - 0s 2ms/step - loss: 11.3821 - accuracy: 0.0014
Epoch 88/200
116/116 [=====] - 0s 2ms/step - loss: 11.3673 - accuracy: 0.0019
Epoch 89/200
116/116 [=====] - 0s 817us/step - loss: 11.3724 - accuracy: 0.0016
Epoch 90/200
116/116 [=====] - 0s 852us/step - loss: 11.3743 - accuracy: 8.1037e-04
Epoch 91/200
116/116 [=====] - 0s 791us/step - loss: 11.3689 - accuracy: 0.0022
Epoch 92/200
116/116 [=====] - 0s 870us/step - loss: 11.3748 - accuracy: 0.0046
Epoch 93/200
116/116 [=====] - 0s 817us/step - loss: 11.3719 - accuracy: 0.0032
Epoch 94/200
116/116 [=====] - 0s 809us/step - loss: 11.3753 - accuracy: 0.0027
Epoch 95/200
116/116 [=====] - 0s 826us/step - loss: 11.3733 - accuracy: 0.0032
Epoch 96/200
116/116 [=====] - 0s 826us/step - loss: 11.3718 - accuracy: 0.0032
Epoch 97/200
116/116 [=====] - 0s 826us/step - loss: 11.3737 - accuracy: 0.0019
Epoch 98/200
116/116 [=====] - 0s 870us/step - loss: 11.3747 - accuracy: 0.0011
Epoch 99/200
116/116 [=====] - 0s 818us/step - loss: 11.3713 - accuracy: 0.0016
Epoch 100/200
116/116 [=====] - 0s 861us/step - loss: 11.3720 - accuracy: 0.0016
Epoch 101/200
116/116 [=====] - 0s 817us/step - loss: 11.3724 - accuracy: 0.0011
Epoch 102/200
116/116 [=====] - 0s 826us/step - loss: 11.3727 - accuracy: 0.0022
Epoch 103/200
116/116 [=====] - 0s 896us/step - loss: 11.3733 - accuracy: 0.0011
Epoch 104/200
116/116 [=====] - 0s 870us/step - loss: 11.3725 - accuracy: 0.0027
Epoch 105/200
116/116 [=====] - 0s 878us/step - loss: 11.3727 - accuracy: 0.0027
Epoch 106/200
116/116 [=====] - 0s 1ms/step - loss: 11.3748 - accuracy: 0.0024
Epoch 107/200
116/116 [=====] - 0s 2ms/step - loss: 11.3779 - accuracy: 5.4025e-04
Epoch 108/200
116/116 [=====] - 0s 2ms/step - loss: 11.3768 - accuracy: 5.4025e-04
Epoch 109/200
116/116 [=====] - 0s 1ms/step - loss: 11.3729 - accuracy: 0.0024
Epoch 110/200
116/116 [=====] - 0s 878us/step - loss: 11.3695 - accuracy: 0.0014
Epoch 111/200
116/116 [=====] - 0s 843us/step - loss: 11.3715 - accuracy: 0.0024
Epoch 112/200
116/116 [=====] - 0s 835us/step - loss: 11.3695 - accuracy: 0.0041
Epoch 113/200
116/116 [=====] - 0s 817us/step - loss: 11.3711 - accuracy: 0.0022
Epoch 114/200
116/116 [=====] - 0s 878us/step - loss: 11.3723 - accuracy: 0.0027
Epoch 115/200
116/116 [=====] - 0s 904us/step - loss: 11.3757 - accuracy: 0.0030

Epoch 116/200
116/116 [=====] - 0s 791us/step - loss: 11.3721 - accuracy: 0.0035
Epoch 117/200
116/116 [=====] - 0s 896us/step - loss: 11.3706 - accuracy: 0.0022
Epoch 118/200
116/116 [=====] - 0s 800us/step - loss: 11.3710 - accuracy: 0.0024
Epoch 119/200
116/116 [=====] - 0s 809us/step - loss: 11.3710 - accuracy: 0.0016
Epoch 120/200
116/116 [=====] - 0s 843us/step - loss: 11.3724 - accuracy: 0.0030
Epoch 121/200
116/116 [=====] - 0s 791us/step - loss: 11.3730 - accuracy: 0.0030
Epoch 122/200
116/116 [=====] - 0s 870us/step - loss: 11.3817 - accuracy: 0.0038
Epoch 123/200
116/116 [=====] - 0s 817us/step - loss: 11.3768 - accuracy: 0.0030
Epoch 124/200
116/116 [=====] - 0s 826us/step - loss: 11.3771 - accuracy: 0.0011
Epoch 125/200
116/116 [=====] - 0s 957us/step - loss: 11.3692 - accuracy: 0.0027
Epoch 126/200
116/116 [=====] - 0s 809us/step - loss: 11.3709 - accuracy: 0.0035
Epoch 127/200
116/116 [=====] - 0s 870us/step - loss: 11.3734 - accuracy: 0.0030
Epoch 128/200
116/116 [=====] - 0s 904us/step - loss: 11.3697 - accuracy: 0.0022
Epoch 129/200
116/116 [=====] - 0s 1ms/step - loss: 11.3746 - accuracy: 0.0022
Epoch 130/200
116/116 [=====] - 0s 1ms/step - loss: 11.3739 - accuracy: 0.0022
Epoch 131/200
116/116 [=====] - 0s 1ms/step - loss: 11.3750 - accuracy: 0.0035
Epoch 132/200
116/116 [=====] - 0s 870us/step - loss: 11.3704 - accuracy: 0.0019
Epoch 133/200
116/116 [=====] - 0s 913us/step - loss: 11.3716 - accuracy: 0.0024
Epoch 134/200
116/116 [=====] - 0s 826us/step - loss: 11.3739 - accuracy: 0.0019
Epoch 135/200
116/116 [=====] - 0s 791us/step - loss: 11.3756 - accuracy: 0.0022
Epoch 136/200
116/116 [=====] - 0s 835us/step - loss: 11.3723 - accuracy: 0.0014
Epoch 137/200
116/116 [=====] - 0s 800us/step - loss: 11.3700 - accuracy: 0.0014
Epoch 138/200
116/116 [=====] - 0s 843us/step - loss: 11.3681 - accuracy: 0.0030
Epoch 139/200
116/116 [=====] - 0s 896us/step - loss: 11.3740 - accuracy: 0.0030
Epoch 140/200
116/116 [=====] - 0s 852us/step - loss: 11.3699 - accuracy: 0.0027
Epoch 141/200
116/116 [=====] - 0s 800us/step - loss: 11.3717 - accuracy: 0.0022
Epoch 142/200
116/116 [=====] - 0s 835us/step - loss: 11.3743 - accuracy: 0.0027
Epoch 143/200
116/116 [=====] - 0s 1ms/step - loss: 11.3707 - accuracy: 0.0014
Epoch 144/200
116/116 [=====] - 0s 2ms/step - loss: 11.3697 - accuracy: 0.0027
Epoch 145/200
116/116 [=====] - 0s 974us/step - loss: 11.3719 - accuracy: 0.0011
Epoch 146/200
116/116 [=====] - 0s 870us/step - loss: 11.3706 - accuracy: 0.0027
Epoch 147/200
116/116 [=====] - 0s 843us/step - loss: 11.3733 - accuracy: 0.0024
Epoch 148/200
116/116 [=====] - 0s 1ms/step - loss: 11.3698 - accuracy: 0.0027
Epoch 149/200
116/116 [=====] - 0s 2ms/step - loss: 11.3701 - accuracy: 0.0022
Epoch 150/200
116/116 [=====] - 0s 2ms/step - loss: 11.3723 - accuracy: 0.0016
Epoch 151/200
116/116 [=====] - 0s 2ms/step - loss: 11.3733 - accuracy: 0.0035
Epoch 152/200
116/116 [=====] - 0s 913us/step - loss: 11.3714 - accuracy: 0.0046
Epoch 153/200
116/116 [=====] - 0s 913us/step - loss: 11.3715 - accuracy: 0.0024

Epoch 154/200
116/116 [=====] - 0s 887us/step - loss: 11.3777 - accuracy: 0.0030
Epoch 155/200
116/116 [=====] - 0s 852us/step - loss: 11.3832 - accuracy: 0.0030
Epoch 156/200
116/116 [=====] - 0s 1ms/step - loss: 11.3709 - accuracy: 0.0032
Epoch 157/200
116/116 [=====] - 0s 965us/step - loss: 11.3707 - accuracy: 0.0035
Epoch 158/200
116/116 [=====] - 0s 852us/step - loss: 11.3709 - accuracy: 0.0027
Epoch 159/200
116/116 [=====] - 0s 1ms/step - loss: 11.3719 - accuracy: 0.0027
Epoch 160/200
116/116 [=====] - 0s 904us/step - loss: 11.3712 - accuracy: 0.0019
Epoch 161/200
116/116 [=====] - 0s 826us/step - loss: 11.3719 - accuracy: 0.0022
Epoch 162/200
116/116 [=====] - 0s 1ms/step - loss: 11.3738 - accuracy: 0.0030
Epoch 163/200
116/116 [=====] - 0s 870us/step - loss: 11.3732 - accuracy: 0.0035
Epoch 164/200
116/116 [=====] - 0s 878us/step - loss: 11.3725 - accuracy: 0.0014
Epoch 165/200
116/116 [=====] - 0s 835us/step - loss: 11.3735 - accuracy: 0.0027
Epoch 166/200
116/116 [=====] - 0s 852us/step - loss: 11.3713 - accuracy: 0.0024
Epoch 167/200
116/116 [=====] - 0s 1ms/step - loss: 11.3708 - accuracy: 0.0030
Epoch 168/200
116/116 [=====] - 0s 809us/step - loss: 11.3730 - accuracy: 0.0016
Epoch 169/200
116/116 [=====] - 0s 1ms/step - loss: 11.3760 - accuracy: 0.0032
Epoch 170/200
116/116 [=====] - 0s 2ms/step - loss: 11.3719 - accuracy: 0.0032
Epoch 171/200
116/116 [=====] - 0s 2ms/step - loss: 11.3770 - accuracy: 0.0027
Epoch 172/200
116/116 [=====] - 0s 1ms/step - loss: 11.3710 - accuracy: 0.0030
Epoch 173/200
116/116 [=====] - 0s 1ms/step - loss: 11.3716 - accuracy: 0.0027
Epoch 174/200
116/116 [=====] - 0s 843us/step - loss: 11.3695 - accuracy: 0.0038
Epoch 175/200
116/116 [=====] - 0s 861us/step - loss: 11.3737 - accuracy: 0.0046
Epoch 176/200
116/116 [=====] - 0s 852us/step - loss: 11.3696 - accuracy: 0.0014
Epoch 177/200
116/116 [=====] - 0s 852us/step - loss: 11.3696 - accuracy: 0.0030
Epoch 178/200
116/116 [=====] - 0s 800us/step - loss: 11.3704 - accuracy: 0.0032
Epoch 179/200
116/116 [=====] - 0s 835us/step - loss: 11.3713 - accuracy: 0.0041
Epoch 180/200
116/116 [=====] - 0s 800us/step - loss: 11.3713 - accuracy: 0.0032
Epoch 181/200
116/116 [=====] - 0s 835us/step - loss: 11.3738 - accuracy: 0.0032
Epoch 182/200
116/116 [=====] - 0s 817us/step - loss: 11.3718 - accuracy: 0.0011
Epoch 183/200
116/116 [=====] - 0s 913us/step - loss: 11.3713 - accuracy: 0.0027
Epoch 184/200
116/116 [=====] - 0s 817us/step - loss: 11.3689 - accuracy: 0.0046
Epoch 185/200
116/116 [=====] - 0s 948us/step - loss: 11.3710 - accuracy: 0.0019
Epoch 186/200
116/116 [=====] - 0s 869us/step - loss: 11.3737 - accuracy: 0.0014
Epoch 187/200
116/116 [=====] - 0s 861us/step - loss: 11.3752 - accuracy: 0.0014
Epoch 188/200
116/116 [=====] - 0s 896us/step - loss: 11.3727 - accuracy: 0.0011
Epoch 189/200
116/116 [=====] - 0s 800us/step - loss: 11.3694 - accuracy: 0.0024
Epoch 190/200
116/116 [=====] - 0s 826us/step - loss: 11.3736 - accuracy: 0.0030
Epoch 191/200
116/116 [=====] - 0s 1ms/step - loss: 11.3788 - accuracy: 0.0030

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Epoch 192/200
116/116 [=====] - 0s 2ms/step - loss: 11.3723 - accuracy: 0.0016
Epoch 193/200
116/116 [=====] - 0s 1ms/step - loss: 11.3728 - accuracy: 0.0022
Epoch 194/200
116/116 [=====] - 0s 1ms/step - loss: 11.3690 - accuracy: 0.0024
Epoch 195/200
116/116 [=====] - 0s 835us/step - loss: 11.3725 - accuracy: 0.0019
Epoch 196/200
116/116 [=====] - 0s 843us/step - loss: 11.3741 - accuracy: 0.0019
Epoch 197/200
116/116 [=====] - 0s 835us/step - loss: 11.3711 - accuracy: 0.0011
Epoch 198/200
116/116 [=====] - 0s 817us/step - loss: 11.3736 - accuracy: 0.0000e+00
Epoch 199/200
116/116 [=====] - 0s 817us/step - loss: 11.3713 - accuracy: 0.0019
Epoch 200/200
116/116 [=====] - 0s 870us/step - loss: 11.3695 - accuracy: 0.0014

```

In [57]:

```

# 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: 11.7652 - accuracy: 0.0000e+00 - 97ms/epoch - 2ms/step
Loss: 11.765235900878906, Accuracy: 0.0

```

In [58]:

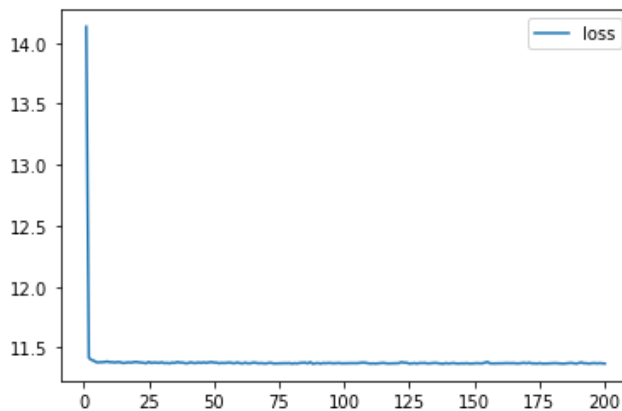
```

# 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[58]: <AxesSubplot:>



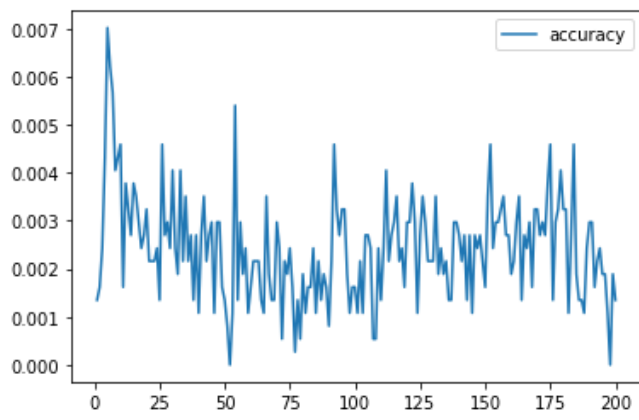
In [59]:

```

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

```

Out[59]: <AxesSubplot:>



In []: