

```
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 [45]: # Define the model - deep neural net
number_input_features = len(X_train[0])
hidden_nodes_layer1 = 40
hidden_nodes_layer2 = 30
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="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_layer2, activation="elu"))

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

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

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

Model: "sequential\_7"

Layer (type)	Output Shape	Param #
=====		
dense_35 (Dense)	(None, 40)	320
dense_36 (Dense)	(None, 30)	1230
dense_37 (Dense)	(None, 30)	930
dense_38 (Dense)	(None, 30)	930
dense_39 (Dense)	(None, 10)	310
=====		
Total params: 3,720		
Trainable params: 3,720		
Non-trainable params: 0		

## Compile the Model

```
In [46]: # Compile the model
nn.compile(loss="mean_squared_error", optimizer="adam", metrics=["accuracy"])
```

## Train the model

```
In [47]: # Train the model
fit_model = nn.fit(X_train,y_train,epochs=200)

Epoch 1/200
116/116 [=====] - 0s 812us/step - loss: 89.3024 - accuracy: 0.0041
```

Epoch 2/200  
116/116 [=====] - 0s 922us/step - loss: 34.7485 - accuracy: 5.4025e-04  
Epoch 3/200  
116/116 [=====] - 0s 1ms/step - loss: 27.4073 - accuracy: 0.0011  
Epoch 4/200  
116/116 [=====] - 0s 1ms/step - loss: 19.1466 - accuracy: 0.0014  
Epoch 5/200  
116/116 [=====] - 0s 2ms/step - loss: 18.8484 - accuracy: 0.0014  
Epoch 6/200  
116/116 [=====] - 0s 1ms/step - loss: 18.8104 - accuracy: 0.0022  
Epoch 7/200  
116/116 [=====] - 0s 852us/step - loss: 18.7229 - accuracy: 0.0035  
Epoch 8/200  
116/116 [=====] - 0s 800us/step - loss: 12.8468 - accuracy: 0.0035  
Epoch 9/200  
116/116 [=====] - 0s 861us/step - loss: 7.0425 - accuracy: 0.0022  
Epoch 10/200  
116/116 [=====] - 0s 800us/step - loss: 6.7282 - accuracy: 0.0041  
Epoch 11/200  
116/116 [=====] - 0s 800us/step - loss: 6.6421 - accuracy: 0.0032  
Epoch 12/200  
116/116 [=====] - 0s 835us/step - loss: 6.5917 - accuracy: 0.0032  
Epoch 13/200  
116/116 [=====] - 0s 809us/step - loss: 6.5462 - accuracy: 0.0030  
Epoch 14/200  
116/116 [=====] - 0s 948us/step - loss: 6.5528 - accuracy: 0.0035  
Epoch 15/200  
116/116 [=====] - 0s 878us/step - loss: 6.5568 - accuracy: 0.0027  
Epoch 16/200  
116/116 [=====] - 0s 826us/step - loss: 6.5538 - accuracy: 0.0027  
Epoch 17/200  
116/116 [=====] - 0s 835us/step - loss: 6.5398 - accuracy: 0.0022  
Epoch 18/200  
116/116 [=====] - 0s 835us/step - loss: 6.5688 - accuracy: 0.0022  
Epoch 19/200  
116/116 [=====] - 0s 983us/step - loss: 6.5710 - accuracy: 0.0030  
Epoch 20/200  
116/116 [=====] - 0s 930us/step - loss: 6.5531 - accuracy: 0.0032  
Epoch 21/200  
116/116 [=====] - 0s 817us/step - loss: 6.5583 - accuracy: 0.0027  
Epoch 22/200  
116/116 [=====] - 0s 948us/step - loss: 6.5420 - accuracy: 0.0022  
Epoch 23/200  
116/116 [=====] - 0s 817us/step - loss: 6.5129 - accuracy: 0.0041  
Epoch 24/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5600 - accuracy: 0.0019  
Epoch 25/200  
116/116 [=====] - 0s 957us/step - loss: 6.5526 - accuracy: 0.0041  
Epoch 26/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5451 - accuracy: 0.0027  
Epoch 27/200  
116/116 [=====] - 0s 3ms/step - loss: 6.6209 - accuracy: 0.0038  
Epoch 28/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5343 - accuracy: 0.0027  
Epoch 29/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5277 - accuracy: 0.0022  
Epoch 30/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5769 - accuracy: 0.0014  
Epoch 31/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5764 - accuracy: 0.0038  
Epoch 32/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5123 - accuracy: 0.0022  
Epoch 33/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5047 - accuracy: 0.0019  
Epoch 34/200  
116/116 [=====] - 0s 870us/step - loss: 6.5119 - accuracy: 0.0027  
Epoch 35/200  
116/116 [=====] - 0s 861us/step - loss: 6.5125 - accuracy: 0.0019  
Epoch 36/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5595 - accuracy: 0.0027  
Epoch 37/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5138 - accuracy: 0.0019  
Epoch 38/200  
116/116 [=====] - 0s 965us/step - loss: 6.5263 - accuracy: 0.0022  
Epoch 39/200  
116/116 [=====] - 0s 939us/step - loss: 6.5293 - accuracy: 0.0024

Epoch 40/200  
116/116 [=====] - 0s 800us/step - loss: 6.5456 - accuracy: 0.0030  
Epoch 41/200  
116/116 [=====] - 0s 896us/step - loss: 6.5144 - accuracy: 0.0014  
Epoch 42/200  
116/116 [=====] - 0s 878us/step - loss: 6.5259 - accuracy: 0.0030  
Epoch 43/200  
116/116 [=====] - 0s 878us/step - loss: 6.5153 - accuracy: 0.0027  
Epoch 44/200  
116/116 [=====] - 0s 1ms/step - loss: 6.4837 - accuracy: 0.0019  
Epoch 45/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5373 - accuracy: 0.0027  
Epoch 46/200  
116/116 [=====] - 0s 2ms/step - loss: 6.5055 - accuracy: 0.0022  
Epoch 47/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5135 - accuracy: 0.0011  
Epoch 48/200  
116/116 [=====] - 0s 904us/step - loss: 6.5463 - accuracy: 0.0022  
Epoch 49/200  
116/116 [=====] - 0s 817us/step - loss: 6.5086 - accuracy: 0.0014  
Epoch 50/200  
116/116 [=====] - 0s 1ms/step - loss: 6.5330 - accuracy: 0.0032  
Epoch 51/200  
116/116 [=====] - 0s 861us/step - loss: 6.5001 - accuracy: 0.0022  
Epoch 52/200  
116/116 [=====] - 0s 896us/step - loss: 6.5015 - accuracy: 0.0014  
Epoch 53/200  
116/116 [=====] - 0s 861us/step - loss: 6.4923 - accuracy: 0.0014  
Epoch 54/200  
116/116 [=====] - 0s 1ms/step - loss: 3.0818 - accuracy: 0.0024  
Epoch 55/200  
116/116 [=====] - 0s 1ms/step - loss: 0.4195 - accuracy: 0.0030  
Epoch 56/200  
116/116 [=====] - 0s 843us/step - loss: 0.4056 - accuracy: 0.0019  
Epoch 57/200  
116/116 [=====] - 0s 817us/step - loss: 0.3984 - accuracy: 0.0019  
Epoch 58/200  
116/116 [=====] - 0s 843us/step - loss: 0.4004 - accuracy: 0.0019  
Epoch 59/200  
116/116 [=====] - 0s 887us/step - loss: 0.4024 - accuracy: 0.0019  
Epoch 60/200  
116/116 [=====] - 0s 861us/step - loss: 0.3902 - accuracy: 0.0016  
Epoch 61/200  
116/116 [=====] - 0s 843us/step - loss: 0.3799 - accuracy: 0.0016  
Epoch 62/200  
116/116 [=====] - 0s 887us/step - loss: 0.3784 - accuracy: 0.0019  
Epoch 63/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3839 - accuracy: 0.0014  
Epoch 64/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3738 - accuracy: 8.1037e-04  
Epoch 65/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3842 - accuracy: 0.0030  
Epoch 66/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3836 - accuracy: 0.0016  
Epoch 67/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3674 - accuracy: 0.0014  
Epoch 68/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3709 - accuracy: 0.0019  
Epoch 69/200  
116/116 [=====] - 0s 904us/step - loss: 0.3676 - accuracy: 0.0016  
Epoch 70/200  
116/116 [=====] - 0s 817us/step - loss: 0.3756 - accuracy: 0.0016  
Epoch 71/200  
116/116 [=====] - 0s 983us/step - loss: 0.3702 - accuracy: 0.0019  
Epoch 72/200  
116/116 [=====] - 0s 896us/step - loss: 0.3737 - accuracy: 5.4025e-04  
Epoch 73/200  
116/116 [=====] - 0s 843us/step - loss: 0.3729 - accuracy: 0.0014  
Epoch 74/200  
116/116 [=====] - 0s 930us/step - loss: 0.3841 - accuracy: 8.1037e-04  
Epoch 75/200  
116/116 [=====] - 0s 939us/step - loss: 0.3757 - accuracy: 0.0016  
Epoch 76/200  
116/116 [=====] - 0s 913us/step - loss: 0.3700 - accuracy: 0.0019  
Epoch 77/200  
116/116 [=====] - 0s 852us/step - loss: 0.3778 - accuracy: 0.0014

Epoch 78/200  
116/116 [=====] - 0s 913us/step - loss: 0.3674 - accuracy: 0.0016  
Epoch 79/200  
116/116 [=====] - 0s 870us/step - loss: 0.3761 - accuracy: 0.0022  
Epoch 80/200  
116/116 [=====] - 0s 904us/step - loss: 0.3766 - accuracy: 5.4025e-04  
Epoch 81/200  
116/116 [=====] - 0s 870us/step - loss: 0.3597 - accuracy: 0.0027  
Epoch 82/200  
116/116 [=====] - 0s 870us/step - loss: 0.3665 - accuracy: 0.0019  
Epoch 83/200  
116/116 [=====] - 0s 835us/step - loss: 0.3746 - accuracy: 0.0022  
Epoch 84/200  
116/116 [=====] - 0s 817us/step - loss: 0.3854 - accuracy: 0.0016  
Epoch 85/200  
116/116 [=====] - 0s 966us/step - loss: 0.3666 - accuracy: 0.0030  
Epoch 86/200  
116/116 [=====] - 0s 826us/step - loss: 0.3609 - accuracy: 0.0038  
Epoch 87/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3617 - accuracy: 0.0024  
Epoch 88/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3656 - accuracy: 0.0011  
Epoch 89/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3673 - accuracy: 0.0027  
Epoch 90/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3652 - accuracy: 0.0022  
Epoch 91/200  
116/116 [=====] - 0s 861us/step - loss: 0.3665 - accuracy: 0.0022  
Epoch 92/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3706 - accuracy: 5.4025e-04  
Epoch 93/200  
116/116 [=====] - 0s 843us/step - loss: 0.3666 - accuracy: 0.0011  
Epoch 94/200  
116/116 [=====] - 0s 965us/step - loss: 0.3690 - accuracy: 0.0014  
Epoch 95/200  
116/116 [=====] - 0s 913us/step - loss: 0.3700 - accuracy: 0.0027  
Epoch 96/200  
116/116 [=====] - 0s 843us/step - loss: 0.3592 - accuracy: 0.0022  
Epoch 97/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3627 - accuracy: 0.0022  
Epoch 98/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3586 - accuracy: 0.0024  
Epoch 99/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3516 - accuracy: 0.0027  
Epoch 100/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3630 - accuracy: 0.0016  
Epoch 101/200  
116/116 [=====] - 0s 974us/step - loss: 0.3565 - accuracy: 0.0022  
Epoch 102/200  
116/116 [=====] - 0s 939us/step - loss: 0.3705 - accuracy: 0.0030  
Epoch 103/200  
116/116 [=====] - 0s 922us/step - loss: 0.3682 - accuracy: 0.0022  
Epoch 104/200  
116/116 [=====] - 0s 922us/step - loss: 0.3538 - accuracy: 0.0019  
Epoch 105/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3650 - accuracy: 0.0019  
Epoch 106/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3626 - accuracy: 0.0014  
Epoch 107/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3508 - accuracy: 0.0024  
Epoch 108/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3610 - accuracy: 0.0035  
Epoch 109/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3585 - accuracy: 0.0022  
Epoch 110/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3541 - accuracy: 0.0030  
Epoch 111/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3548 - accuracy: 0.0027  
Epoch 112/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3552 - accuracy: 0.0014  
Epoch 113/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3641 - accuracy: 0.0016  
Epoch 114/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3557 - accuracy: 0.0024  
Epoch 115/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3560 - accuracy: 0.0027

Epoch 116/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3573 - accuracy: 0.0035  
Epoch 117/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3617 - accuracy: 0.0022  
Epoch 118/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3534 - accuracy: 0.0032  
Epoch 119/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3595 - accuracy: 0.0016  
Epoch 120/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3587 - accuracy: 0.0011  
Epoch 121/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3500 - accuracy: 0.0024  
Epoch 122/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3596 - accuracy: 0.0035  
Epoch 123/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3548 - accuracy: 0.0035  
Epoch 124/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3557 - accuracy: 0.0024  
Epoch 125/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3517 - accuracy: 0.0038  
Epoch 126/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3566 - accuracy: 0.0019  
Epoch 127/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3530 - accuracy: 0.0022  
Epoch 128/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3548 - accuracy: 0.0019  
Epoch 129/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3615 - accuracy: 0.0019  
Epoch 130/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3505 - accuracy: 0.0027  
Epoch 131/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3548 - accuracy: 0.0027  
Epoch 132/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3557 - accuracy: 0.0014  
Epoch 133/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3499 - accuracy: 0.0027  
Epoch 134/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3499 - accuracy: 0.0019  
Epoch 135/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3523 - accuracy: 0.0019  
Epoch 136/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3552 - accuracy: 0.0030  
Epoch 137/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3483 - accuracy: 0.0022  
Epoch 138/200  
116/116 [=====] - 0s 809us/step - loss: 0.3577 - accuracy: 0.0019  
Epoch 139/200  
116/116 [=====] - 0s 861us/step - loss: 0.3537 - accuracy: 0.0019  
Epoch 140/200  
116/116 [=====] - 0s 913us/step - loss: 0.3540 - accuracy: 0.0035  
Epoch 141/200  
116/116 [=====] - 0s 861us/step - loss: 0.3573 - accuracy: 0.0019  
Epoch 142/200  
116/116 [=====] - 0s 974us/step - loss: 0.3563 - accuracy: 0.0019  
Epoch 143/200  
116/116 [=====] - 0s 870us/step - loss: 0.3524 - accuracy: 0.0016  
Epoch 144/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3663 - accuracy: 0.0027  
Epoch 145/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3491 - accuracy: 5.4025e-04  
Epoch 146/200  
116/116 [=====] - 0s 904us/step - loss: 0.3510 - accuracy: 0.0014  
Epoch 147/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3513 - accuracy: 0.0019  
Epoch 148/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3481 - accuracy: 0.0016  
Epoch 149/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3471 - accuracy: 0.0024  
Epoch 150/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3488 - accuracy: 0.0011  
Epoch 151/200  
116/116 [=====] - 0s 930us/step - loss: 0.3515 - accuracy: 0.0019  
Epoch 152/200  
116/116 [=====] - 0s 841us/step - loss: 0.3494 - accuracy: 5.4025e-04  
Epoch 153/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3490 - accuracy: 0.0027

Epoch 154/200  
116/116 [=====] - 0s 843us/step - loss: 0.3464 - accuracy: 0.0030  
Epoch 155/200  
116/116 [=====] - 0s 817us/step - loss: 0.3501 - accuracy: 0.0030  
Epoch 156/200  
116/116 [=====] - 0s 922us/step - loss: 0.3502 - accuracy: 0.0014  
Epoch 157/200  
116/116 [=====] - 0s 904us/step - loss: 0.3490 - accuracy: 0.0027  
Epoch 158/200  
116/116 [=====] - 0s 930us/step - loss: 0.3527 - accuracy: 0.0024  
Epoch 159/200  
116/116 [=====] - 0s 809us/step - loss: 0.3579 - accuracy: 0.0030  
Epoch 160/200  
116/116 [=====] - 0s 974us/step - loss: 0.3502 - accuracy: 0.0032  
Epoch 161/200  
116/116 [=====] - 0s 835us/step - loss: 0.3522 - accuracy: 0.0019  
Epoch 162/200  
116/116 [=====] - 0s 861us/step - loss: 0.3501 - accuracy: 0.0027  
Epoch 163/200  
116/116 [=====] - 0s 991us/step - loss: 0.3500 - accuracy: 0.0035  
Epoch 164/200  
116/116 [=====] - 0s 852us/step - loss: 0.3547 - accuracy: 0.0027  
Epoch 165/200  
116/116 [=====] - 0s 904us/step - loss: 0.3606 - accuracy: 0.0016  
Epoch 166/200  
116/116 [=====] - 0s 974us/step - loss: 0.3477 - accuracy: 0.0027  
Epoch 167/200  
116/116 [=====] - 0s 957us/step - loss: 0.3474 - accuracy: 0.0022  
Epoch 168/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3574 - accuracy: 0.0043  
Epoch 169/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3484 - accuracy: 0.0041  
Epoch 170/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3584 - accuracy: 0.0027  
Epoch 171/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3498 - accuracy: 0.0035  
Epoch 172/200  
116/116 [=====] - 0s 904us/step - loss: 0.3520 - accuracy: 0.0030  
Epoch 173/200  
116/116 [=====] - 0s 817us/step - loss: 0.3452 - accuracy: 0.0046  
Epoch 174/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3503 - accuracy: 0.0041  
Epoch 175/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3478 - accuracy: 0.0035  
Epoch 176/200  
116/116 [=====] - 0s 922us/step - loss: 0.3499 - accuracy: 0.0032  
Epoch 177/200  
116/116 [=====] - 0s 870us/step - loss: 0.3499 - accuracy: 0.0030  
Epoch 178/200  
116/116 [=====] - 0s 843us/step - loss: 0.3446 - accuracy: 0.0030  
Epoch 179/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3470 - accuracy: 0.0027  
Epoch 180/200  
116/116 [=====] - 0s 843us/step - loss: 0.3472 - accuracy: 0.0032  
Epoch 181/200  
116/116 [=====] - 0s 835us/step - loss: 0.3480 - accuracy: 0.0046  
Epoch 182/200  
116/116 [=====] - 0s 843us/step - loss: 0.3471 - accuracy: 0.0019  
Epoch 183/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3462 - accuracy: 0.0032  
Epoch 184/200  
116/116 [=====] - 0s 904us/step - loss: 0.3519 - accuracy: 0.0027  
Epoch 185/200  
116/116 [=====] - 0s 800us/step - loss: 0.3466 - accuracy: 0.0027  
Epoch 186/200  
116/116 [=====] - 0s 878us/step - loss: 0.3524 - accuracy: 0.0051  
Epoch 187/200  
116/116 [=====] - 0s 913us/step - loss: 0.3480 - accuracy: 0.0049  
Epoch 188/200  
116/116 [=====] - 0s 835us/step - loss: 0.3462 - accuracy: 0.0046  
Epoch 189/200  
116/116 [=====] - 0s 835us/step - loss: 0.3445 - accuracy: 0.0038  
Epoch 190/200  
116/116 [=====] - 0s 1ms/step - loss: 0.3475 - accuracy: 0.0035  
Epoch 191/200  
116/116 [=====] - 0s 2ms/step - loss: 0.3472 - accuracy: 0.0046

```

Epoch 192/200
116/116 [=====] - 0s 2ms/step - loss: 0.3535 - accuracy: 0.0046
Epoch 193/200
116/116 [=====] - 0s 817us/step - loss: 0.3487 - accuracy: 0.0030
Epoch 194/200
116/116 [=====] - 0s 844us/step - loss: 0.3460 - accuracy: 0.0022
Epoch 195/200
116/116 [=====] - 0s 1ms/step - loss: 0.3447 - accuracy: 0.0030
Epoch 196/200
116/116 [=====] - 0s 1ms/step - loss: 0.3513 - accuracy: 0.0035
Epoch 197/200
116/116 [=====] - 0s 904us/step - loss: 0.3477 - accuracy: 0.0035
Epoch 198/200
116/116 [=====] - 0s 1ms/step - loss: 0.3515 - accuracy: 0.0041
Epoch 199/200
116/116 [=====] - 0s 1ms/step - loss: 0.3476 - accuracy: 0.0035
Epoch 200/200
116/116 [=====] - 0s 896us/step - loss: 0.3466 - accuracy: 0.0032

```

In [48]:

```

# 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: 3.5081 - accuracy: 0.0057 - 103ms/epoch - 3ms/step
Loss: 3.5080618858337402, Accuracy: 0.005668016150593758

```

In [49]:

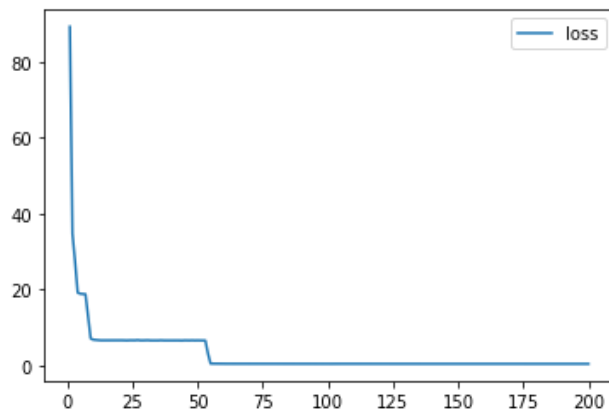
```

# 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[49]: <AxesSubplot:>



In [50]:

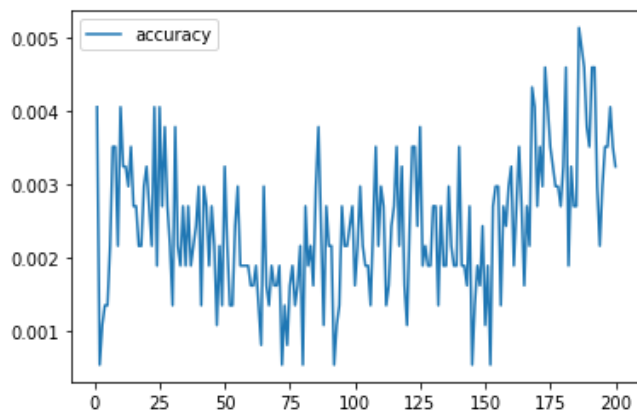
```

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

```

Out[50]: <AxesSubplot:>





In [ ]: