

```

In [1]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error
from sklearn.model_selection import GridSearchCV
import tensorflow as tf
from tensorflow.keras.wrappers.scikit_learn import KerasRegressor
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.optimizers import Adam
from tensorflow.keras import regularizers

# Step 1: Data Preprocessing
data = pd.read_csv('Seasons_Stats.csv') # Replace with the path to your data

# Drop unnecessary columns
data = data.drop(['blank1', 'blank2'], axis=1)

# Handle missing values
data = data.dropna()

# Normalize numerical features
scaler = MinMaxScaler()
column_names = list(data.columns)
numerical_features = column_names[4:]
data[numerical_features] = scaler.fit_transform(data[numerical_features])

# Step 2: Feature Selection
selected_features = column_names[4:]
target_variable = 'PER'

# Step 3: Model Construction
X = data[selected_features]
y = data[target_variable]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

def create_model():
    model = Sequential()
    model.add(Dense(32, activation='relu', input_shape=(len(selected_features),)))
    model.add(Dense(16, activation='relu', kernel_regularizer=regularizers.l2(0.01)))
    model.add(Dense(1, activation='linear'))
    optimizer = Adam(learning_rate=0.001)
    model.compile(optimizer=optimizer, loss='mean_squared_error')
    return model

model = create_model()

# Step 4: Model Training
learning_rate = 0.001
optimizer = Adam(learning_rate=learning_rate)
model.compile(optimizer=optimizer, loss='mean_squared_error')

```

```

epochs = 100
batch_size = 32

model.fit(X_train, y_train, epochs=epochs, batch_size=batch_size, verbose=1)

history = model.fit(X_train, y_train, epochs=epochs, batch_size=batch_size,

# Step 5: Model Evaluation
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print('Mean Squared Error:', mse)

# Step 6: Hyperparameter Tuning
param_grid = {'epochs': [50, 100, 150],
              'batch_size': [32, 64, 128]}

model = KerasRegressor(build_fn=create_model, verbose=0)
grid_search = GridSearchCV(model, param_grid, cv=3)
grid_search.fit(X_train, y_train)

best_params = grid_search.best_params_
print('Best Parameters:', best_params)

best_model = grid_search.best_estimator_
y_pred_best = best_model.predict(X_test)
mse_best = mean_squared_error(y_test, y_pred_best)
print('Best Model Mean Squared Error:', mse_best)

```

Metal device set to: Apple M2

Epoch 1/100

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s.cc:128] Failed to get CPU frequency: 0 Hz

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Mean Squared Error: 5.17550409548232e-05

```

```

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y:74: DeprecationWarning: KerasRegressor is deprecated, use Sci-Keras (http
s://github.com/adriangb/scikeras) instead. See https://www.adriangb.com/scike
ras/stable/migration.html for help migrating.

```

```

    model = KerasRegressor(build_fn=create_model, verbose=0)

```

```

Best Parameters: {'batch_size': 32, 'epochs': 150}

```

```

Best Model Mean Squared Error: 7.17525373581752e-05

```

In [2]: `import matplotlib.pyplot as plt`

```

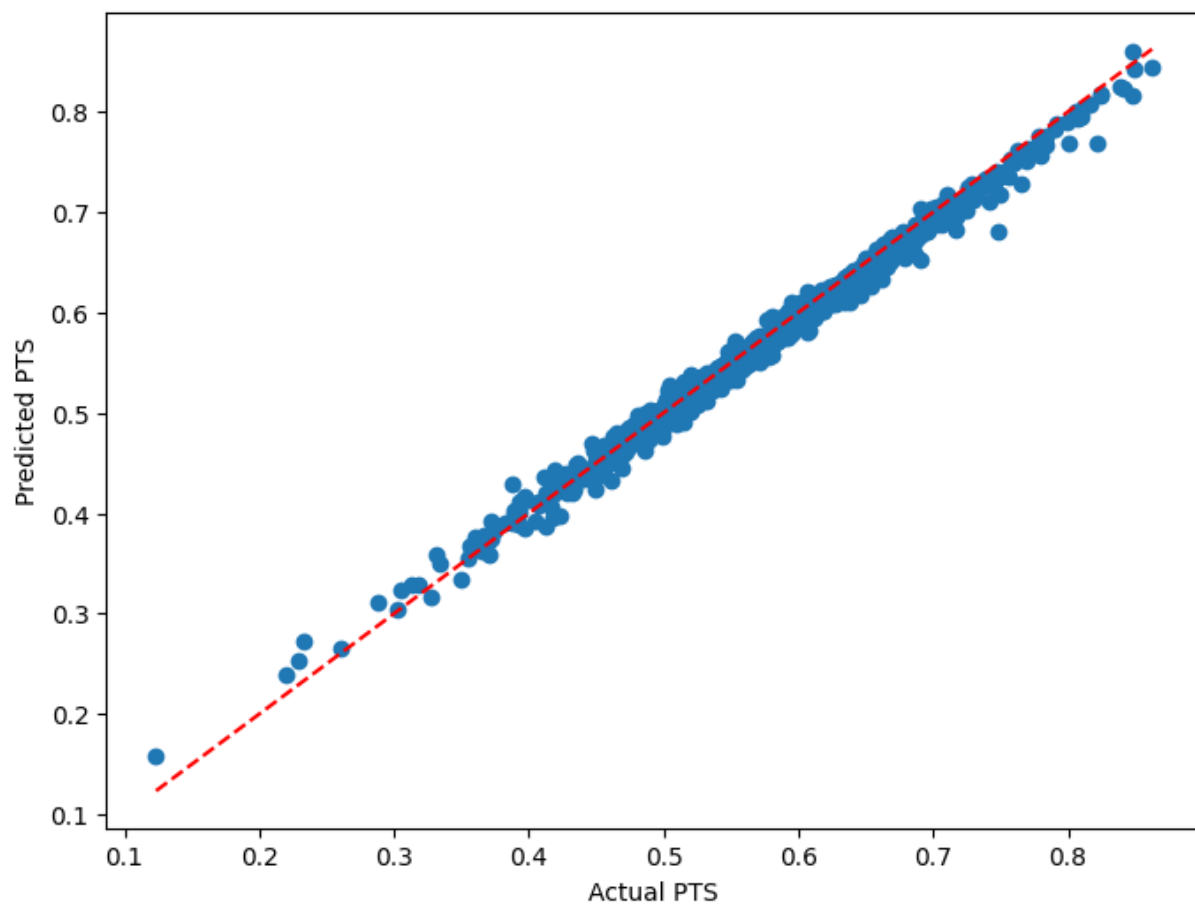
# Scatter plot of actual vs predicted values
plt.figure(figsize=(8, 6))
plt.scatter(y_test, y_pred_best)
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--')
plt.xlabel('Actual PER')
plt.ylabel('Predicted PER')
plt.title('Actual vs Predicted PER')
plt.show()

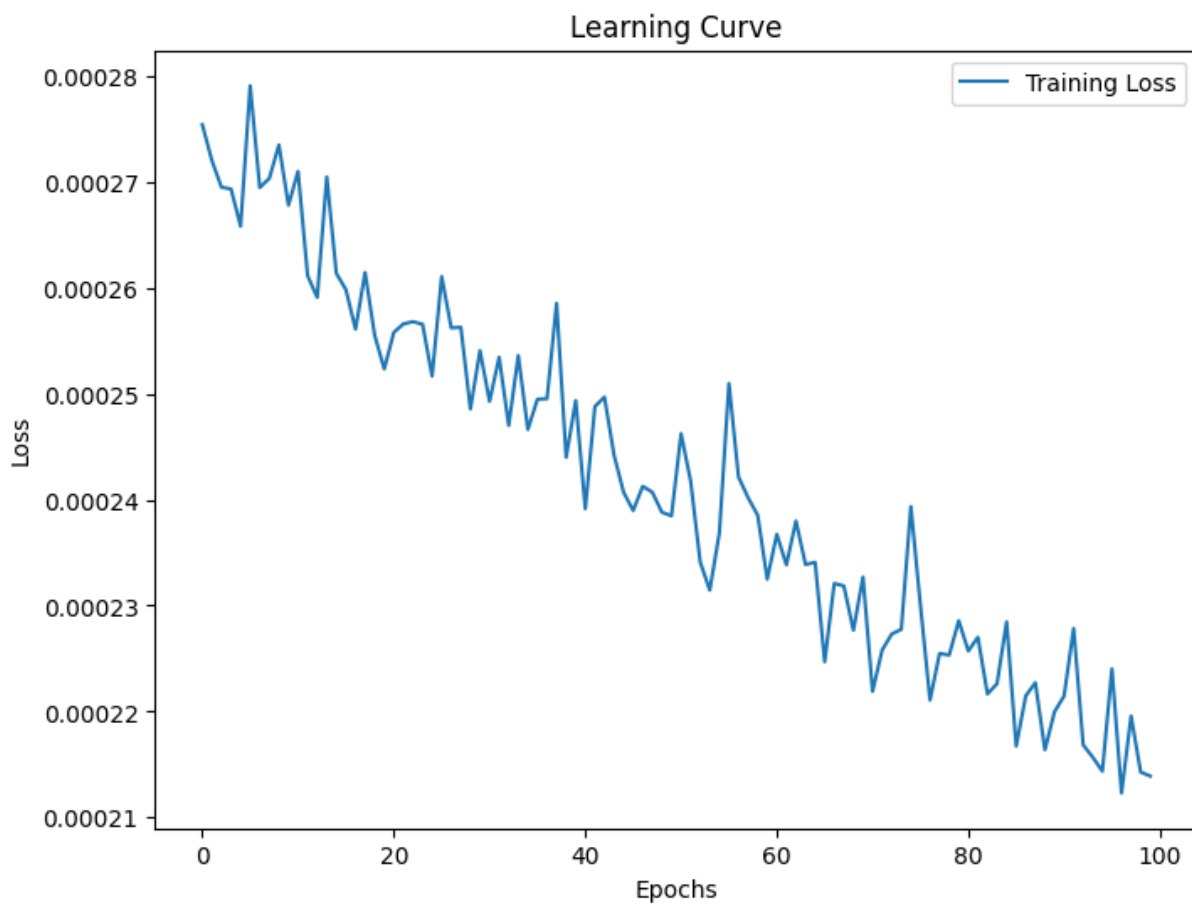
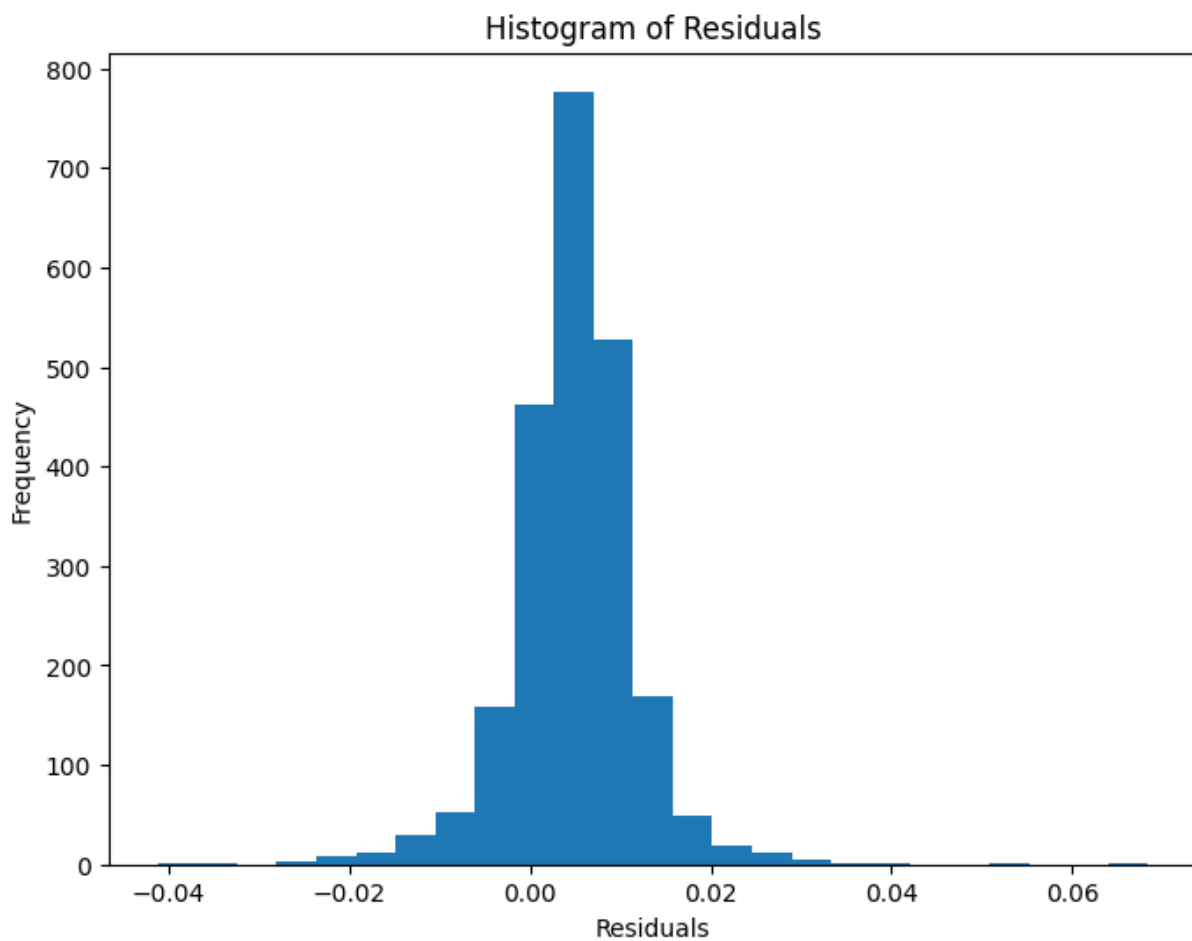
# Histogram of residuals
residuals = y_test - y_pred_best
plt.figure(figsize=(8, 6))
plt.hist(residuals, bins=25)
plt.xlabel('Residuals')
plt.ylabel('Frequency')
plt.title('Histogram of Residuals')
plt.show()

plt.figure(figsize=(8, 6))
plt.plot(history.history['loss'], label='Training Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Learning Curve')
plt.legend()
plt.show()

```

Actual vs Predicted PTS





```
In [3]: import pandas as pd

results = {'Model': ['Base Model', 'Best Model'],
           'MSE': [mse, mse_best]}
results_df = pd.DataFrame(results)

print(results_df)
```

	Model	MSE
0	Base Model	0.000052
1	Best Model	0.000072

```
In [ ]:
```