# pca\_open\_office

January 18, 2022

# 1 Principal Component Analysis

Below are the Principal Component Analysis on the first 7 features:

- 1. index: Index for each data point
- 2. sim: Index for each simulation
- 3. tem: Temperature of air supplied to the zone (°C)
- 4. mass: Total mass flow rate of air supply to the zone (kg/s)
- 5. n occ: Number of occupants in zone
- 6. x: location of occupant of interest on x-axis (m)
- 7. y: location of occupant of interest on y-axis (m)

The last 3 features are treated as target predictions for evaluation purpose.

- 1. MRT: Mean radiant temperature of the occupant of interest's head and chest (°C)
- 2. T: Average temperature of air surrounding the occupant of interest (°C)
- 3. V: Average speed of air surrounding the occupant of interest occupant (°C)

```
[1]: data_loc = '../data/preprocessed/'
```

# 1.1 Open Office

## 1.1.1 1. Standardize the data

```
[2]: import pandas as pd
from sklearn.preprocessing import StandardScaler

open_office_data = pd.read_csv(data_loc+"open_office.csv")
open_office_data.isna().sum()
```

```
[2]: index
                0
                0
      sim
      tem
                0
                0
     mass
                0
     n
      occ
                0
                0
      х
                0
     У
                0
     MRT
     Τ
                0
```

V 8 dtype: int64

```
This dataset seems to have null values.
[3]: # Remove the index column
     open_office_data.drop(columns=['index', 'sim'], inplace=True)
     # Remove null data
     open_office_data = open_office_data.dropna()
     # Seperate the data into features and labels
     X = open_office_data.drop(columns=['MRT', 'T', 'V'])
     y_MRT = open_office_data['MRT']
     y_T = open_office_data['T']
     y_V = open_office_data['V']
     # Standardize the data
     scaler = StandardScaler()
     scaler.fit(X)
     X_std = pd.DataFrame(scaler.transform(X), columns=X.columns)
     X_std.describe()
[3]:
                     tem
                                  mass
                                                               0.00
     count 4.871000e+03 4.871000e+03 4.871000e+03 4.871000e+03 4.871000e+03
            7.243687e-16 -7.691104e-16 5.969813e-16 -2.552031e-15 1.479325e-15
    mean
            1.000103e+00 1.000103e+00 1.000103e+00 1.000103e+00 1.000103e+00
     std
           -1.870995e+00 -1.405293e+00 -1.567282e+00 -1.414649e+00 -1.415396e+00
    min
     25%
           -7.598156e-01 -8.609269e-01 -7.994826e-01 -1.414649e+00 -7.033713e-01
```

```
7.356918e-02 -9.881505e-02 -3.168297e-02 7.068890e-01 6.777462e-02
50%
75%
       6.291590e-01 8.810431e-01 7.361167e-01 7.068890e-01 8.363501e-01
       2.573724e+00 1.860901e+00 1.503916e+00 7.068890e-01 1.532952e+00
max
count 4.871000e+03
mean
       2.411744e-15
std
       1.000103e+00
     -1.117982e+00
min
25%
     -1.117982e+00
50%
      8.944685e-01
75%
      8.944685e-01
      8.944685e-01
max
```

#### 1.1.2 2. Perform PCA

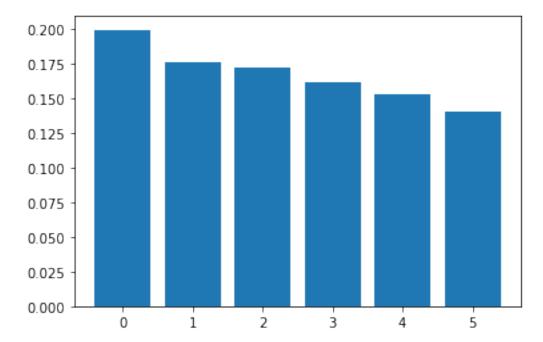
```
[4]: # Perform PCA
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
import numpy as np
from numpy import savetxt

pca = PCA()
pca.fit(X_std)

e_vectors = pca.components_ # The eigenvectors
evr = pca.explained_variance_ratio_ # The variance explained by each eigenvector

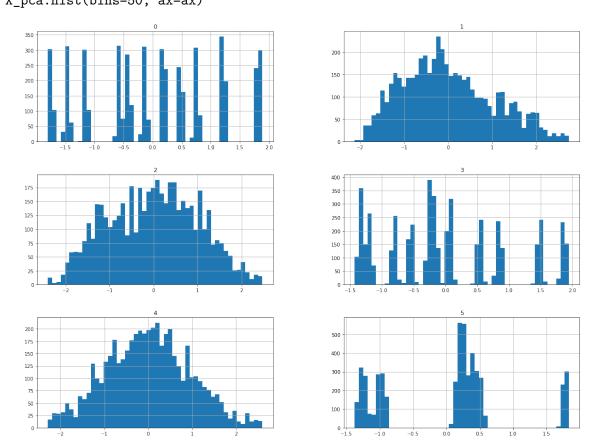
# Save the eigenvectors to a text file
savetxt("../reports/data/pca_open_office_eigenvectors.csv", e_vectors,u
delimiter=',')

plt.bar(range(len(evr)), evr)
plt.savefig('../figures/evr_pca_open_office.png')
```



```
[5]: # Tramsform the data
X_pca = pd.DataFrame(pca.transform(X_std))
fig, ax = plt.subplots(figsize=(20,15))
X_pca.hist(bins=50, ax=ax)
fig.savefig('../figures/hist_pca_open_office.png')
```

/tmp/ipykernel\_21455/4076575879.py:4: UserWarning: To output multiple subplots,
the figure containing the passed axes is being cleared
 X\_pca.hist(bins=50, ax=ax)



After performing PCA on the features, the distribution graph shows more normal distribution features (8 features).

# 1.1.3 3. Compare models' results with and without PCA on predicting MRT

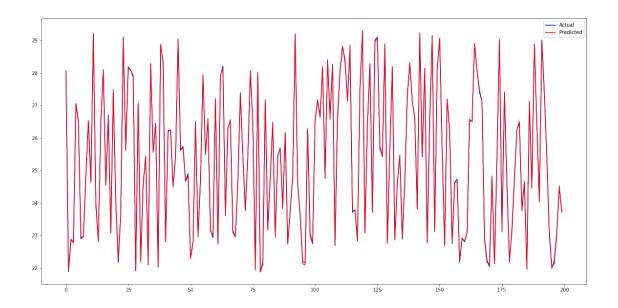
# 3.1 Split the data into training and testing

# 3.2 Train and evaluate the Linear Regression model

## 3.2.1 Without PCA

```
[7]: from sklearn.linear_model import LinearRegression
     from sklearn.metrics import mean_squared_error
     import pickle
     import time
     # Train linear regression model on training data
     model = LinearRegression()
     start = time.time()
     model.fit(X_train, y_train)
     stop = time.time()
     lr_train_time = stop - start
     print(f"Training time: {lr_train_time}s")
     # Save the model to a pickle file
     filename = '../reports/models/lr_MRT_open_office_model.pkl'
     pickle.dump(model, open(filename, 'wb'))
     # Predict on test data
     y_pred = model.predict(X_test)
     # Evaluate the model
     mse = mean_squared_error(y_test, y_pred)
     print(f'Mean squared error: {mse}')
     # Plot the predictions and actual values
     plt.figure(figsize=(20,10))
     plt.plot(range(200), y_test[:200], color='blue', label='Actual')
     plt.plot(range(200), y_pred[:200], color='red', label='Predicted')
     plt.legend()
     plt.show()
```

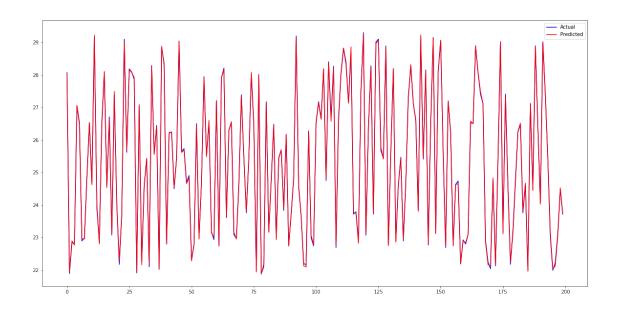
Training time: 0.005346536636352539s Mean squared error: 0.002502969797609541



# **3.2.2** With PCA

```
[8]: # Train linear regression model on training data
     start = time.time()
     model.fit(X_train_pca, y_train_pca)
     stop = time.time()
     lr_pca_train_time = stop - start
     print(f"Training time: {lr_pca_train_time}s")
     # Predict on test data
     y_pred_pca = model.predict(X_test_pca)
     # Save the model to a pickle file
     filename = '../reports/models/lr_MRT_pca_open_office_model.pkl'
     pickle.dump(model, open(filename, 'wb'))
     # Evaluate the model
     mse = mean_squared_error(y_test_pca, y_pred_pca)
     print(f'Mean squared error: {mse}')
     # Plot the predictions and actual values
     plt.figure(figsize=(20,10))
     plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
     plt.plot(range(200), y_pred_pca[:200], color='red', label='Predicted')
     plt.legend()
    plt.show()
```

Training time: 0.007262229919433594s
Mean squared error: 0.0025029697976095343



# 3.3 Train and evaluate the Feedforward Neural Network model

```
[9]: # Create neural network model
     from keras.models import Sequential
     from keras.layers import Dense
     N_NEURONS = 1024
     N_LAYERS = 4
    model = Sequential()
     model.add(Dense(units=N_NEURONS, input_dim=X.shape[1], activation='relu'))
     for i in range(N_LAYERS-1):
         model.add(Dense(units=N_NEURONS, activation='relu'))
     model.add(Dense(units=1, activation='linear')) # Output layer
    model.compile(loss='mean_squared_error', optimizer='adam')
     # Train the model
     start = time.time()
     model.fit(X_train, y_train, epochs=100, verbose=0)
     stop = time.time()
     nn_train_time = stop - start
     print(f"Training time: {nn_train_time}s")
     # Predict on test data
     y_pred_nn = model.predict(X_test)
     model.save('../reports/models/nn_MRT_open_office_model.pkl')
```

```
# Evaluate the model
mse = mean_squared_error(y_test, y_pred_nn)
print(f'Mean squared error: {mse}')
# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred_nn[:200], color='red', label='Predicted')
plt.legend()
plt.show()
2022-01-18 18:10:58.452571: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcudart.so.10.1
2022-01-18 18:11:00.235584: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcuda.so.1
2022-01-18 18:11:00.277612: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.278207: I
tensorflow/core/common runtime/gpu/gpu device.cc:1716] Found device 0 with
pciBusID: 0000:04:00.0 name: GeForce GTX 1060 6GB computeCapability: 6.1
coreClock: 1.7845GHz coreCount: 10 deviceMemorySize: 5.93GiB
deviceMemoryBandwidth: 178.99GiB/s
2022-01-18 18:11:00.278261: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcudart.so.10.1
2022-01-18 18:11:00.280817: I
tensorflow/stream executor/platform/default/dso loader.cc:48] Successfully
opened dynamic library libcublas.so.10
2022-01-18 18:11:00.282432: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcufft.so.10
2022-01-18 18:11:00.282790: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcurand.so.10
2022-01-18 18:11:00.286776: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcusolver.so.10
2022-01-18 18:11:00.288329: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcusparse.so.10
2022-01-18 18:11:00.293957: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
```

```
opened dynamic library libcudnn.so.7
2022-01-18 18:11:00.294147: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.294814: I
tensorflow/stream executor/cuda/cuda gpu executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.295351: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1858] Adding visible gpu
devices: 0
2022-01-18 18:11:00.309030: I
tensorflow/core/platform/profile_utils/cpu_utils.cc:104] CPU Frequency:
2022-01-18 18:11:00.311605: I tensorflow/compiler/xla/service/service.cc:168]
XLA service 0x563e63c88e40 initialized for platform Host (this does not
guarantee that XLA will be used). Devices:
2022-01-18 18:11:00.311652: I tensorflow/compiler/xla/service/service.cc:176]
StreamExecutor device (0): Host, Default Version
2022-01-18 18:11:00.557467: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.558124: I tensorflow/compiler/xla/service/service.cc:168]
XLA service 0x563e63cf48f0 initialized for platform CUDA (this does not
guarantee that XLA will be used). Devices:
2022-01-18 18:11:00.558147: I tensorflow/compiler/xla/service/service.cc:176]
StreamExecutor device (0): GeForce GTX 1060 6GB, Compute Capability 6.1
2022-01-18 18:11:00.558401: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.558958: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1716] Found device 0 with
properties:
pciBusID: 0000:04:00.0 name: GeForce GTX 1060 6GB computeCapability: 6.1
coreClock: 1.7845GHz coreCount: 10 deviceMemorySize: 5.93GiB
deviceMemoryBandwidth: 178.99GiB/s
2022-01-18 18:11:00.559028: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcudart.so.10.1
2022-01-18 18:11:00.559071: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcublas.so.10
2022-01-18 18:11:00.559110: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
```

opened dynamic library libcufft.so.10

```
2022-01-18 18:11:00.559149: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcurand.so.10
2022-01-18 18:11:00.559191: I
tensorflow/stream executor/platform/default/dso loader.cc:48] Successfully
opened dynamic library libcusolver.so.10
2022-01-18 18:11:00.559230: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcusparse.so.10
2022-01-18 18:11:00.559270: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcudnn.so.7
2022-01-18 18:11:00.559379: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.559975: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:00.560485: I
tensorflow/core/common runtime/gpu/gpu device.cc:1858] Adding visible gpu
devices: 0
2022-01-18 18:11:00.560546: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
opened dynamic library libcudart.so.10.1
2022-01-18 18:11:01.236847: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1257] Device interconnect
StreamExecutor with strength 1 edge matrix:
2022-01-18 18:11:01.236888: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1263]
2022-01-18 18:11:01.236898: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1276] 0:
2022-01-18 18:11:01.237188: I
tensorflow/stream executor/cuda/cuda gpu executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:01.237710: I
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:982] successful NUMA node
read from SysFS had negative value (-1), but there must be at least one NUMA
node, so returning NUMA node zero
2022-01-18 18:11:01.238129: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1402] Created TensorFlow device
(/job:localhost/replica:0/task:0/device:GPU:0 with 4679 MB memory) -> physical
GPU (device: 0, name: GeForce GTX 1060 6GB, pci bus id: 0000:04:00.0, compute
capability: 6.1)
2022-01-18 18:11:01.918703: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully
```

opened dynamic library libcublas.so.10

Training time: 43.761592864990234s

WARNING:tensorflow:From /home/khiem/anaconda3/lib/python3.8/site-packages/tensorflow/python/training/tracking/tracking.py:111:

Model.state\_updates (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.

Instructions for updating:

This property should not be used in TensorFlow 2.0, as updates are applied automatically.

WARNING:tensorflow:From /home/khiem/anaconda3/lib/python3.8/site-packages/tensorflow/python/training/tracking/tracking.py:111: Layer.updates (from tensorflow.python.keras.engine.base\_layer) is deprecated and will be removed in a future version.

Instructions for updating:

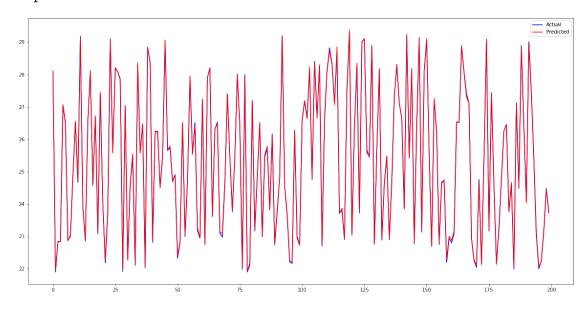
This property should not be used in TensorFlow 2.0, as updates are applied automatically.

2022-01-18 18:11:45.662754: W tensorflow/python/util/util.cc:348] Sets are not currently considered sequences, but this may change in the future, so consider avoiding using them.

INFO:tensorflow:Assets written to:

../reports/models/nn\_MRT\_open\_office\_model.pkl/assets

Mean squared error: 0.002490786140369075



## 3.3.2 With PCA

[10]: # Train the model
start = time.time()
model.fit(X\_train\_pca, y\_train\_pca, epochs=100, verbose=0)

```
stop = time.time()
nn_pca_train_time = stop - start
print(f"Training time: {nn_pca_train_time}s")

# Predict on test data
y_pred_nn_pca = model.predict(X_test_pca)
model.save('../reports/models/nn_MRT_pca_open_office_model.pkl')

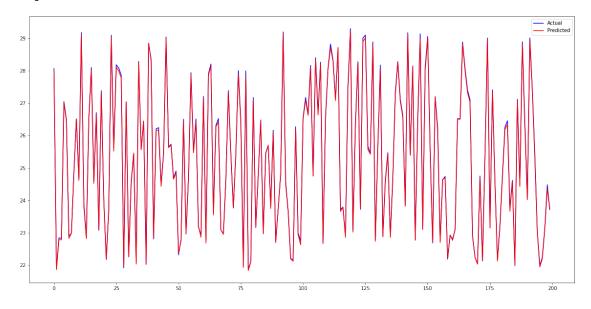
# Evaluate the model
mse = mean_squared_error(y_test_pca, y_pred_nn_pca)
print(f'Mean squared error: {mse}')

# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred_nn_pca[:200], color='red', label='Predicted')
plt.legend()
plt.show()
```

Training time: 44.0262987613678s

INFO:tensorflow:Assets written to:
../reports/models/nn\_MRT\_pca\_open\_office\_model.pkl/assets

Mean squared error: 0.0027803035776146155



```
[11]: # Save training times to a csv file
import csv

with open('../reports/data/training_times_open_office_MRT.csv', 'w') as csvfile:
```

```
writer = csv.writer(csvfile)
writer.writerow(['Model', 'Training time (s)'])
writer.writerow(['Linear regression', lr_train_time])
writer.writerow(['Linear regression with PCA', lr_pca_train_time])
writer.writerow(['Neural network', nn_train_time])
writer.writerow(['Neural network with PCA', nn_pca_train_time])
```

## 1.1.4 4. Compare models' results with and without PCA on predicting T

## 4.1 Split the data into training and testing

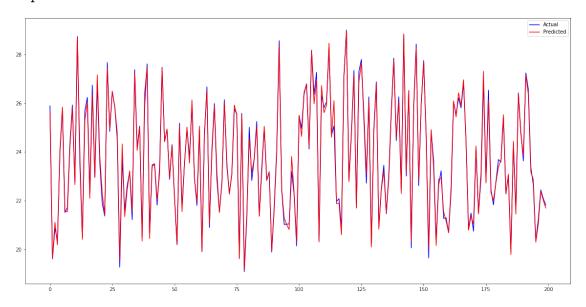
## 4.2 Train and evaluate the Linear Regression model

```
[13]: from sklearn.linear_model import LinearRegression
      from sklearn.metrics import mean_squared_error
      import pickle
      import time
      # Train linear regression model on training data
      model = LinearRegression()
      start = time.time()
      model.fit(X_train, y_train)
      stop = time.time()
      lr_train_time = stop - start
      print(f"Training time: {lr_train_time}s")
      # Save the model to a pickle file
      filename = '../reports/models/lr_T_open_office_model.pkl'
      pickle.dump(model, open(filename, 'wb'))
      # Predict on test data
      y_pred = model.predict(X_test)
      # Evaluate the model
      mse = mean_squared_error(y_test, y_pred)
```

```
print(f'Mean squared error: {mse}')

# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred[:200], color='red', label='Predicted')
plt.legend()
plt.show()
```

Training time: 0.006529092788696289s Mean squared error: 0.04119175262435056



# **4.2.2** With PCA

```
[14]: # Train linear regression model on training data
start = time.time()
model.fit(X_train_pca, y_train_pca)
stop = time.time()
lr_pca_train_time = stop - start
print(f"Training time: {lr_pca_train_time}s")

# Predict on test data
y_pred_pca = model.predict(X_test_pca)

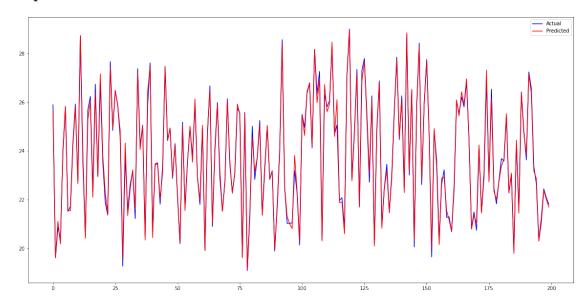
# Save the model to a pickle file
filename = '../reports/models/lr_T_pca_open_office_model.pkl'
pickle.dump(model, open(filename, 'wb'))

# Evaluate the model
```

```
mse = mean_squared_error(y_test_pca, y_pred_pca)
print(f'Mean squared error: {mse}')

# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred_pca[:200], color='red', label='Predicted')
plt.legend()
plt.show()
```

Training time: 0.005934476852416992s Mean squared error: 0.041191752624350535



## 4.3 Train and evaluate the Feedforward Neural Network model

```
[15]: # Create neural network model
from keras.models import Sequential
from keras.layers import Dense

N_NEURONS = 1024
N_LAYERS = 4

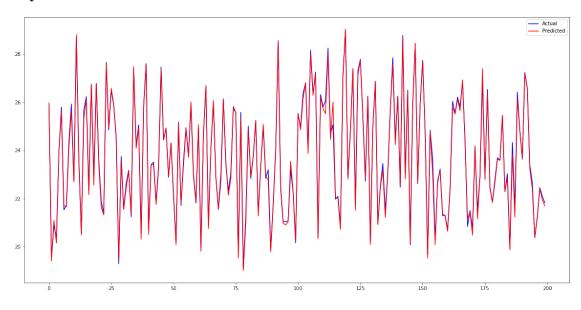
model = Sequential()
model.add(Dense(units=N_NEURONS, input_dim=X.shape[1], activation='relu'))
for i in range(N_LAYERS-1):
    model.add(Dense(units=N_NEURONS, activation='relu'))
model.add(Dense(units=1, activation='linear')) # Output layer
```

```
model.compile(loss='mean_squared_error', optimizer='adam')
# Train the model
start = time.time()
model.fit(X_train, y_train, epochs=100, verbose=0)
stop = time.time()
nn_train_time = stop - start
print(f"Training time: {nn_train_time}s")
# Predict on test data
y_pred_nn = model.predict(X_test)
model.save('../reports/models/nn_T_open_office_model.pkl')
# Evaluate the model
mse = mean_squared_error(y_test, y_pred_nn)
print(f'Mean squared error: {mse}')
# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred_nn[:200], color='red', label='Predicted')
plt.legend()
plt.show()
```

Training time: 45.3539400100708s INFO:tensorflow:Assets written to:

 $../{\tt reports/models/nn\_T\_open\_office\_model.pkl/assets}$ 

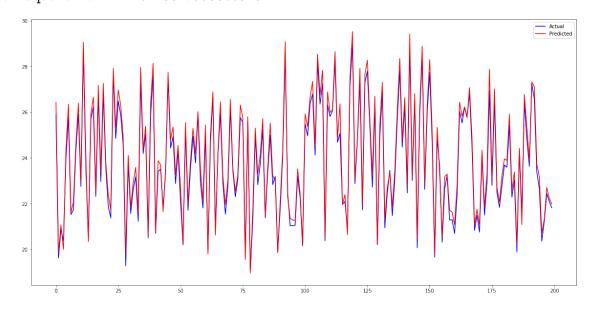
Mean squared error: 0.026644152618731248



#### 4.3.2 With PCA

```
[16]: # Train the model
      start = time.time()
      model.fit(X_train_pca, y_train_pca, epochs=100, verbose=0)
      stop = time.time()
      nn_pca_train_time = stop - start
      print(f"Training time: {nn_pca_train_time}s")
      # Predict on test data
      y_pred_nn_pca = model.predict(X_test_pca)
      model.save('../reports/models/nn_T_pca_open_office_model.pkl')
      # Evaluate the model
      mse = mean_squared_error(y_test_pca, y_pred_nn_pca)
      print(f'Mean squared error: {mse}')
      # Plot the predictions and actual values
      plt.figure(figsize=(20,10))
      plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
      plt.plot(range(200), y_pred_nn_pca[:200], color='red', label='Predicted')
      plt.legend()
     plt.show()
```

Training time: 46.28824257850647s
INFO:tensorflow:Assets written to:
../reports/models/nn\_T\_pca\_open\_office\_model.pkl/assets
Mean squared error: 0.13827998860915247



```
[17]: # Save training times to a csv file
import csv

with open('../reports/data/training_times_open_office_T.csv', 'w') as csvfile:
    writer = csv.writer(csvfile)
    writer.writerow(['Model', 'Training time (s)'])
    writer.writerow(['Linear regression', lr_train_time])
    writer.writerow(['Linear regression with PCA', lr_pca_train_time])
    writer.writerow(['Neural network', nn_train_time])
    writer.writerow(['Neural network with PCA', nn_pca_train_time])
```

# 1.1.5 5. Compare models' results with and without PCA on predicting V

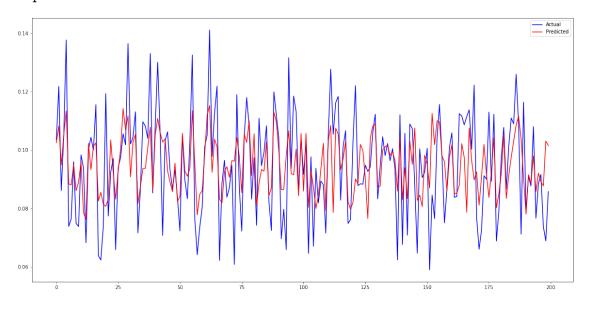
## 5.1 Split the data into training and testing

# 5.2 Train and evaluate the Linear Regression model

```
5.2.1 Without PCA
[19]: X_train.isna().sum()
[19]: tem
              0
     mass
              0
              0
     n
      осс
              0
              0
      X
      dtype: int64
[20]: y_V.isna().sum()
[20]: 0
[21]: from sklearn.linear_model import LinearRegression
      from sklearn.metrics import mean_squared_error
      import pickle
      import time
```

```
# Train linear regression model on training data
model = LinearRegression()
start = time.time()
model.fit(X_train, y_train)
stop = time.time()
lr_train_time = stop - start
print(f"Training time: {lr_train_time}s")
# Save the model to a pickle file
filename = '../reports/models/lr_V_open_office_model.pkl'
pickle.dump(model, open(filename, 'wb'))
# Predict on test data
y_pred = model.predict(X_test)
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
print(f'Mean squared error: {mse}')
# Plot the predictions and actual values
plt.figure(figsize=(20,10))
plt.plot(range(200), y_test[:200], color='blue', label='Actual')
plt.plot(range(200), y_pred[:200], color='red', label='Predicted')
plt.legend()
plt.show()
```

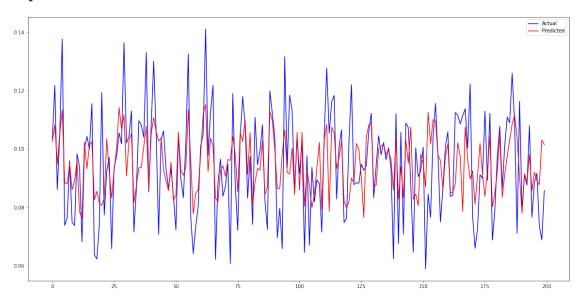
Training time: 0.007966041564941406s
Mean squared error: 0.0002148180985995211



#### **5.2.2** With PCA

```
[22]: # Train linear regression model on training data
      start = time.time()
      model.fit(X_train_pca, y_train_pca)
      stop = time.time()
      lr_pca_train_time = stop - start
      print(f"Training time: {lr_pca_train_time}s")
      # Predict on test data
      y_pred_pca = model.predict(X_test_pca)
      # Save the model to a pickle file
      filename = '../reports/models/lr_V_pca_open_office_model.pkl'
      pickle.dump(model, open(filename, 'wb'))
      # Evaluate the model
      mse = mean_squared_error(y_test_pca, y_pred_pca)
      print(f'Mean squared error: {mse}')
      # Plot the predictions and actual values
      plt.figure(figsize=(20,10))
      plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
      plt.plot(range(200), y_pred_pca[:200], color='red', label='Predicted')
     plt.legend()
      plt.show()
```

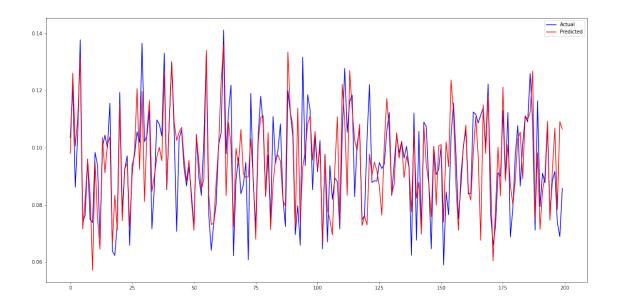
Training time: 0.006435394287109375s
Mean squared error: 0.0002148180985995211



## 5.3 Train and evaluate the Feedforward Neural Network model

```
[23]: # Create neural network model
      from keras.models import Sequential
      from keras.layers import Dense
      N_NEURONS = 1024
      N_LAYERS = 4
      model = Sequential()
      model.add(Dense(units=N NEURONS, input_dim=X.shape[1], activation='relu'))
      for i in range(N_LAYERS-1):
          model.add(Dense(units=N_NEURONS, activation='relu'))
      model.add(Dense(units=1, activation='linear')) # Output layer
      model.compile(loss='mean_squared_error', optimizer='adam')
      # Train the model
      start = time.time()
      model.fit(X_train, y_train, epochs=100, verbose=0)
      stop = time.time()
      nn_train_time = stop - start
      print(f"Training time: {nn_train_time}s")
      # Predict on test data
      y_pred_nn = model.predict(X test)
      model.save('../reports/models/nn_V_open_office_model.pkl')
      # Evaluate the model
      mse = mean_squared_error(y_test, y_pred_nn)
      print(f'Mean squared error: {mse}')
      # Plot the predictions and actual values
      plt.figure(figsize=(20,10))
      plt.plot(range(200), y test[:200], color='blue', label='Actual')
      plt.plot(range(200), y_pred_nn[:200], color='red', label='Predicted')
      plt.legend()
     plt.show()
```

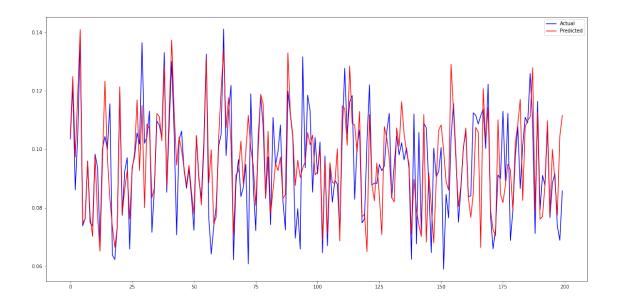
```
Training time: 47.35318875312805s
INFO:tensorflow:Assets written to:
../reports/models/nn_V_open_office_model.pkl/assets
Mean squared error: 0.0001326037106000784
```



# 5.3.2 With PCA

```
[24]: # Train the model
      start = time.time()
      model.fit(X_train_pca, y_train_pca, epochs=100, verbose=0)
      stop = time.time()
      nn_pca_train_time = stop - start
      print(f"Training time: {nn_pca_train_time}s")
      # Predict on test data
      y_pred_nn_pca = model.predict(X_test_pca)
      model.save('../reports/models/nn_V_pca_open_office_model.pkl')
      # Evaluate the model
      mse = mean_squared_error(y_test_pca, y_pred_nn_pca)
      print(f'Mean squared error: {mse}')
      # Plot the predictions and actual values
      plt.figure(figsize=(20,10))
      plt.plot(range(200), y_test_pca[:200], color='blue', label='Actual')
      plt.plot(range(200), y_pred_nn_pca[:200], color='red', label='Predicted')
     plt.legend()
     plt.show()
```

Training time: 44.344775438308716s
INFO:tensorflow:Assets written to:
../reports/models/nn\_V\_pca\_open\_office\_model.pkl/assets
Mean squared error: 0.00018198284814540812



```
[25]: # Save training times to a csv file
import csv

with open('../reports/data/training_times_open_office_V.csv', 'w') as csvfile:
    writer = csv.writer(csvfile)
    writer.writerow(['Model', 'Training time (s)'])
    writer.writerow(['Linear regression', lr_train_time])
    writer.writerow(['Linear regression with PCA', lr_pca_train_time])
    writer.writerow(['Neural network', nn_train_time])
    writer.writerow(['Neural network with PCA', nn_pca_train_time])
```