#### Import the libraries

```
In [1]: #importing the libraries
import tensorflow as tf
from tensorflow import keras
import numpy as np
import pandas as pd
from sklearn.metrics import mean_absolute_error
import matplotlib.pyplot as plt
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\losses.py:2976: The name tf.losses.sparse\_softmax\_cross\_entropy is deprecated. Please use tf.compat.v1.losses .sparse\_softmax\_cross\_entropy instead.

### Load the data and data pre-processing

```
In [2]: #load the extracted features
    baseline_df = pd.read_excel('extracted_features_baseline.xlsx')
    toolwear_df = pd.read_excel('extracted_features_toolwear.xlsx')

In [3]: good_features = baseline_df.values
    bad_features = toolwear_df.values
```

## train test split

from sklearn.model\_selection import train\_test\_split good\_train, good\_test = train\_test\_split(good\_features, test\_size=0.2, random\_state=5)

bad train, bad test = train test split(bad features, test size=0.2, random state=5)

```
In [5]: #data scalling
    from sklearn.preprocessing import StandardScaler

sc = StandardScaler()
    good_train = sc.fit_transform(good_train)
    good_test = sc.transform(good_test)
    bad_train = sc.fit_transform(bad_train)
    bad_test = sc.transform(bad_test)
```

# Specifying the number of encoded features to be investigated

```
In [6]: feature_size = [5, 10, 20, 40, 60]
```

# Train and Evaluate Autoencoders with different hidden layer size (encoded features)

```
In [7]: result = []
        # The number of encoded feature of autoencoder is varied in each iteration of the for loop
        for size in feature_size:
            condensed_f = size
            #constructing the good autoencoder model
            #input layer which number of neurons equals the number of original features
            l in good = keras.Input(good features.shape[1])
            #hidden layer which condenses the feature into the specified number of condensed features
            l_condensed_good = keras.layers.Dense(condensed_f)(l_in_good)
            #output layer which is the same as the input
            l_out_good = keras.layers.Dense(good_features.shape[1])(l_condensed_good)
            #defining the good autoencoder
            autoencoder_good = keras.Model(l_in_good, l_out_good)
            #compile the model
            autoencoder_good.compile(optimizer='adam', loss='mse')
            #train the model
```

```
autoencoder good.fit(good train, good train, epochs = 50, batch size = 8, validation split = 0.1)
#The autoencoders are asked to predict both classes of test data
GAE_pred_good = autoencoder_good.predict(good_test)
GAE pred bad = autoencoder good.predict(bad test)
#The MSE of each entry in the entire dataset is stored in a temporary array
#The 2 temporary arrays will be used to determine the margin between 2 classes (it needs all the values to
GAE MSE bad = []
GAE_MSE_good = []
#The MSE of each entry in the test datasets are computed and stored in the temporary array
for i in range(len(bad test)):
    GAE MSE good.append(mean absolute error(good test[i],GAE pred good[i]))
    GAE MSE bad.append(mean absolute error(bad test[i],GAE pred bad[i]))
#For each variation of window size, the margin between 2 classes will be stored in 'difference' array
difference = []
for x in range(1,20):
    window = x
    GAE_average_good = []
   GAE average bad = []
    for ind in range(len(GAE MSE good) - window + 1):
        GAE average good.append(np.mean(GAE MSE good[ind:ind+window]))
        GAE average bad.append(np.mean(GAE MSE bad[ind:ind+window]))
    #calculates the minimum difference in MSE between 2 classes and stores in the array
   difference.append(min(GAE average bad) - max(GAE average good))
#The array 'difference' is obtained for each autoencoder with different number of encoded features. The ent.
result.append(difference)
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\backend.py:1398: The name tf.executing\_eagerly\_outside\_functions is deprecated. Please use tf.compat.v1.executing eagerly outside functions instead.

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\optimizers\\_\_init\_\_.py:309: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimize r instead.

#### Epoch 1/50

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\utils\tf\_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

```
=========] - 1s 7ms/step - loss: 1.1079 - val loss: 1.0250
38/38 [===
Epoch 2/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.9638 - val loss: 0.9201
Epoch 3/50
38/38 [============] - 0s 3ms/step - loss: 0.8571 - val loss: 0.8201
Epoch 4/50
38/38 [=====
       Epoch 5/50
Epoch 6/50
38/38 [====
           =========] - Os 2ms/step - loss: 0.6129 - val loss: 0.6019
Epoch 7/50
38/38 [====
         Epoch 8/50
38/38 [=====
          Epoch 9/50
38/38 [====
           ========] - Os 2ms/step - loss: 0.5034 - val_loss: 0.5106
Epoch 10/50
38/38 [======
        Fnoch 11/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.4601 - val loss: 0.4722
Epoch 12/50
38/38 [========== ] - 0s 2ms/step - loss: 0.4440 - val loss: 0.4556
Epoch 13/50
Epoch 14/50
38/38 [====
           Epoch 15/50
38/38 [====
             ========] - Os 2ms/step - loss: 0.4082 - val loss: 0.4216
Epoch 16/50
38/38 [=====
           ========] - 0s 3ms/step - loss: 0.3984 - val loss: 0.4130
Epoch 17/50
Epoch 18/50
```

```
Epoch 19/50
38/38 [============= ] - 0s 2ms/step - loss: 0.3751 - val loss: 0.3961
Epoch 20/50
38/38 [============ ] - 0s 2ms/step - loss: 0.3689 - val loss: 0.3930
Epoch 21/50
38/38 [=====
            ========] - Os 2ms/step - loss: 0.3636 - val loss: 0.3909
Epoch 22/50
38/38 [======
         Epoch 23/50
38/38 [=====
          Epoch 24/50
38/38 [====
              ========] - Os 2ms/step - loss: 0.3525 - val_loss: 0.3840
Epoch 25/50
38/38 [============= ] - 0s 2ms/step - loss: 0.3501 - val loss: 0.3842
Epoch 26/50
Epoch 27/50
Epoch 28/50
          38/38 [=====
Epoch 29/50
38/38 [====
              ========] - 0s 2ms/step - loss: 0.3430 - val_loss: 0.3814
Epoch 30/50
             ========] - Os 2ms/step - loss: 0.3424 - val_loss: 0.3806
38/38 [====
Epoch 31/50
38/38 [============ ] - 0s 2ms/step - loss: 0.3413 - val loss: 0.3797
Epoch 32/50
38/38 [==============] - 0s 2ms/step - loss: 0.3408 - val_loss: 0.3782
Epoch 33/50
Epoch 34/50
38/38 [============= ] - 0s 2ms/step - loss: 0.3389 - val loss: 0.3785
Epoch 35/50
Epoch 36/50
38/38 [============ ] - 0s 2ms/step - loss: 0.3376 - val loss: 0.3792
Epoch 37/50
Epoch 38/50
38/38 [============== ] - 0s 3ms/step - loss: 0.3367 - val_loss: 0.3784
Epoch 39/50
38/38 [====
                    ==] - 0s 3ms/step - loss: 0.3364 - val_loss: 0.3778
Epoch 40/50
Epoch 41/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.3359 - val loss: 0.3784
Epoch 42/50
38/38 [============] - 0s 2ms/step - loss: 0.3358 - val loss: 0.3769
Epoch 43/50
38/38 [=====
          =========] - Os 2ms/step - loss: 0.3356 - val_loss: 0.3760
Epoch 44/50
              =======] - 0s 2ms/step - loss: 0.3353 - val_loss: 0.3777
38/38 [====
Epoch 45/50
38/38 [====
                =======] - 0s 2ms/step - loss: 0.3354 - val loss: 0.3765
Epoch 46/50
38/38 [=====
          Epoch 47/50
38/38 [=============] - 0s 2ms/step - loss: 0.3350 - val loss: 0.3768
Epoch 48/50
Epoch 49/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.3347 - val loss: 0.3769
Epoch 50/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.3343 - val loss: 0.3758
3/3 [======] - 0s 2ms/step
3/3 [=======] - 0s 2ms/step
Epoch 1/50
38/38 [============] - 1s 6ms/step - loss: 1.2304 - val loss: 1.1386
Epoch 2/50
38/38 [====
            ========] - Os 2ms/step - loss: 0.9853 - val loss: 0.9347
Epoch 3/50
38/38 [===
           Epoch 4/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.6823 - val loss: 0.6496
Epoch 5/50
Epoch 6/50
Epoch 7/50
Epoch 8/50
38/38 [===
           Epoch 9/50
```

```
Epoch 10/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.3723 - val loss: 0.3951
Epoch 11/50
38/38 [==
           =======] - Os 2ms/step - loss: 0.3494 - val loss: 0.3758
Epoch 12/50
38/38 [=====
           =======] - Os 2ms/step - loss: 0.3288 - val loss: 0.3577
Epoch 13/50
38/38 [=============] - 0s 2ms/step - loss: 0.3107 - val_loss: 0.3437
Epoch 14/50
38/38 [======
       Epoch 15/50
Epoch 16/50
38/38 [========== ] - 0s 2ms/step - loss: 0.2729 - val loss: 0.3095
Epoch 17/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.2638 - val loss: 0.3005
Epoch 18/50
Epoch 19/50
38/38 [============ ] - 0s 2ms/step - loss: 0.2483 - val loss: 0.2891
Epoch 20/50
38/38 [=====
         ===============] - 0s 2ms/step - loss: 0.2410 - val loss: 0.2825
Epoch 21/50
        38/38 [====
Epoch 22/50
38/38 [============] - 0s 2ms/step - loss: 0.2294 - val loss: 0.2723
Epoch 23/50
Epoch 24/50
38/38 [============ ] - 0s 2ms/step - loss: 0.2185 - val loss: 0.2623
Epoch 25/50
Epoch 26/50
38/38 [=====
        Epoch 27/50
38/38 [============] - 0s 2ms/step - loss: 0.2043 - val loss: 0.2522
Epoch 28/50
38/38 [============] - 0s 2ms/step - loss: 0.2003 - val_loss: 0.2485
Epoch 29/50
38/38 [=====
        Epoch 30/50
Fnoch 31/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.1886 - val loss: 0.2393
Epoch 32/50
Epoch 33/50
Epoch 34/50
38/38 [=====
        Epoch 35/50
38/38 [=====
          =========] - 0s 2ms/step - loss: 0.1771 - val loss: 0.2300
Epoch 36/50
38/38 [============ ] - 0s 2ms/step - loss: 0.1745 - val loss: 0.2277
Epoch 37/50
Epoch 38/50
Epoch 39/50
38/38 [============= ] - 0s 2ms/step - loss: 0.1688 - val loss: 0.2226
Epoch 40/50
38/38 [============ ] - 0s 3ms/step - loss: 0.1672 - val loss: 0.2203
Epoch 41/50
Epoch 42/50
Epoch 43/50
38/38 [=====
       Epoch 44/50
38/38 [====
        =========] - 0s 3ms/step - loss: 0.1612 - val_loss: 0.2145
Epoch 45/50
Epoch 46/50
Epoch 47/50
Epoch 48/50
38/38 [===========] - 0s 3ms/step - loss: 0.1577 - val_loss: 0.2102
Epoch 49/50
      38/38 [=====
Epoch 50/50
38/38 [=====
```

```
3/3 [=======] - 0s 2ms/step
Epoch 1/50
Epoch 2/50
38/38 [====
        Epoch 3/50
38/38 [=====
      Epoch 4/50
38/38 [====
      Epoch 5/50
38/38 [====
          Epoch 6/50
Epoch 7/50
Epoch 8/50
Epoch 9/50
       38/38 [=====
Epoch 10/50
          ========] - Os 2ms/step - loss: 0.2392 - val_loss: 0.2726
38/38 [====
Epoch 11/50
38/38 [=====
          Epoch 12/50
38/38 [============ ] - 0s 2ms/step - loss: 0.1954 - val loss: 0.2349
Epoch 13/50
38/38 [==============] - 0s 3ms/step - loss: 0.1793 - val_loss: 0.2213
Epoch 14/50
Epoch 15/50
Epoch 16/50
Epoch 17/50
38/38 [=============] - 0s 3ms/step - loss: 0.1361 - val loss: 0.1839
Epoch 18/50
38/38 [============] - 0s 2ms/step - loss: 0.1289 - val_loss: 0.1753
Epoch 19/50
Epoch 20/50
38/38 [====
                ===] - 0s 2ms/step - loss: 0.1157 - val_loss: 0.1640
Epoch 21/50
Epoch 22/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.1052 - val loss: 0.1562
Epoch 23/50
38/38 [============] - 0s 2ms/step - loss: 0.1006 - val loss: 0.1513
Epoch 24/50
38/38 [=====
        Epoch 25/50
           =======] - Os 3ms/step - loss: 0.0924 - val loss: 0.1426
38/38 [====
Epoch 26/50
38/38 [====
            =======] - Os 3ms/step - loss: 0.0889 - val loss: 0.1394
Epoch 27/50
38/38 [=====
        Epoch 28/50
38/38 [=============] - 0s 2ms/step - loss: 0.0831 - val loss: 0.1337
Epoch 29/50
Epoch 30/50
38/38 [=========== ] - 0s 2ms/step - loss: 0.0779 - val loss: 0.1278
Epoch 31/50
38/38 [=========== ] - 0s 5ms/step - loss: 0.0755 - val loss: 0.1254
Epoch 32/50
38/38 [============] - 0s 2ms/step - loss: 0.0735 - val_loss: 0.1233
Epoch 33/50
38/38 [============] - 0s 6ms/step - loss: 0.0714 - val_loss: 0.1196
Epoch 34/50
38/38 [=====
          ========] - 0s 3ms/step - loss: 0.0698 - val loss: 0.1184
Epoch 35/50
38/38 [====
         ==========] - 0s 6ms/step - loss: 0.0680 - val loss: 0.1171
Epoch 36/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.0661 - val loss: 0.1138
Epoch 37/50
Epoch 38/50
Epoch 39/50
Epoch 40/50
        38/38 [=====
Epoch 41/50
```

```
Epoch 42/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.0575 - val loss: 0.1015
Epoch 43/50
38/38 [==
         Epoch 44/50
38/38 [=====
         ========] - Os 3ms/step - loss: 0.0550 - val loss: 0.0978
Epoch 45/50
38/38 [============] - 0s 2ms/step - loss: 0.0542 - val_loss: 0.0952
Epoch 46/50
38/38 [============ ] - 0s 2ms/step - loss: 0.0530 - val loss: 0.0932
Epoch 47/50
Epoch 48/50
38/38 [========== ] - 0s 2ms/step - loss: 0.0509 - val loss: 0.0898
Epoch 49/50
38/38 [========== ] - 0s 3ms/step - loss: 0.0499 - val loss: 0.0885
Epoch 50/50
38/38 [============] - 0s 3ms/step - loss: 0.0490 - val loss: 0.0869
3/3 [======] - 0s 2ms/step
3/3 [======
      Epoch 1/50
      38/38 [=====
Epoch 2/50
38/38 [====
      Epoch 3/50
38/38 [============] - 0s 2ms/step - loss: 0.4643 - val loss: 0.4341
Epoch 4/50
Epoch 5/50
38/38 [============] - 0s 3ms/step - loss: 0.2783 - val_loss: 0.2898
Epoch 6/50
Epoch 7/50
38/38 [=====
     Epoch 8/50
38/38 [============] - 0s 3ms/step - loss: 0.1634 - val loss: 0.1897
Epoch 9/50
38/38 [============] - 0s 3ms/step - loss: 0.1418 - val_loss: 0.1706
Epoch 10/50
38/38 [=====
      Epoch 11/50
Fnoch 12/50
Epoch 13/50
Epoch 14/50
Epoch 15/50
38/38 [=====
      Epoch 16/50
38/38 [=====
       Epoch 17/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.0659 - val loss: 0.0940
Epoch 18/50
Epoch 19/50
Epoch 20/50
38/38 [============= ] - 0s 2ms/step - loss: 0.0534 - val loss: 0.0818
Epoch 21/50
38/38 [============ ] - 0s 2ms/step - loss: 0.0500 - val loss: 0.0776
Epoch 22/50
Epoch 23/50
Epoch 24/50
38/38 [======
      Epoch 25/50
       38/38 [====
Epoch 26/50
Epoch 27/50
Epoch 28/50
Epoch 29/50
Epoch 30/50
     38/38 [=====
Epoch 31/50
38/38 [============ ] - 0s 2ms/step - loss: 0.0301 - val loss: 0.0513
```

```
Epoch 32/50
38/38 [============ ] - 0s 2ms/step - loss: 0.0289 - val loss: 0.0488
Epoch 33/50
38/38 [============= ] - 0s 2ms/step - loss: 0.0278 - val loss: 0.0472
Epoch 34/50
38/38 [=====
            =========] - 0s 2ms/step - loss: 0.0264 - val loss: 0.0444
Epoch 35/50
38/38 [=====
         Epoch 36/50
38/38 [=====
          Epoch 37/50
38/38 [====
              ========] - Os 2ms/step - loss: 0.0232 - val_loss: 0.0395
Epoch 38/50
38/38 [============= ] - 0s 2ms/step - loss: 0.0224 - val loss: 0.0378
Epoch 39/50
Epoch 40/50
Epoch 41/50
          38/38 [=====
Epoch 42/50
             ========] - Os 2ms/step - loss: 0.0192 - val_loss: 0.0324
38/38 [=====
Epoch 43/50
38/38 [=====
             ========] - 0s 2ms/step - loss: 0.0186 - val_loss: 0.0311
Epoch 44/50
38/38 [============ ] - 0s 2ms/step - loss: 0.0180 - val loss: 0.0304
Epoch 45/50
Epoch 46/50
Epoch 47/50
38/38 [============ ] - 0s 6ms/step - loss: 0.0164 - val loss: 0.0277
Epoch 48/50
38/38 [========== ] - 0s 6ms/step - loss: 0.0159 - val loss: 0.0266
Epoch 49/50
38/38 [============ ] - 0s 4ms/step - loss: 0.0156 - val loss: 0.0259
Epoch 50/50
3/3 [=======] - 0s 2ms/step
        3/3 [======
Epoch 1/50
38/38 [====
                 ======] - 1s 5ms/step - loss: 1.2152 - val_loss: 0.8030
Epoch 2/50
Epoch 3/50
38/38 [=========== ] - 0s 3ms/step - loss: 0.3475 - val loss: 0.3527
Epoch 4/50
38/38 [============] - 0s 4ms/step - loss: 0.2430 - val loss: 0.2673
Epoch 5/50
38/38 [====
           =========] - 0s 4ms/step - loss: 0.1822 - val_loss: 0.2160
Epoch 6/50
              =======] - 0s 4ms/step - loss: 0.1436 - val loss: 0.1786
38/38 [====
Epoch 7/50
38/38 [===
                =======] - Os 5ms/step - loss: 0.1171 - val loss: 0.1524
Epoch 8/50
38/38 [====
          Epoch 9/50
38/38 [============] - 0s 3ms/step - loss: 0.0846 - val loss: 0.1192
Epoch 10/50
Epoch 11/50
38/38 [========== ] - 0s 3ms/step - loss: 0.0662 - val loss: 0.0969
Epoch 12/50
38/38 [========== ] - 0s 4ms/step - loss: 0.0599 - val loss: 0.0893
Epoch 13/50
38/38 [=============] - 0s 4ms/step - loss: 0.0540 - val_loss: 0.0816
Epoch 14/50
38/38 [======
         Epoch 15/50
38/38 [=====
             ========] - Os 3ms/step - loss: 0.0453 - val loss: 0.0692
Epoch 16/50
            =========] - Os 4ms/step - loss: 0.0417 - val loss: 0.0647
38/38 [====
Epoch 17/50
38/38 [========== ] - 0s 3ms/step - loss: 0.0386 - val loss: 0.0608
Epoch 18/50
Epoch 19/50
Epoch 20/50
Epoch 21/50
38/38 [====
          Epoch 22/50
```

```
Epoch 23/50
    38/38 [============ ] - 0s 2ms/step - loss: 0.0254 - val loss: 0.0418
    Epoch 24/50
    38/38 [==
              Epoch 25/50
    38/38 [=====
              Epoch 26/50
    38/38 [============] - 0s 2ms/step - loss: 0.0216 - val_loss: 0.0351
    Epoch 27/50
    38/38 [============ ] - 0s 2ms/step - loss: 0.0207 - val loss: 0.0334
    Epoch 28/50
    Epoch 29/50
    38/38 [=========== ] - 0s 2ms/step - loss: 0.0186 - val loss: 0.0307
    Epoch 30/50
    38/38 [============] - 0s 2ms/step - loss: 0.0175 - val loss: 0.0289
    Epoch 31/50
    38/38 [============] - 0s 3ms/step - loss: 0.0167 - val loss: 0.0278
    Epoch 32/50
    38/38 [=========== ] - 0s 3ms/step - loss: 0.0160 - val loss: 0.0269
    Epoch 33/50
             38/38 [======
    Epoch 34/50
           38/38 [=====
    Epoch 35/50
    38/38 [============] - 0s 3ms/step - loss: 0.0140 - val loss: 0.0231
    Epoch 36/50
    Epoch 37/50
    Epoch 38/50
    Epoch 39/50
    38/38 [============] - 0s 3ms/step - loss: 0.0117 - val loss: 0.0195
    Epoch 40/50
    38/38 [============] - 0s 3ms/step - loss: 0.0112 - val loss: 0.0183
    Epoch 41/50
    38/38 [=============] - 0s 2ms/step - loss: 0.0110 - val_loss: 0.0181
    Epoch 42/50
    38/38 [=====
           Epoch 43/50
    Epoch 44/50
    38/38 [=========== ] - 0s 2ms/step - loss: 0.0096 - val loss: 0.0159
    Epoch 45/50
    Epoch 46/50
    Epoch 47/50
    38/38 [=====
           Epoch 48/50
    38/38 [=====
             Epoch 49/50
    38/38 [============ ] - 0s 3ms/step - loss: 0.0080 - val loss: 0.0137
    Epoch 50/50
    38/38 [============] - 0s 3ms/step - loss: 0.0076 - val_loss: 0.0125
    3/3 [======] - 0s 2ms/step
    3/3 [=======] - 0s 1ms/step
In [8]: #plotting the result
    window sizes = list(range(1,len(result[0])+1))
    plt.plot(window_sizes, result[0], label = "5 features")
    plt.plot(window_sizes, result[1], label = "10 features")
    plt.plot(window_sizes, result[2], label = "20 features")
    plt.plot(window_sizes, result[3], label = "40 features")
plt.plot(window_sizes, result[4], label = "60 features")
    plt.xlabel("Window Size")
    plt.ylabel("Margin")
    plt.legend(loc='lower right')
    plt.xticks([2, 4, 6, 8, 10, 12, 14, 16, 18, 20])
```

```
Out[8]: ([<matplotlib.axis.XTick at 0x22d091f3b90>,
           <matplotlib.axis.XTick at 0x22d0e3e7110>,
           <matplotlib.axis.XTick at 0x22d0e615010>,
           <matplotlib.axis.XTick at 0x22d0f76a210>,
           <matplotlib.axis.XTick at 0x22d0f770690>,
           <matplotlib.axis.XTick at 0x22d0cfa38d0>,
           <matplotlib.axis.XTick at 0x22d0f773dd0>,
           <matplotlib.axis.XTick at 0x22d0f77e190>,
           <matplotlib.axis.XTick at 0x22d0f7845d0>,
           <matplotlib.axis.XTick at 0x22d0f786910>],
          [Text(2, 0, '2'),
Text(4, 0, '4'),
           Text(6, 0, '6'),
Text(8, 0, '8'),
           Text(10, 0, '10'),
           Text(12, 0, '12'),
           Text(14, 0, '14'),
Text(16, 0, '16'),
           Text(18, 0, '18'),
           Text(20, 0, '20')])
            0.2
            0.1
            0.0
        Margin
           -0.1
           -0.2
                                                                           5 features
                                                                           10 features
           -0.3
                                                                           20 features
                                                                           40 features
                                                                           60 features
           -0.4
                       2
                                            8
                                                   10
                                                          12
                                                                  14
                                                                                       20
                              4
                                      6
                                                                         16
                                                                                18
                                              Window Size
```

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