

BMEG 3103 (AI)

PROJECT 2

AU, Wai Tak (1155175068)
CHAN, Cheuk Ka (1155174356)

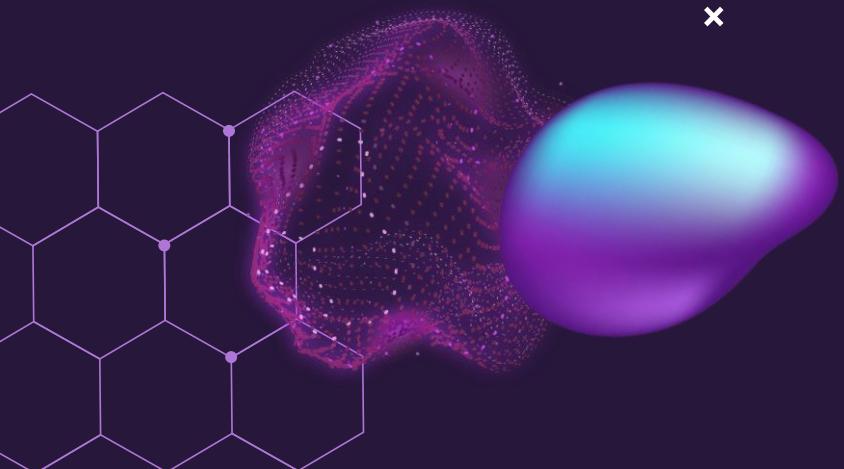
OUR TEAM



AU Wai Tak

7.

x



CHAN Cheuk Ka

9.



x



INTRODUCTION

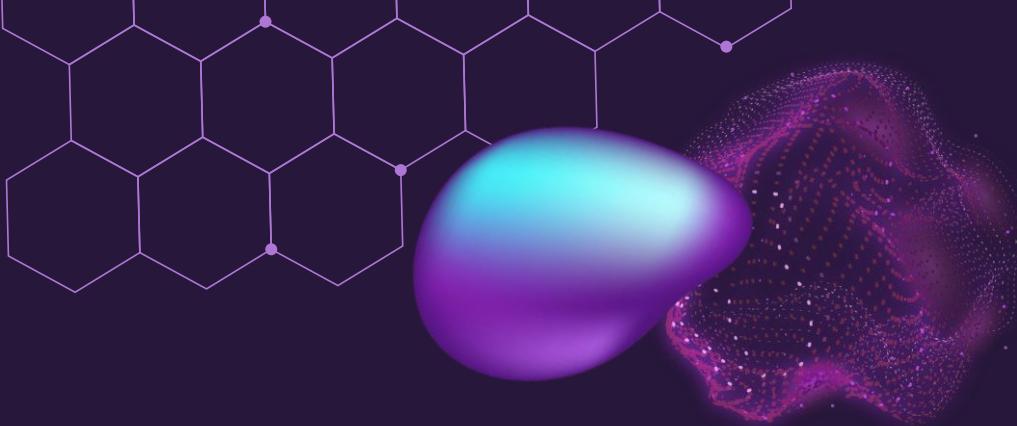
This project's task is to create a machine learning algorithm that can classify the health of patients into 5 categories.



01

DATA PRE-PROCESSING

+ data sanitisation



x



+



IRRELEVANT COLUMNS

We dropped the first few useless columns pertaining username, timestamp, etc.

A	B	C	D	E	new
1	user_name	raw_timestamp_part_1	raw_timestamp_part_2	cvted_timestamp	no
2	carlitos	1323084231	788290	5/12/201111:23	no
3	2 carlitos	1323084231	808298	5/12/201111:23	no
4	3 carlitos	1323084231	820366	5/12/201111:23	no
5	4 carlitos	1323084232	120339	5/12/201111:23	no
6	5 carlitos	1323084232	196328	5/12/201111:23	no
7	6 carlitos	1323084232	304277	5/12/201111:23	no
8	7 carlitos	1323084232	368296	5/12/201111:23	no
9	8 carlitos	1323084232	440390	5/12/201111:23	no
10	9 carlitos	1323084232	484323	5/12/201111:23	no
11	10 carlitos	1323084232	484434	5/12/201111:23	no
12	11 carlitos	1323084232	500302	5/12/201111:23	no
13	12 carlitos	1323084232	528316	5/12/201111:23	no
14	13 carlitos	1323084232	560359	5/12/201111:23	no
15	14 carlitos	1323084232	576390	5/12/201111:23	no
16	15 carlitos	1323084232	604281	5/12/201111:23	no
17	16 carlitos	1323084232	644202	5/12/201111:23	no



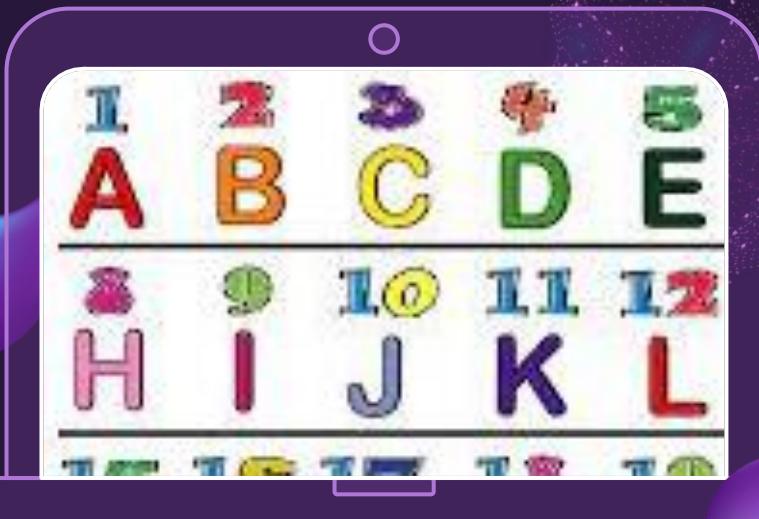
ERRONEOUS DATA

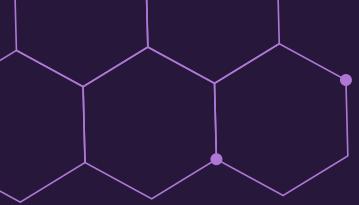
We trashed the columns containing empty cells and errors

			NA	NA	
			NA	NA	
			NA	NA	
			NA	NA	
			NA	NA	
			NA	NA	
			NA	NA	
			NA	NA	
2.713152	#DIV/0!	#DIV/0!	-94.3	3	5.6
			NA	NA	
			NA	NA	
			NA	NA	

DIGITISE CLASS

We changed class letters to numbers





SANITISED

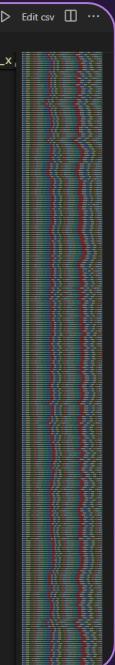
After sanitisation, we acquired a 19622×55 CSV matrix (excluding labels).



project2_trainingset_sanitised.csv

dataset > project2_trainingset_sanitised.csv > data

```
1 ,_belt_x,magnet_belt_y,magnet_belt_z,roll_arm,pitch_arm,yaw_arm,total_accel_arm,gyros_arm_x,gyros_arm_y,gyros_arm_z,accel_arm_x,
19589 1771,1,-0.27,0.58,0.26,1,-3,-7,-571,242,-10,-147,7,17,-47,1,49,-0.98,-0.75,-0.92,241,-292,411,-618,-251,5
19590 137,5,-0.37,0.88,0.33,-1,46,25,-564,255,5,-148,8,74,-46,2,49,-0.42,0.51,-0.15,245,-310,-281,413,-639,-286,5
19591 112,7,-0.45,1.25,0.61,7,61,35,-552,284,9,-147,10,1,-48,5,44,0.32,1.27,0.64,211,-270,-264,389,-653,-266,5
19592 1,-0.5,1.43,0.64,6,68,46,-540,309,-1,-146,10,6,-50,1,38,0.5,1.28,0.89,139,-233,-255,379,-654,-259,5
19593 -0.72,2,18,0.46,-8,100,50,-496,371,7,-144,12,8,-55,5,34,0.4,1.8,0.61,72,-208,-256,306,-666,-177,5
19594 1,14,-0.87,2.5,0.13,-14,120,54,-427,440,38,-144,15,5,-59,9,34,0.16,2,12,0.51,57,-210,-246,243,-665,-137,5
19595 -0.93,2,63,0.25,-20,129,48,-407,453,38,-144,17,3,-62,4,33,0.08,2,33,0.49,50,-210,-245,224,-672,-132,5
19596 1,14,-0.95,2,71,0,44,-12,125,43,-359,497,44,-145,19,-64,9,33,0.08,2,62,0.51,37,-217,-237,185,-673,-117,5
19597 32,-7,116,42,-342,502,49,-145,21,1,-68,33,0.08,2,83,0.64,33,-224,-234,157,-672,-108,5
19598 972912,25,-0.5,1,04,0.23,53,231,-66,-78,583,134,0,0,0,39,-0.61,4,56,1,12,-237,298,33,-703,382,803,5
19599 1838,24,-0.32,0.79,0.13,46,222,-62,-44,591,134,0,0,0,34,-0.67,4,61,1,28,-107,389,70,-693,517,890,5
19600 125287,28,0.43,0.37,-0.28,84,253,-74,10,596,123,0,0,0,52,0.19,4,5,2,24,505,7,-463,839,905,5
19601 1469457,28,0.43,0.39,-0.38,82,252,-70,10,601,117,0,0,0,51,0.22,4,06,1,84,38,496,4,-408,852,878,5
19602 18,27,0.58,0.02,-0.34,88,236,-75,15,590,135,0,0,0,52,0.51,3.1,1.36,64,501,0,-204,821,777,5
19603 1136895,27,0.66,-0.27,-0.44,78,241,-79,-8,589,163,0,0,0,48,0.4,2,2,79,1,1,81,461,-14,-92,765,700,5
19604 16087114,26,0.71,-0.43,-0.49,73,237,-77,-17,589,146,0,0,0,47,0.61,2,76,1,07,89,450,-9,-68,752,684,5
19605 1899,26,0.8,-0.83,-0.69,64,243,-60,-54,583,166,0,0,0,44,0.71,2,1,1,26,66,425,-14,2,712,630,5
19606 139,23,1.04,-1.38,-0.59,43,214,-73,-141,553,180,0,0,0,41,1,09,1,35,1,64,63,393,-24,67,653,537,5
19607 106,23,1.11,-1.33,-0.71,30,215,-73,-177,540,172,0,0,0,40,1,2,1,3,1,76,38,391,-35,89,633,512,5
19608 13196,23,1.17,-1.27,-0.79,78,213,-75,-189,535,189,0,0,0,41,1,27,1,28,1,82,40,394,-44,91,629,488,5
19609 158974,22,1.32,-1,32,-0.89,11,200,-80,-219,523,185,0,0,0,42,1,33,1,14,1,84,52,401,-53,109,613,460,5
19610 190316,20,0.51,-0,5,-0.72,-63,99,-154,-568,257,36,0,0,0,39,1,4,-3,1,-0,59,-201,323,5,-706,273,620,5
19611 13207178,20,0.48,-0,43,-0,79,-58,99,-164,-567,241,44,0,0,0,37,1,38,-2,94,-0,52,-193,307,8,-699,229,596,5
19612 1,7372696,20,0.56,-0.42,-0.79,-49,92,-165,-574,248,29,0,0,0,36,1,38,-2,94,-0,48,-193,291,3,-683,152,555,5
19613 1,5303382,20,0.66,-0.39,-0.72,-52,87,-167,-576,232,42,0,0,0,34,1,46,-3,08,-0,49,-203,266,-4,-672,119,540,5
19614 1,3970337,18,0.61,-0.32,-0.49,-55,77,-152,-586,220,26,0,0,0,33,1,38,-2,86,-0,56,-227,236,-1,-622,-11,480,5
19615 1,0376856,19,0.55,-0.37,-0.34,-53,67,-160,-595,211,24,0,0,0,32,1,36,-2,83,-0,61,-227,221,0,-592,-85,431,5
19616 1,3,9202735,19,0.58,-0,48,-0,39,-53,57,-169,-591,210,12,0,0,0,31,1,33,-2,89,-0,71,-230,200,-8,-571,-126,404,5
19617 16,9489792,19,0,43,-0,4,-0,97,-45,58,-173,-609,165,1,0,0,0,25,1,51,-2,18,-1,02,-241,33,-17,-377,-420,173,5
19618 i130785,19,0,34,-0,31,-0,51,-42,70,-167,-624,127,8,0,0,0,27,1,75,-1,91,-0,38,-255,-50,-30,-226,-570,27,5
19619 1,4998335,19,0,32,-0,26,-0,36,-42,66,-168,-618,134,0,0,0,0,29,1,73,-1,75,-0,25,-271,-68,-37,-205,-587,6,5
19620 14,5256342,18,0,24,-0,24,0,05,-41,62,-164,116,116,0,0,0,29,1,59,-1,36,0,-271,-91,-43,-151,-635,-36,5
19621 i18037,19,0,22,-0,27,0,21,-38,54,-170,-621,113,-9,0,0,0,29,1,54,-1,2,0,05,-263,-99,-114,-551,-70,5
19622 1458728,19,0,13,-0,14,0,34,-40,42,-176,-626,116,0,0,0,0,32,1,48,-0,9,0,05,-270,-141 Col 55: class 98,5
19623 1,0,36,-38,38,-176,-627,119,2,0,0,0,33,1,38,-0,64,0,08,-278,-159,-52,-60,-686,-110,5
```



02

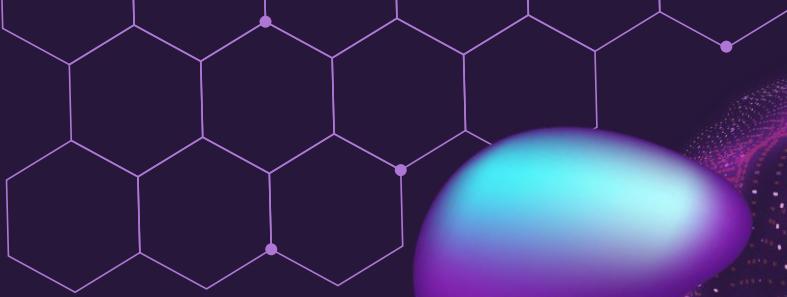


TRAINING AI MODEL

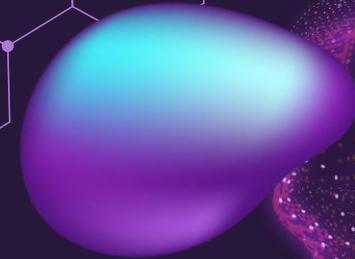
+ hyperparameters



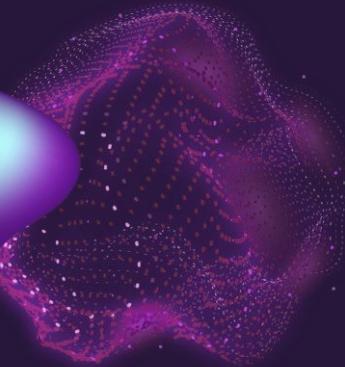
+



x

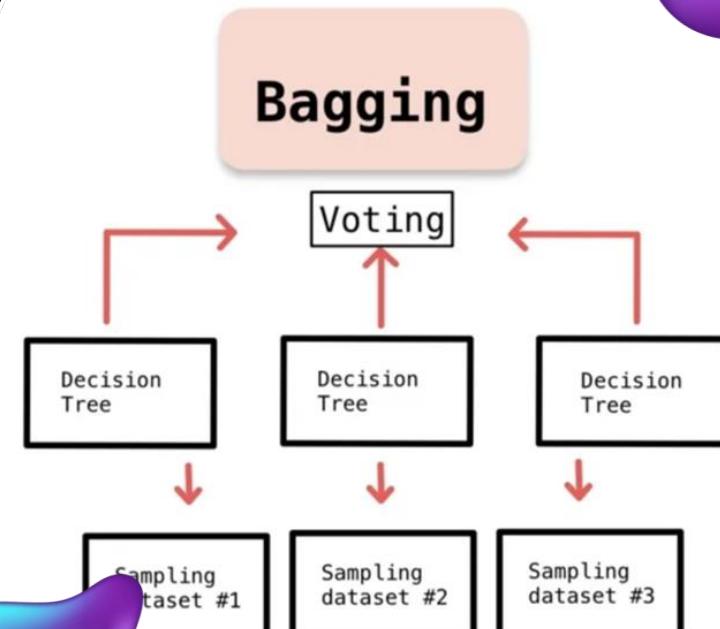


x



MODEL FAMILY

Ensemble learning (bagging)

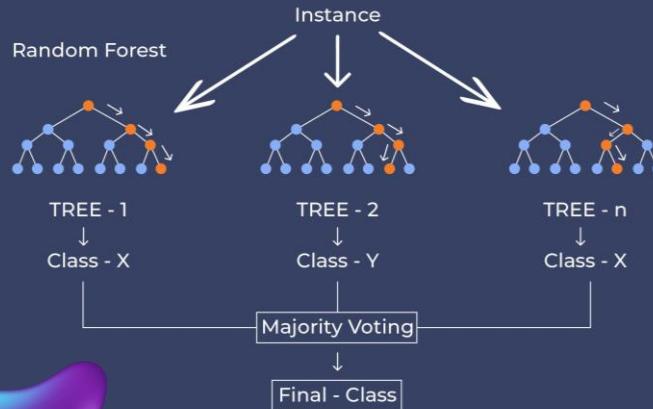


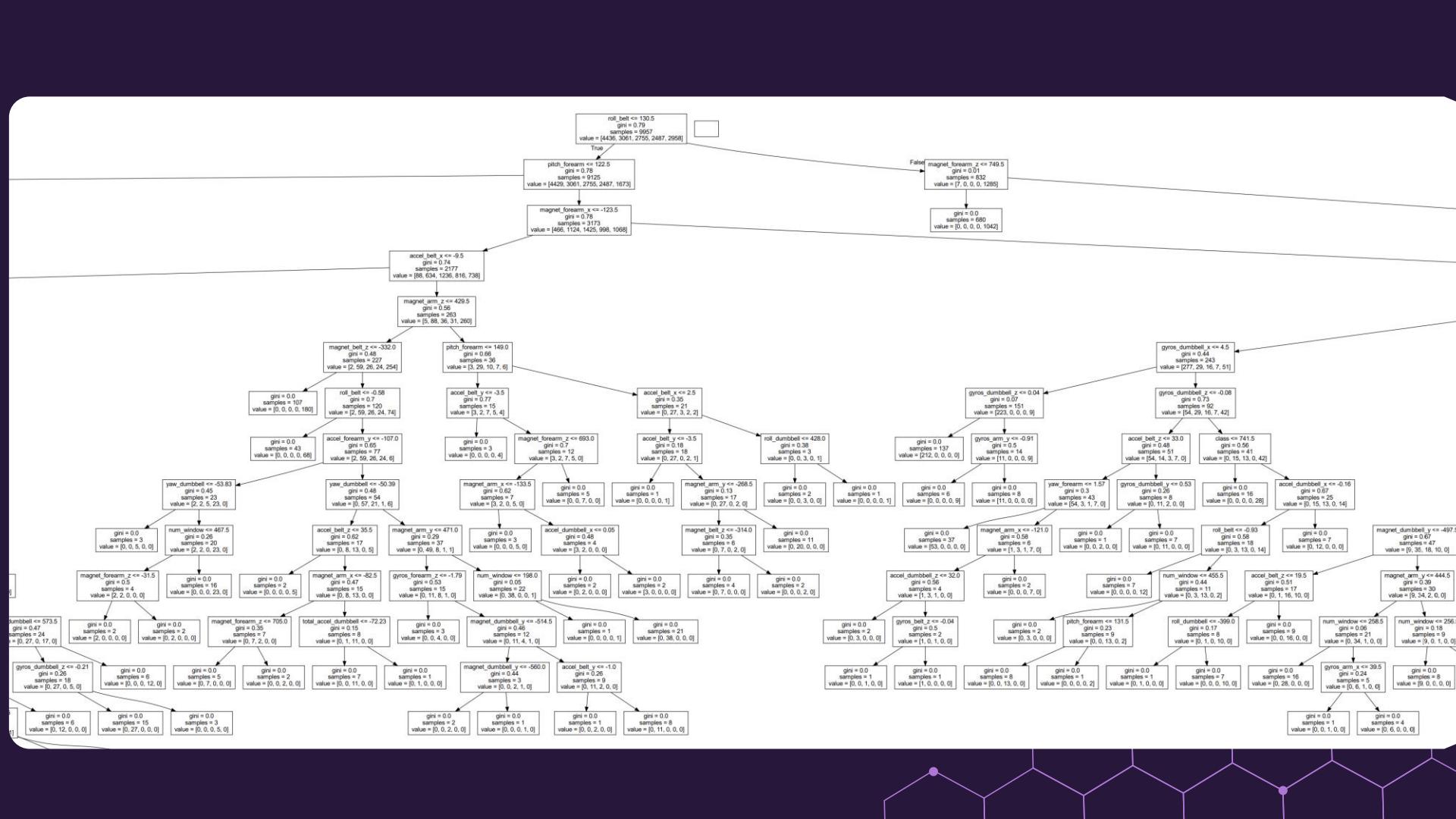
RANDOM FOREST CLASSIFIER

From sklearn, Random Forest consists of fully grown trees built on bootstrapped data and the majority vote rule to make predictions.

RANDOM FOREST

CLASSIFICATION





HYPERPARAMETERS

x



SPLIT

We split 20% of the training data for testing

x



ESTIMATORS

We started from $n_estimators = 100$, then incremented by 50 until it reached >95% accuracy at 200

x



METRICS

x

x

on the 20% testing data, after hyperparameters tuning

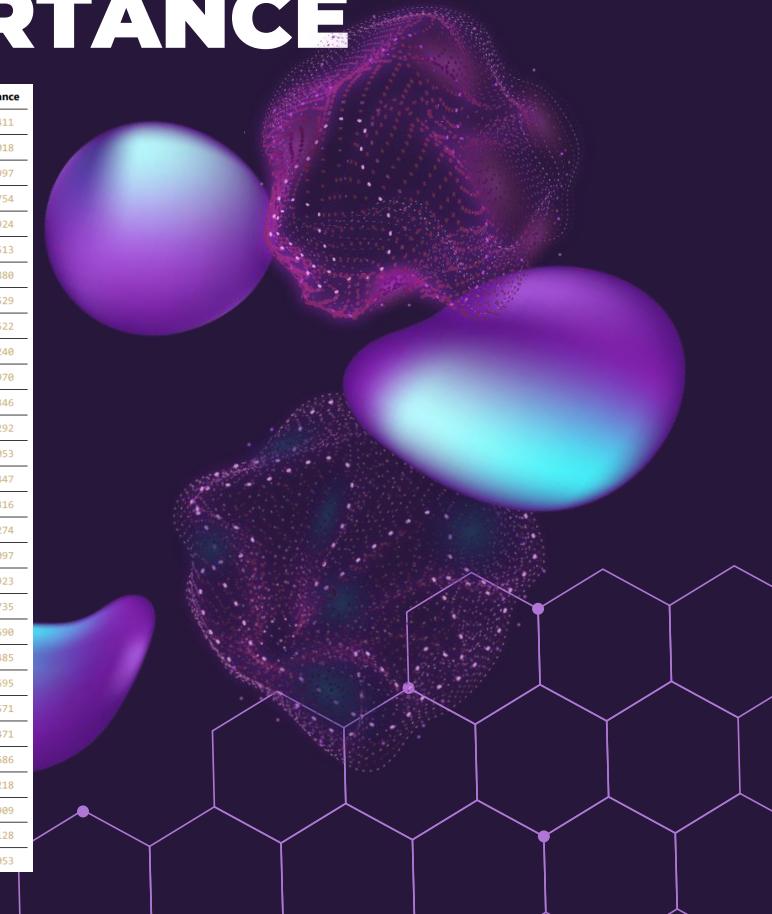
<u>ACCURACY</u>	0.9982165605095541																									
<u>Precision</u>	0.9982171198022268																									
<u>Recall</u>	0.9982165605095541																									
<u>CONFUSION MATRIX</u>	<table><tbody><tr><td>1097</td><td>0</td><td>0</td><td>0</td><td>0</td></tr><tr><td>2</td><td>731</td><td>0</td><td>0</td><td>0</td></tr><tr><td>0</td><td>1</td><td>685</td><td>0</td><td>0</td></tr><tr><td>0</td><td>0</td><td>2</td><td>668</td><td>1</td></tr><tr><td>0</td><td>0</td><td>0</td><td>1</td><td>737</td></tr></tbody></table>	1097	0	0	0	0	2	731	0	0	0	0	1	685	0	0	0	0	2	668	1	0	0	0	1	737
1097	0	0	0	0																						
2	731	0	0	0																						
0	1	685	0	0																						
0	0	2	668	1																						
0	0	0	1	737																						



The metrics are already very good, so we didn't further tune the hyperparameters. ˘＼(ツ)／-

✗ FEATURE IMPORTANCE

Feature label	Importance	Feature label	Importance
num_window	0.092442	gyros_dumbbell_y	0.014411
roll_belt	0.074710	accel_forearm_z	0.014018
yaw_belt	0.051930	accel_arm_x	0.013997
magnet_dumbbell_z	0.045757	total_accel_belt	0.013754
pitch_forearm	0.045721	yaw_arm	0.012924
magnet_dumbbell_y	0.043380	magnet_arm_y	0.012513
pitch_belt	0.041529	magnet_forearm_x	0.011880
roll_forearm	0.035012	magnet_forearm_y	0.011529
magnet_dumbbell_x	0.030000	pitch_dumbbell	0.010522
roll_dumbbell	0.026281	pitch_arm	0.010240
magnet_belt_y	0.025142	yaw_forearm	0.009970
accel_dumbbell_y	0.024722	magnet_arm_z	0.009346
accel_belt_z	0.024186	accel_arm_y	0.008292
magnet_belt_z	0.022496	accel_belt_y	0.008053
accel_dumbbell_z	0.020494	accel_forearm_y	0.007447
accel_forearm_x	0.019814	gyros_arm_y	0.007316
roll_arm	0.017512	gyros_arm_x	0.007274
magnet_belt_x	0.016070	gyros_belt_y	0.007097
yaw_dumbbell	0.015871	accel_belt_x	0.006923
magnet_arm_x	0.015695	gyros_dumbbell_x	0.006735
total_accel_dumbbell	0.015565	accel_arm_z	0.006690
gyros_belt_z	0.015538	gyros_forearm_y	0.006485
magnet_forearm_z	0.015358	total_accel_forearm	0.005695
accel_dumbbell_x	0.014625	gyros_belt_x	0.005471
		gyros_dumbbell_z	0.004686
		gyros_forearm_z	0.004218
		gyros_forearm_x	0.003989
		gyros_arm_z	0.003128
		new_window	0.000053



x

CLASS PREDICTION

The unseen data was pre-processed the same manner as described before.



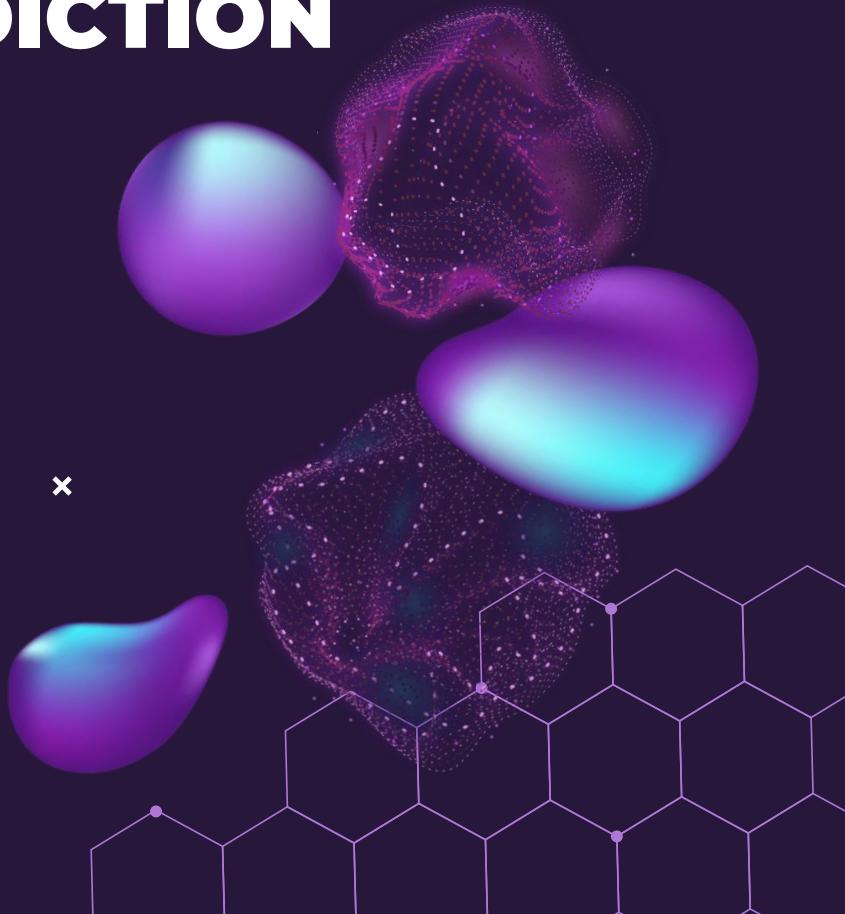
Prediction results:

[A B D B B E D E A B A A C A D B C E D B]

* We had to change the 1/2/3/4/5 back into A/B/C/D/E due to the pre-processing steps

x

x





A dark purple background featuring abstract geometric patterns. In the upper left, a hexagonal grid of light purple lines forms a partial hexagon. In the lower left, a larger hexagonal shape is composed of a dense network of small dots and lines, with a large, glowing blue-purple sphere resting on it. In the upper right, a single dot is connected by lines to several other dots, forming a small cluster. In the lower right, another hexagonal grid of light purple lines is visible. The overall aesthetic is futuristic and minimalist.

FIN.