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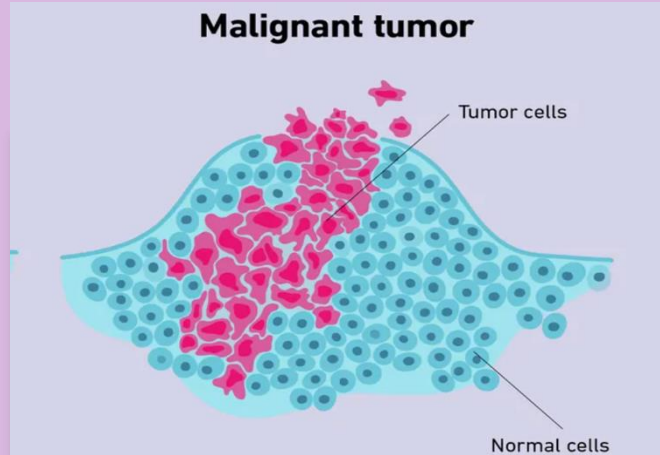
Results of our prediction model and its accuracy.

INTRODUCTION

In this presentation, we will look at the two different types of tumors, and how we can use various features to distinguish one type from another. We will delve into how we applied Discriminant Analysis on the characteristic data of tumors such as size, shape, texture, location, and **histopathological** features to accurately classify tumors. Earlier identification of the type of tumor can ultimately lead to better diagnosis and treatment options for patients.

Histopathological – study of diseased cells and tissues using a microscope.

MALIGNANT AND BENIGN TUMORS



Malignant Tumors:

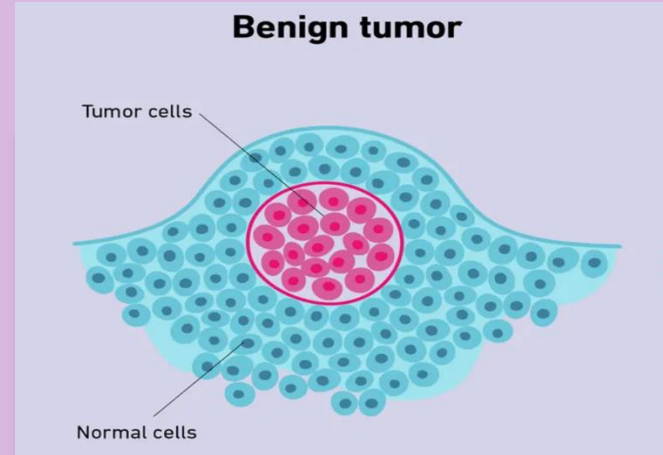
Cancerous

May invade surrounding tissue

Most grow rapidly*

Irregular shape

Needs treatment



Benign Tumors:

Not cancerous

Doesn't invade surrounding tissue

Most grow slowly*

Smooth shape

May not need treatment



DATASET

SOURCE: KAGGLE

The dataset contains 570 rows of different tumor data containing 30 different features of the tumor and their actual classification(as Benign or Malignant). The following are the attributes in this dataset.

1	id	9	concave points_mean	17	compactness_se	25	area_worst
2	radius_mean	10	symmetry_mean	18	concavity_se	26	smoothness_worst
3	texture_mean	11	fractal_dimension_mean	19	concave points_se	27	compactness_worst
4	perimeter_mean	12	radius_se	20	symmetry_se	28	concavity_worst
5	area_mean	13	texture_se	21	fractal_dimension_se	29	concave points_worst
6	smoothness_mean	14	perimeter_se	22	radius_worst	30	symmetry_worst
7	compactness_mean	15	area_se	23	texture_worst	31	fractal_dimension_worst
8	concavity_mean	16	smoothness_se	24	perimeter_worst	32	Diagnosis



OPTIMIZATION

Objective:

The main goal is to maximize the **Accuracy** of Cancer Prediction using **Discriminant Analysis**.

- Predicting the cancer with a very high accuracy improves the business of a diagnostic centre.

Best Feature Selection:

- We have 30 columns(features of tumor) in our dataset but most of them are redundant and doesn't contribute to the determination of the type of tumor. So, we used Genetic Algorithm to find the best 5 high quality features that are relevant to the problem.
- We found this using 'GeneticSelectionCV' module from sklearn-genetic package in Python.
The best features are shown in the adjacent image.

	Features	Score
23	area_worst	112598.431564
3	area_mean	53991.655924
13	area_se	8758.504705
22	perimeter_worst	3665.035416
2	perimeter_mean	2011.102864
20	radius_worst	491.689157
0	radius_mean	266.104917
12	perimeter_se	250.571896
21	texture_worst	174.449400
1	texture_mean	93.897508
26	concavity_worst	39.516915
10	radius_se	34.675247
6	concavity_mean	19.712354
25	compactness_worst	19.314922
27	concave points_worst	13.485419
7	concave points_mean	10.544025



OPTIMIZATION

Discriminant Analysis (optimization model that we adopted):

- It is a statistic tool used in business and marketing by analysts to classify/categorize certain data. If a high % of data is correctly classified, then the Analysis is said to be successful.

Objective Function: Accuracy (of prediction) – must be Maximized.

Features deciding the prediction of tumor: area_worst, area_mean, area_se, perimeter_worst, perimeter_mean

Constraints for Cutoff value of score: $-7735.3 < \text{Cutoff} < 7735.3$

Score = SUMPRODUCT(Features,Weights).

Since the Maximum value of the sum of the 5 features is 7735.3, the cut-off value cannot be greater than 7735.3(if weights = 1) and cannot be smaller than -7735.3(if weights = -1)

Constraints for Weights of features: $-1 < \text{Weights} < 1$



OPTIMIZATION RESULTS & ACCURACY

Discriminate the type of tumor/cancer based on its features using Diuscriminant Analysis (Features selected using Genetic Algorithm)															
Weights for discriminant function															
	area_worst	area_mean	area_se	perimeter_mean	perimeter_worst										
	0.9603204	0.03475749	0.2413133	0.047735802	0.642361846					M - Malignant					
										B - Benign					
Cutoff value for classification															
	954.40045													Max	
														7735.3	
										Classification Matrix					
										(Actual along side, predicted along top)				Sum of features	
id	area_worst	area_mean	area_se	perimeter_mean	perimeter_worst	diagnosis	Score	Prediction			M	B			
842302	2019	1001	153.4	122.8	184.6	M	2135.1386	M						3480.8	
842517	1956	1326	74.08	132.9	158.8	M	2050.7028	M		M	176	36		3647.78	
84300903	1709	1203	94.03	130	152.5	M	1809.8574	M		B	8	349		3288.53	
84348301	567.7	386.1	27.23	77.58	98.87	M	632.3784	B						1157.48	
84358402	1575	1297	94.44	135.1	152.2	M	1684.5914	M						3253.74	
843786	741.6	477.1	27.19	82.57	103.4	M	805.6795	B						1431.86	
844359	1606	1040	53.91	119.6	153.2	M	1695.5506	M						2972.71	
84458202	897	577.9	50.96	90.2	110.6	M	969.1421	M						1726.66	
844981	739.3	519.8	24.32	87.5	106.2	M	806.29629	B						1477.12	
84501001	711.4	475.9	23.94	83.97	97.65	M	772.2251	B						1392.86	
845636	1150	797.8	40.51	102.7	123.8	M	1226.3005	M						2214.81	
84610002	1299	781	54.16	103.6	136.5	M	1380.2992	M						2374.26	
846226	1332	1123	116.2	132.4	151.7	M	1449.9866	M						2855.3	
846381	876.5	782.7	36.58	103.7	112	M	954.64752	M						1911.48	
84667401	697.7	578.3	19.21	93.6	108.8	M	769.10849	B						1497.61	
84799002	943.2	658.8	32.55	96.73	124.1	M	1020.8618	M						1855.38	
848406	1138	684.5	45.4	94.74	123.4	M	1211.3817	M						2086.04	
84862001	1315	798.8	54.18	108.1	136.8	M	1396.6954	M						2412.88	
849014	2398	1260	112.4	130	186.8	M	2499.9653	M						4087.2	
8510426	711.2	566.3	23.56	87.46	99.7	B	776.56685	B						1488.22	
8510653	630.5	520	14.67	85.63	96.09	B	692.90816	B						1346.89	
8510824	314.9	273.9	15.7	60.34	65.13	B	360.43101	B						729.97	



OPTIMIZATION RESULTS & ACCURACY

Total number of tumor samples	569
Number of samples correctly predicted as Malignant	176
Number of samples correctly predicted as Benign	349
Number of Benign tumors wrongly predicted as Malignant	8
Number of Malignant tumors wrongly predicted as Benign	36
Accuracy (Percentage of tumor samples that are correctly predicted)	$= (349+176)/569$ $= 0.9227$ $= \mathbf{92.27\%}$

CONCLUSION:

- We managed to achieve a very good Maximum Accuracy of 92.27% by choosing the best features for Discriminative analysis using Genetic Algorithm. (We previously got a maximum accuracy of only 70% when we used all the features for Discriminative Analysis)
- This prediction can also be achieved by Multiple Linear Regression by training and testing a model. We wish to perform this and compare the results and accuracy of both models.

**THANKS
FOR YOUR
ATTENTION!!**

ANY QUESTIONS!?

