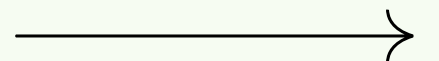


DATAVISION

Healthcare Dataset

Benítez, Figueroa, Gavenda e Ibañez



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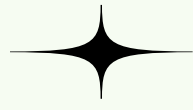
Resultados

05

Métricas

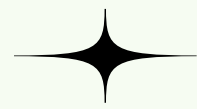
06

Conclusión



Problema

- Tiempos de espera prolongados para obtener resultados
 - Demoras en el diagnóstico y tratamiento debido a la alta demanda
 - Recursos hospitalarios (laboratorios, médicos, equipamiento) limitados.
 - Dificultad para priorizar pacientes según nivel de urgencia
-

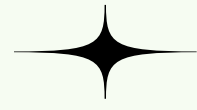


Necesidad

Surge la necesidad de contar con herramientas que permitan anticipar el resultado probable de un test médico, incluso antes de que esté disponible, utilizando variables clínicas y administrativas recogidas durante la internación.

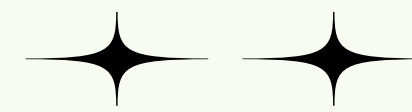


Modelo predictivo capaz de anticipar el resultado de un test médico realizado a un paciente en su ingreso al hospital



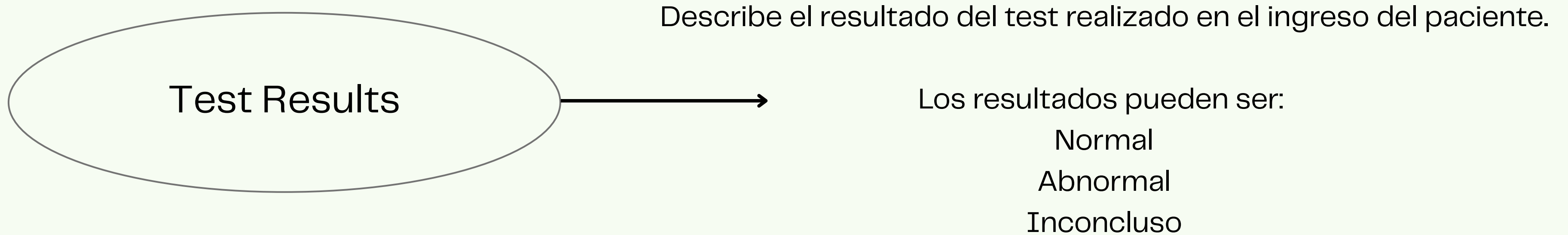
Beneficios del modelo

- Anticipación del diagnóstico
 - Priorización inteligente
 - Eficiencia operativa.
 - Reducción de costos
-

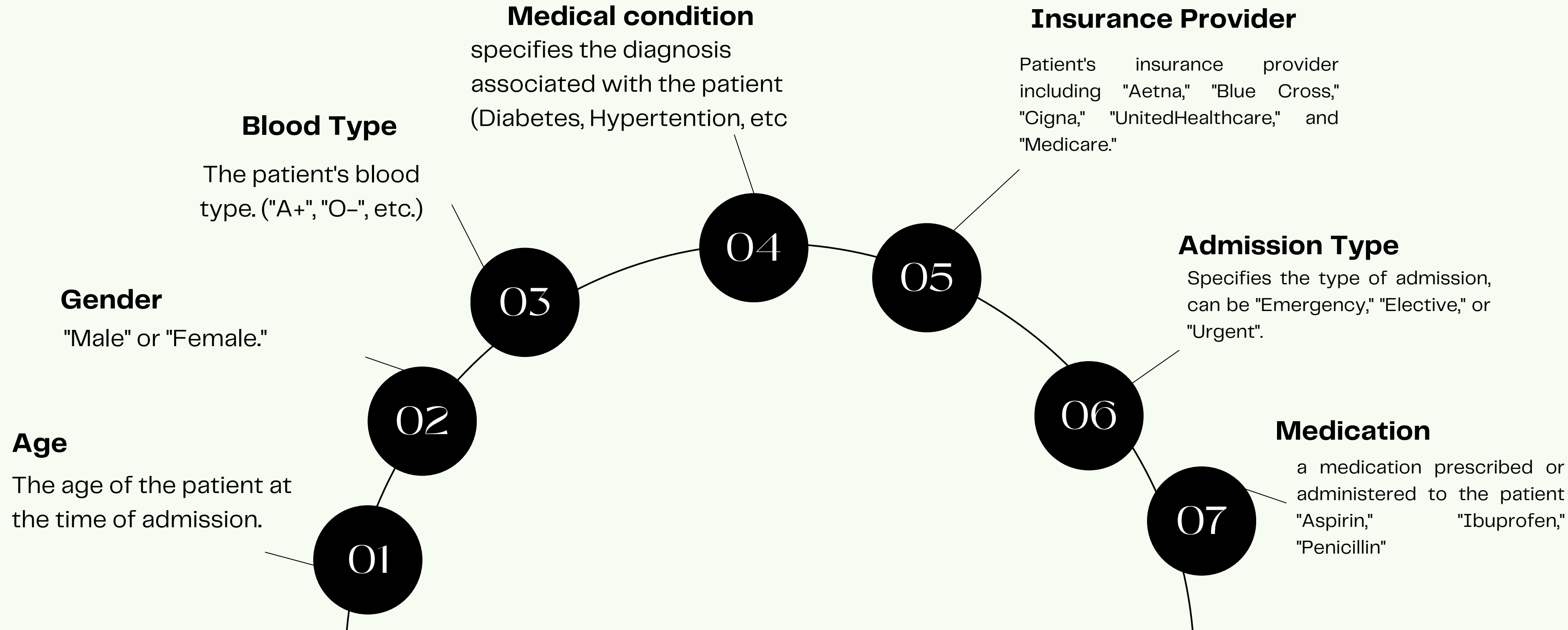


EDA

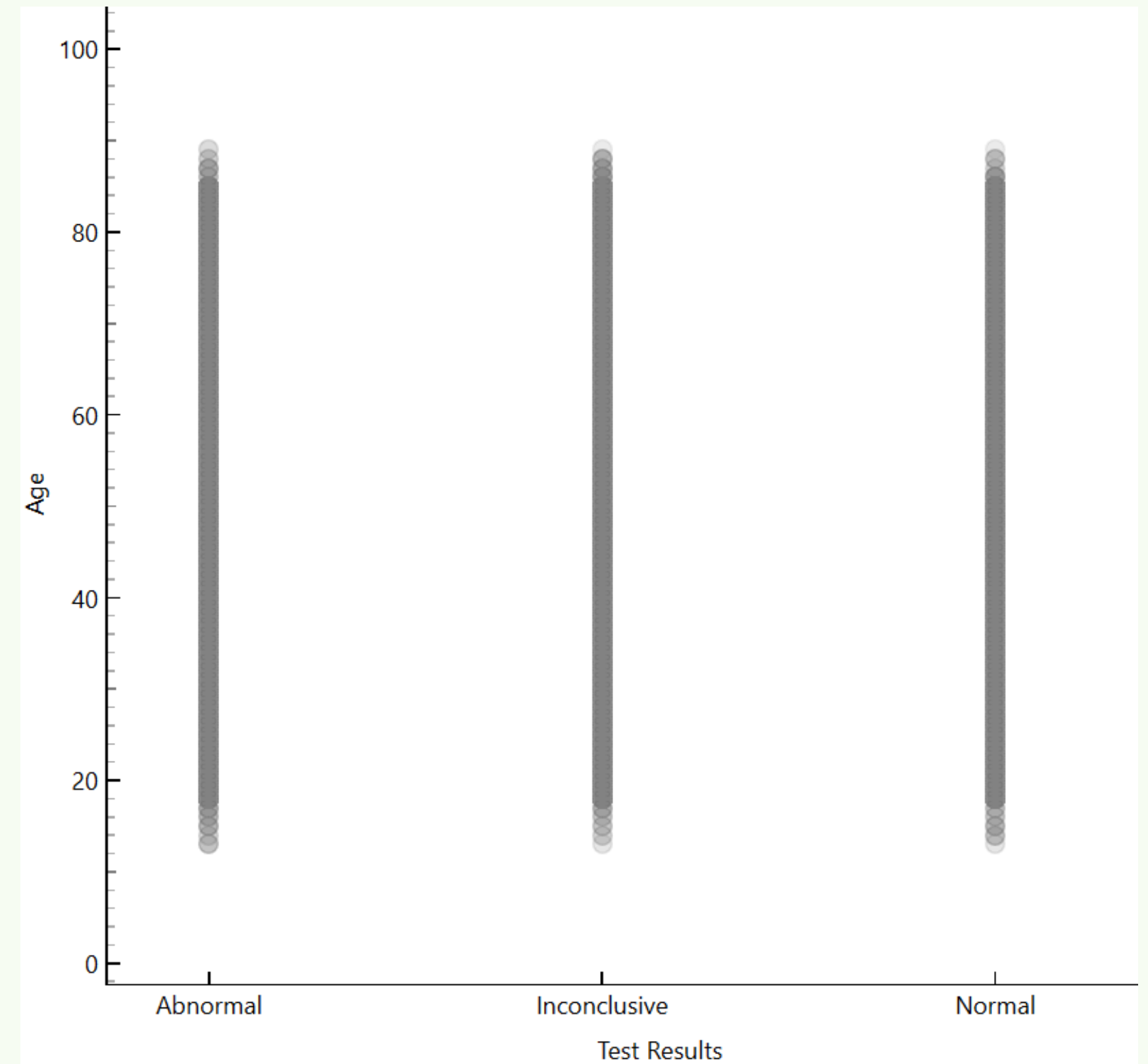
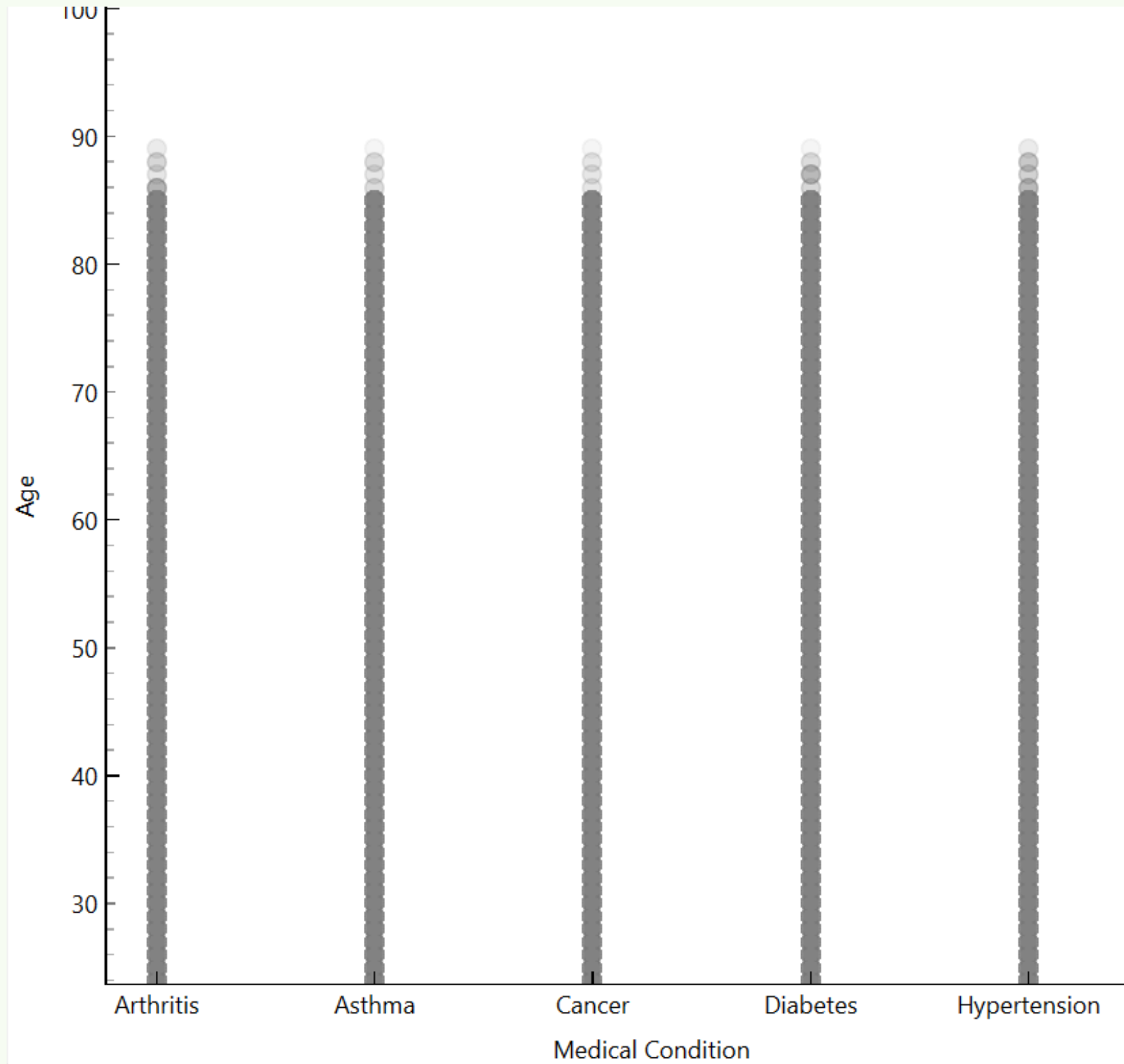
Variable Objetivo ✨




Variables Independientes



Scatter Plott



Feature Statistics

Name	Distribution	Mean	Mode	Median	Dispersion	Min.	Max.	Missing
N Age		51.54	38	52	0.38	13	89	0 (0 %)

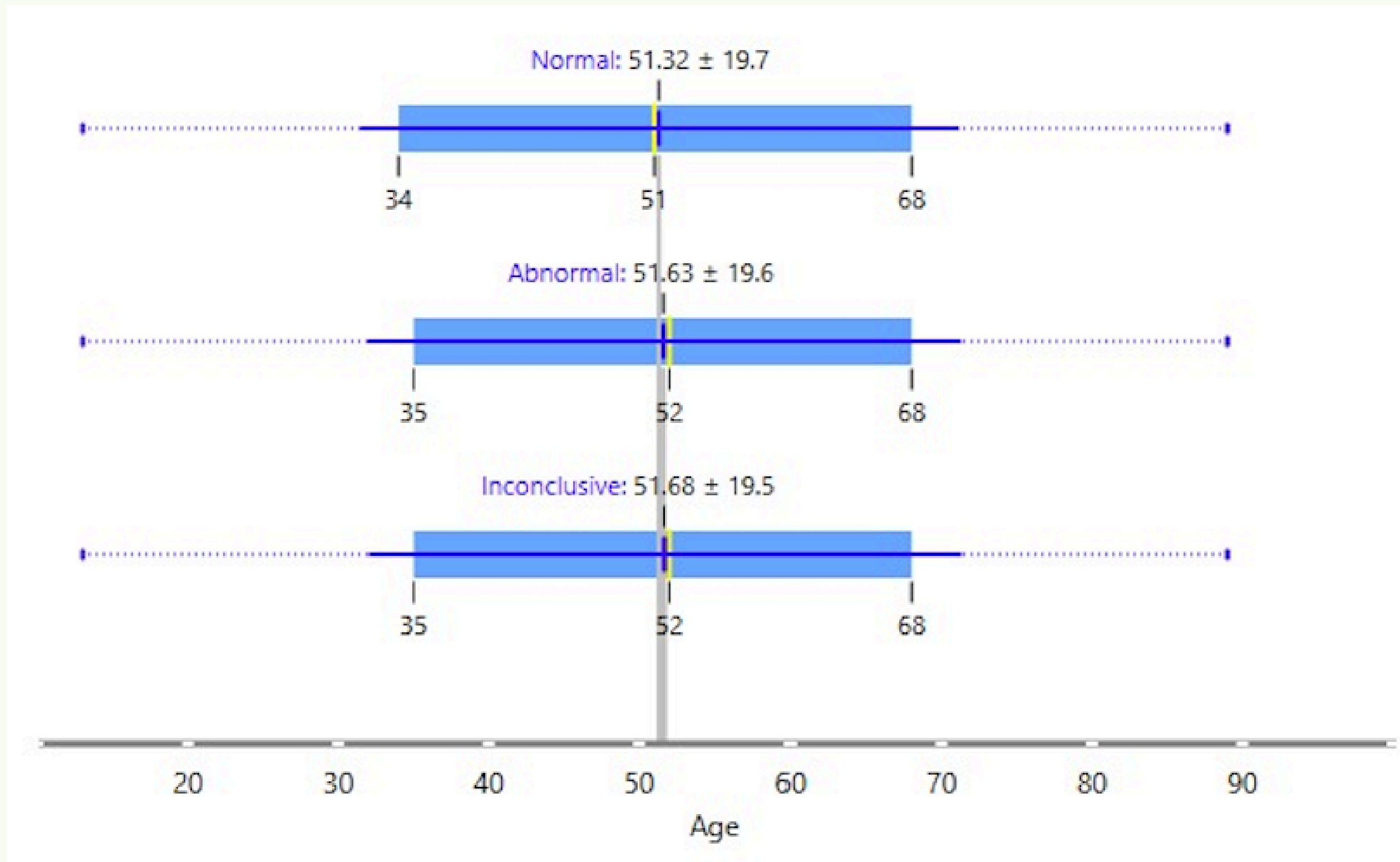
	Name	Distribution	Mean	Mode	Median	Dispersion	Min.	Max.	Missing
C	Blood Type			A-		2.08			0 (0 %)
C	Gender			Male		0.693			0 (0 %)
C	Insurance Provider			Cigna		1.61			0 (0 %)
C	Medical Condition			Arthritis		1.79			0 (0 %)
C	Medication			Lipitor		1.61			0 (0 %)
C	Test Results			Abnormal		1.1			0 (0 %)

Abnormal

Inconclusive

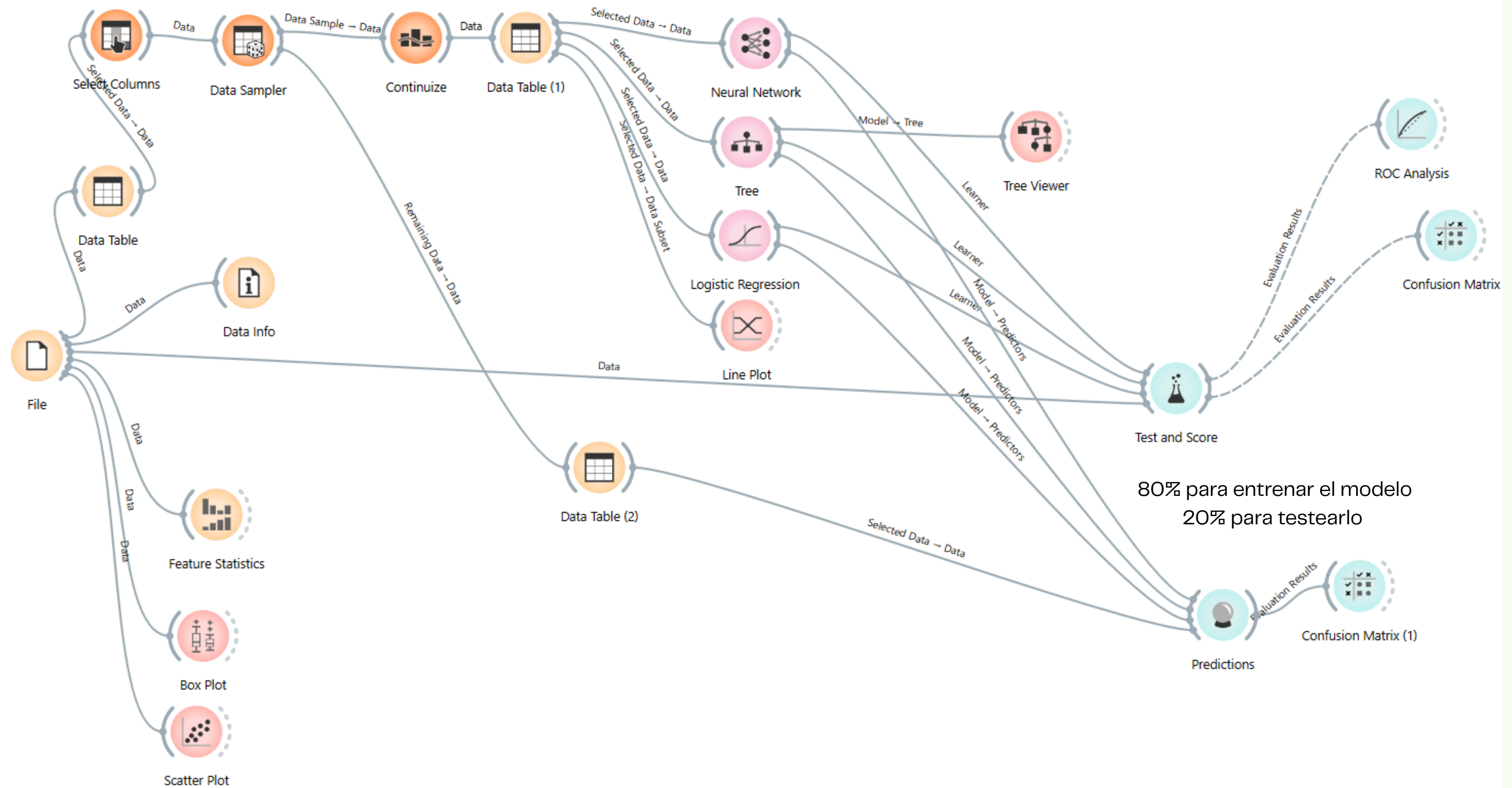
Normal

Box Plot

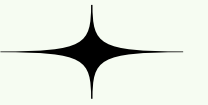


La edad no influye en el resultado del test

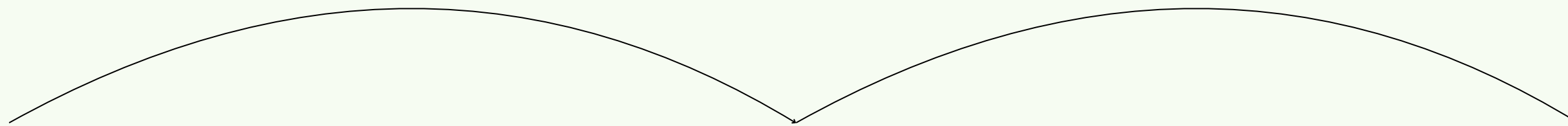
PIPELINE



80% para entrenar el modelo
20% para testearlo



Análisis de Métricas



F1

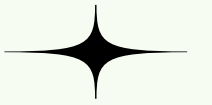
Representa el balance entre la precisión y el recall

RECALL

Capacidad del modelo para detectar correctamente los casos positivos reales (en este caso, pacientes con resultados “Abnormal”).

PRECISIÓN

De las veces que el modelo predijo positiva (“Abnormal”) cuantas veces acertó

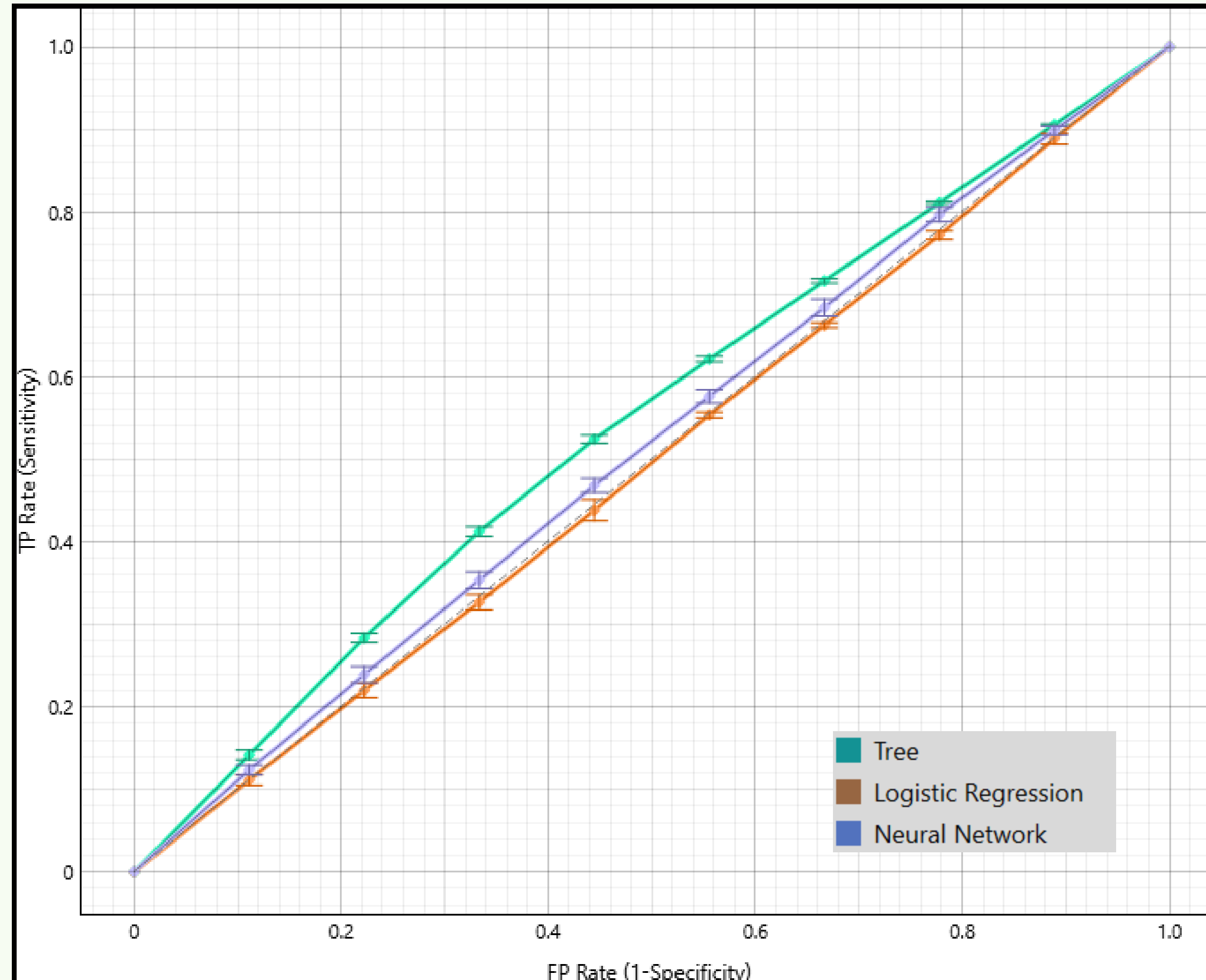


Análisis de Modelos

Model	AUC	CA	F1	Prec	Recall	MCC	
Tree	0.548	0.386	0.386	0.386	0.386	0.080	→ Mejor rendimiento
Neural Network	0.514	0.346	0.345	0.345	0.346	0.018	→ Leve mejora del rendimiento
Logistic Regression	0.499	0.334	0.330	0.334	0.334	0.000	→ Bajo rendimiento



Curvas ROC



Ninguno de los modelos logra una discriminación clara entre los casos positivos y negativos

MATRICES DE CONFUSIÓN

Tree

		Predicted			Σ
		Abnormal	Inconclusive	Normal	
Actual	Abnormal	1550	1126	1073	3749
	Inconclusive	1286	1292	1065	3643
	Normal	1250	1160	1298	3708
	Σ	4086	3578	3436	11100

		Predicted			Σ
		Abnormal	Inconclusive	Normal	
Actual	Abnormal	1442	1025	1282	3749
	Inconclusive	1415	1004	1224	3643
	Normal	1518	992	1198	3708
	Σ	4375	3021	3704	11100

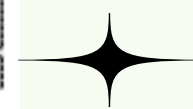
Regresión Logística

Los modelos cometen una alta proporción de errores en la clasificación de los resultados del test médico

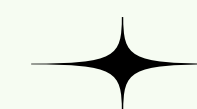
		Predicted			Σ
		Abnormal	Inconclusive	Normal	
Actual	Abnormal	1014	1497	1238	3749
	Inconclusive	916	1504	1223	3643
	Normal	910	1482	1316	3708
	Σ	2840	4483	3777	11100

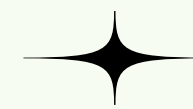
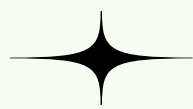
Red Neuronal

	Tree	error	Logistic Regression	error	Neural Network	error	Test Results
1	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.33 : 0.33 : 0.34 → Normal	0.657	0.30 : 0.54 : 0.16 → Inconclusive	0.835	Normal
2	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.34 : 0.32 : 0.34 → Abnormal	0.660	0.40 : 0.30 : 0.30 → Abnormal	0.596	Abnormal
3	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.32 : 0.34 : 0.34 → Inconclusive	0.665	0.30 : 0.32 : 0.39 → Normal	0.614	Normal
4	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.34 : 0.32 : 0.34 → Normal	0.682	0.34 : 0.28 : 0.39 → Normal	0.723	Inconclusive
5	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.35 : 0.31 : 0.34 → Abnormal	0.654	0.48 : 0.21 : 0.31 → Abnormal	0.519	Abnormal
6	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.36 : 0.32 : 0.33 → Abnormal	0.684	0.24 : 0.25 : 0.50 → Normal	0.747	Inconclusive
7	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.34 : 0.33 : 0.33 → Abnormal	0.668	0.19 : 0.55 : 0.26 → Inconclusive	0.738	Normal
8	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.33 : 0.34 : 0.33 → Inconclusive	0.661	0.42 : 0.30 : 0.28 → Abnormal	0.698	Inconclusive
9	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.35 : 0.32 : 0.33 → Abnormal	0.654	0.27 : 0.40 : 0.33 → Inconclusive	0.727	Abnormal
10	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.34 : 0.34 : 0.32 → Abnormal	0.677	0.23 : 0.44 : 0.33 → Inconclusive	0.668	Normal
11	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.34 : 0.33 : 0.34 → Normal	0.663	0.36 : 0.37 : 0.27 → Inconclusive	0.733	Normal
12	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.33 : 0.32 : 0.35 → Normal	0.667	0.43 : 0.18 : 0.40 → Abnormal	0.573	Abnormal
13	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.32 : 0.33 : 0.35 → Normal	0.667	0.24 : 0.26 : 0.50 → Normal	0.739	Inconclusive
14	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.36 : 0.33 : 0.31 → Abnormal	0.686	0.34 : 0.24 : 0.42 → Normal	0.580	Normal
15	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.33 : 0.33 : 0.34 → Normal	0.668	0.19 : 0.33 : 0.48 → Normal	0.675	Inconclusive
16	1.00 : 0.00 : 0.00 → Abnormal	1.000	0.34 : 0.32 : 0.34 → Normal	0.683	0.59 : 0.13 : 0.28 → Abnormal	0.865	Inconclusive
17	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.35 : 0.33 : 0.32 → Abnormal	0.654	0.23 : 0.43 : 0.35 → Inconclusive	0.774	Abnormal
18	1.00 : 0.00 : 0.00 → Abnormal	0.000	0.35 : 0.33 : 0.33 → Abnormal	0.647	0.25 : 0.43 : 0.33 → Inconclusive	0.654	Abnormal



	Tree	error	Logistic Regression	error	Neural Network	error	Test Results
1136	0.00 : 0.00 : 1.00 → Normal	0.000	0.35 : 0.32 : 0.33 → Abnormal	0.669	0.22 : 0.43 : 0.35 → Inconclusive	0.646	Normal
1137	0.00 : 0.00 : 1.00 → Normal	0.000	0.35 : 0.33 : 0.32 → Abnormal	0.677	0.30 : 0.45 : 0.25 → Inconclusive	0.751	Normal
1138	0.00 : 0.00 : 1.00 → Normal	0.000	0.33 : 0.33 : 0.33 → Abnormal	0.667	0.13 : 0.36 : 0.51 → Normal	0.488	Normal
1139	0.00 : 0.00 : 1.00 → Normal	1.000	0.33 : 0.32 : 0.35 → Normal	0.680	0.30 : 0.43 : 0.28 → Inconclusive	0.573	Inconclusive
1140	0.00 : 0.00 : 1.00 → Normal	0.000	0.34 : 0.32 : 0.34 → Normal	0.662	0.33 : 0.49 : 0.18 → Inconclusive	0.822	Normal
1141	0.00 : 0.00 : 1.00 → Normal	0.000	0.32 : 0.33 : 0.35 → Normal	0.653	0.26 : 0.41 : 0.33 → Inconclusive	0.666	Normal
1142	0.00 : 0.00 : 1.00 → Normal	0.000	0.33 : 0.34 : 0.33 → Inconclusive	0.666	0.21 : 0.41 : 0.39 → Inconclusive	0.613	Normal
1143	0.00 : 0.00 : 1.00 → Normal	0.000	0.32 : 0.34 : 0.34 → Inconclusive	0.663	0.27 : 0.52 : 0.21 → Inconclusive	0.792	Normal
1144	0.00 : 0.00 : 1.00 → Normal	1.000	0.32 : 0.34 : 0.34 → Normal	0.677	0.28 : 0.34 : 0.38 → Normal	0.723	Abnormal
1145	0.00 : 0.00 : 1.00 → Normal	1.000	0.33 : 0.32 : 0.34 → Normal	0.677	0.16 : 0.16 : 0.68 → Normal	0.836	Inconclusive
1146	0.00 : 0.00 : 1.00 → Normal	1.000	0.34 : 0.33 : 0.33 → Abnormal	0.670	0.34 : 0.25 : 0.41 → Normal	0.755	Inconclusive
1147	0.00 : 0.00 : 1.00 → Normal	1.000	0.34 : 0.33 : 0.33 → Abnormal	0.660	0.33 : 0.55 : 0.13 → Inconclusive	0.673	Abnormal
1148	0.00 : 0.00 : 1.00 → Normal	0.000	0.33 : 0.34 : 0.34 → Inconclusive	0.664	0.36 : 0.51 : 0.12 → Inconclusive	0.876	Normal
1149	0.00 : 0.00 : 1.00 → Normal	0.000	0.36 : 0.33 : 0.31 → Abnormal	0.692	0.35 : 0.25 : 0.40 → Normal	0.600	Normal
1150	0.00 : 0.00 : 1.00 → Normal	0.000	0.34 : 0.33 : 0.33 → Abnormal	0.667	0.18 : 0.36 : 0.46 → Normal	0.536	Normal
1151	0.00 : 0.00 : 1.00 → Normal	1.000	0.34 : 0.33 : 0.33 → Abnormal	0.663	0.33 : 0.27 : 0.40 → Normal	0.671	Abnormal
1152	0.00 : 0.00 : 1.00 → Normal	0.000	0.33 : 0.33 : 0.34 → Normal	0.662	0.40 : 0.25 : 0.35 → Abnormal	0.647	Normal





¡Muchas Gracias!