





The liver is a vital visceral organ that plays a crucial role in various metabolic processes, including digestion, detoxification, regulation of blood sugar levels, and production of important proteins. Liver diseases can range from mild to life-threatening illnesses.

Most of the common liver diseases are:

- Fatty liver disease associated with excessive alcohol, diabetes and obesity
- Hepatitis: inflammation caused by viruses
- Cirrhosis: advanced scarring of the liver tissue caused by long-term damage, often due to chronic alcoholism, viral hepatitis, or other factors
- Primary liver cancer (hepatocellular carcinoma)
 originates in the liver cells and is often associated with
 underlying liver disease.



Problem Analysis and Motivation

- Problem: Accurate and early diagnosis of liver disease is critical for effective treatment and patient well-being.
- Motivation: Improving diagnosis can lead to better healthcare outcomes, reduced medical costs, and enhanced patient care.



Diagnosis

Blood tests or Liver function tests which is the basis of this project in the major way of diagnosing liver disease measures enzymes, bilirubin, and proteins to assess how well the liver is functioning.



Improving Healthcare service delivery - Contribute to enhanced patient care, reduced medical costs, and improved healthcare outcomes through reliable diagnosis.

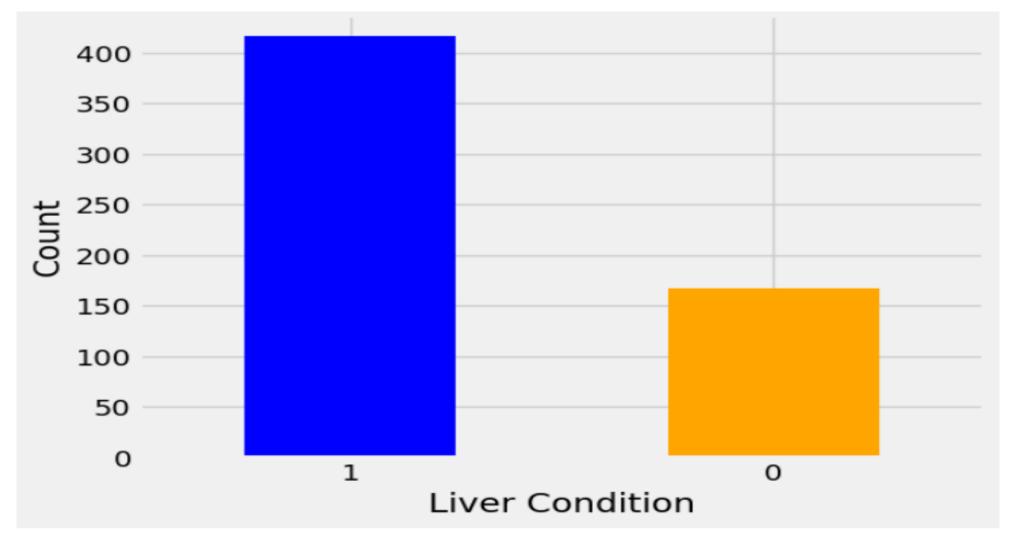


Early Detection - Identify features that contribute to early detection of liver disease, enabling timely medical interventions.

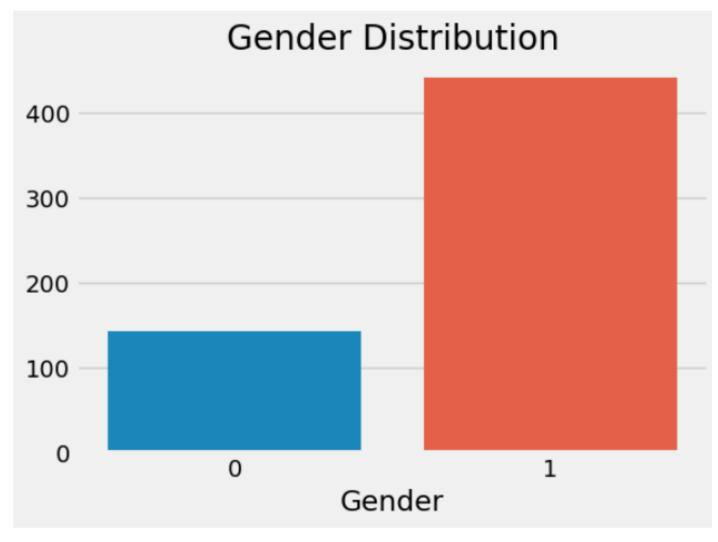


Accurate Diagnosis - Develop predictive models to accurately diagnose liver disease based on patient data.

Dataset Description

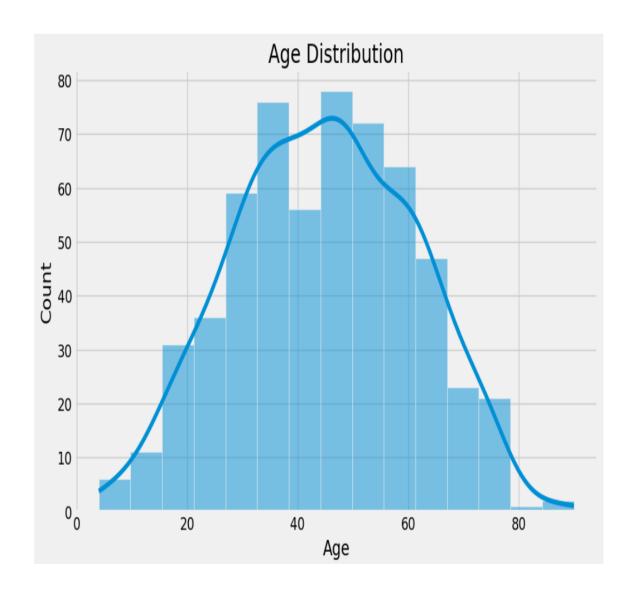


Number of patients diagnosed with liver disease: 416 Number of patients with normal liver: 167



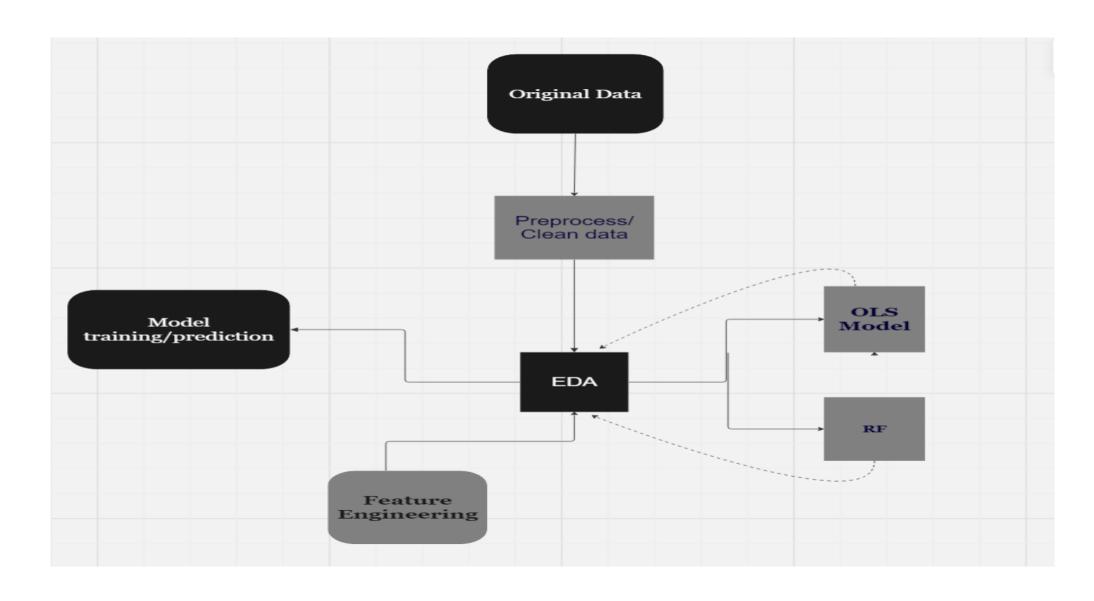
Number of patients that are male: 441 Number of patients that are female: 142

Age distribution



	LiverCondition	Gender	Mean_Age
0	0	0	42.740000
1	0	1	40.598291
2	1	0	43.347826
3	1	1	46.950617

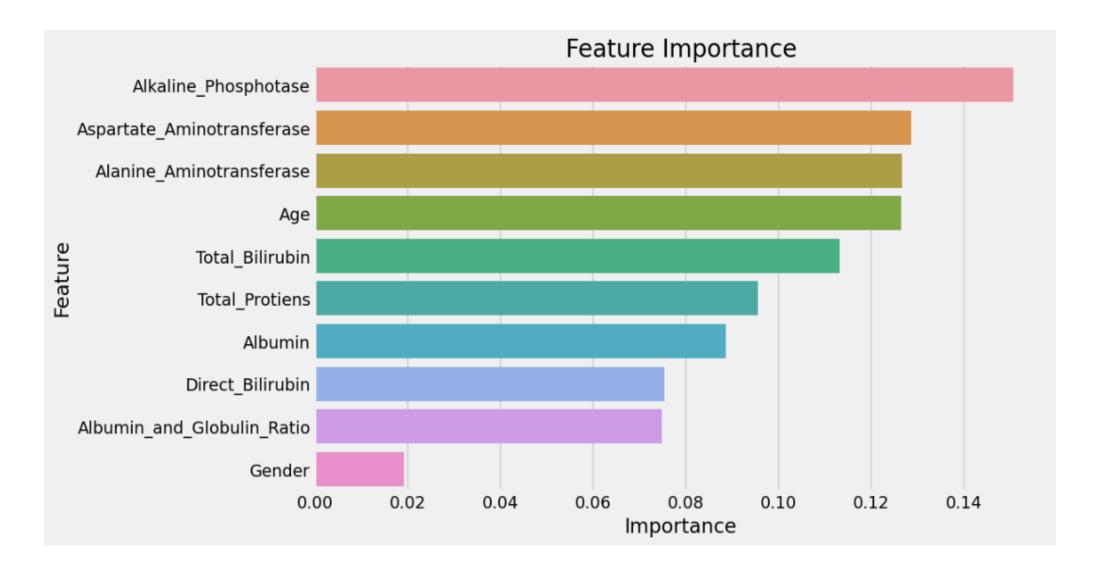
Approach



OLS Regression Results

Model:	iverCondition OLS	R-squared: Adj. R-squ			0.120 0.105	
	Least Squares		F-statistic:		7.812	
	, 12 Aug 2023	•		8.3		
Time:	12:14:18		nood:	-:	327.11	
No. Observations:	583	AIC:			676.2	
Df Residuals:	572	BIC:			724.3	
Df Model:	10					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	0.3088	0.162	1.903	0.058	-0.010	0.628
Age	0.0032	0.001	2.804	0.005	0.001	0.005
Gender	0.0510	0.042	1.214	0.225	-0.032	0.133
Total_Bilirubin	0.0001	0.006	0.023	0.982	-0.011	0.012
Direct_Bilirubin	0.0225	0.013	1.677	0.094	-0.004	0.049
Alkaline_Phosphotase	0.0002	7.74e-05	2.517	0.012	4.28e-05	0.000
Alanine_Aminotransferase	0.0004	0.000	2.367	0.018	6.57e-05	0.001
Aspartate_Aminotransferas	se -6.355e-05	0.000	-0.617	0.537	-0.000	0.000
Total_Protiens	0.0760	0.039	1.972	0.049	0.000	0.152
Albumin	-0.1391	0.071	-1.959	0.051	-0.279	0.000
Albumin_and_Globulin_Rat:	io 0.0545	0.107	0.509		-0.156	0.265
Omnibus:	343.370	Durbin-Wat		========	1.943	
Prob(Omnibus):	0.000			7		
Skew:	-0.689	Prob(JB):	-	6.2	23e-18	
Kurtosis:	1.833	Cond. No.		4.5	51e+03	

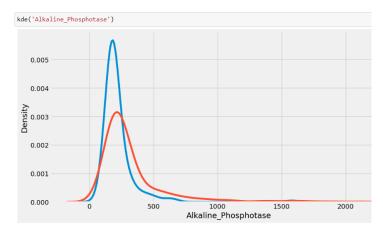
RandomForest

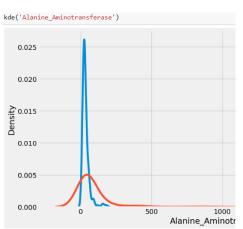


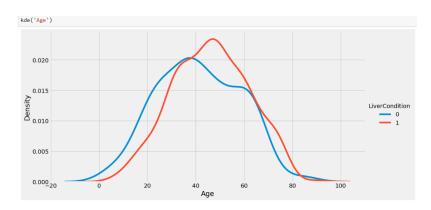
Heatmap

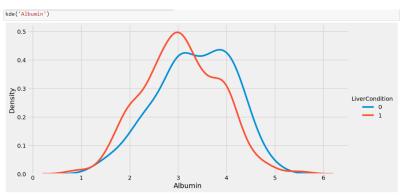
Age	1	0.057	0.012	0.0075	0.08	-0.087	-0.02	-0.19	-0.27	-0.22	0.14	1.0
Gender	0.057	1	0.089	0.1	-0.027	0.082	0.08	-0.089	-0.094	-0.0034	0.082	0.8
Total_Bilirubin	0.012	0.089	1	0.87	0.21	0.21	0.24	-0.0081	-0.22	-0.21	0.22	
Direct_Bilirubin	0.0075	0.1	0.87	1	0.23	0.23	0.26	-0.00014	-0.23	-0.2	0.25	0.6
Alkaline_Phosphotase	0.08	-0.027	0.21	0.23	1	0.13	0.17	-0.029	-0.17	-0.23	0.18	
Alanine_Aminotransferase	-0.087	0.082	0.21	0.23	0.13	1	0.79	-0.043	-0.03	-0.0024	0.16	0.4
Aspartate_Aminotransferase	-0.02	0.08	0.24	0.26	0.17	0.79	1	-0.026	-0.085	-0.07	0.15	0.2
Total_Protiens	-0.19	-0.089	-0.0081	-0.00014	-0.029	-0.043	-0.026	1	0.78	0.23	-0.035	0.2
Albumin	-0.27	-0.094	-0.22	-0.23	-0.17	-0.03	-0.085	0.78	1	0.69	-0.16	0.0
Albumin_and_Globulin_Ratio	-0.22	-0.0034	-0.21	-0.2	-0.23	-0.0024	-0.07	0.23	0.69	1	-0.16	
LiverCondition	0.14	0.082	0.22	0.25	0.18	0.16	0.15	-0.035	-0.16	-0.16	1	-0.2
	Age	Gender	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphotase	Alanine_Aminotransferase	Aspartate_Aminotransferase	Total_Protiens	Albumin	Albumin_and_Globulin_Ratio	LiverCondition	

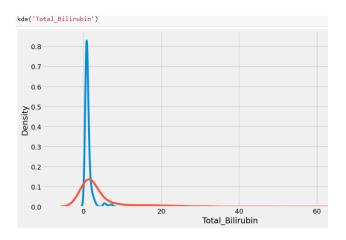
Kernel density

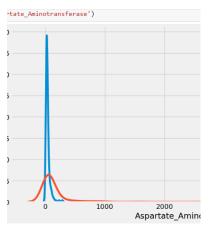


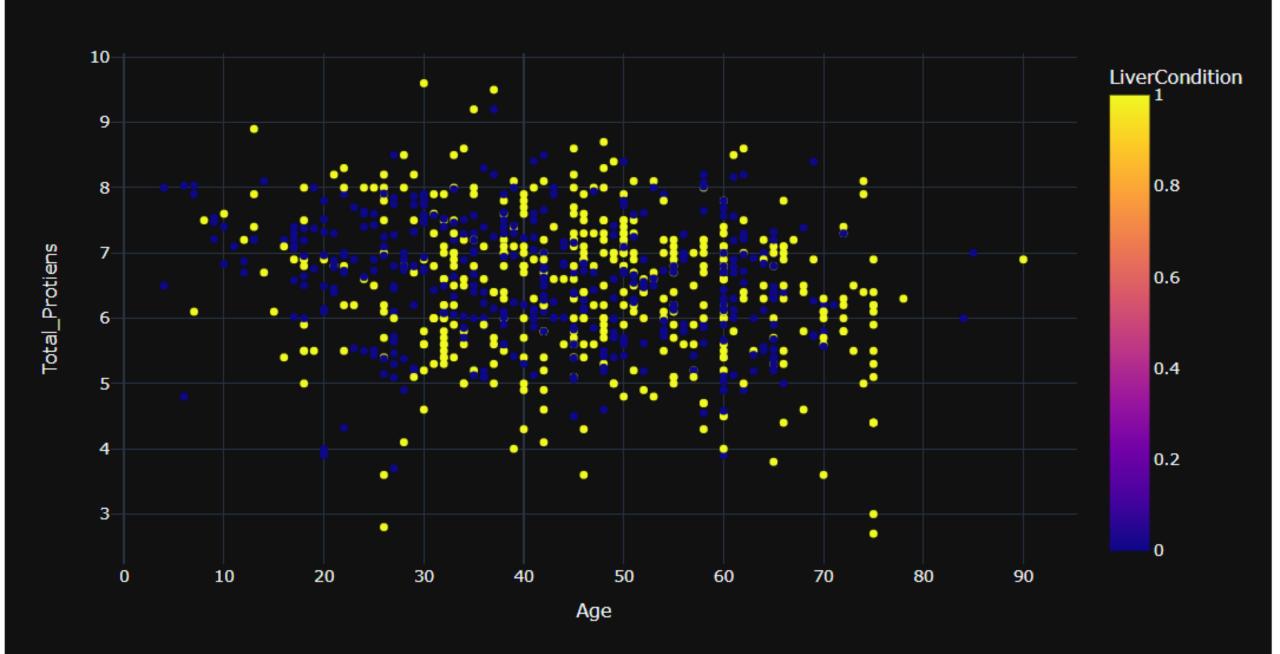


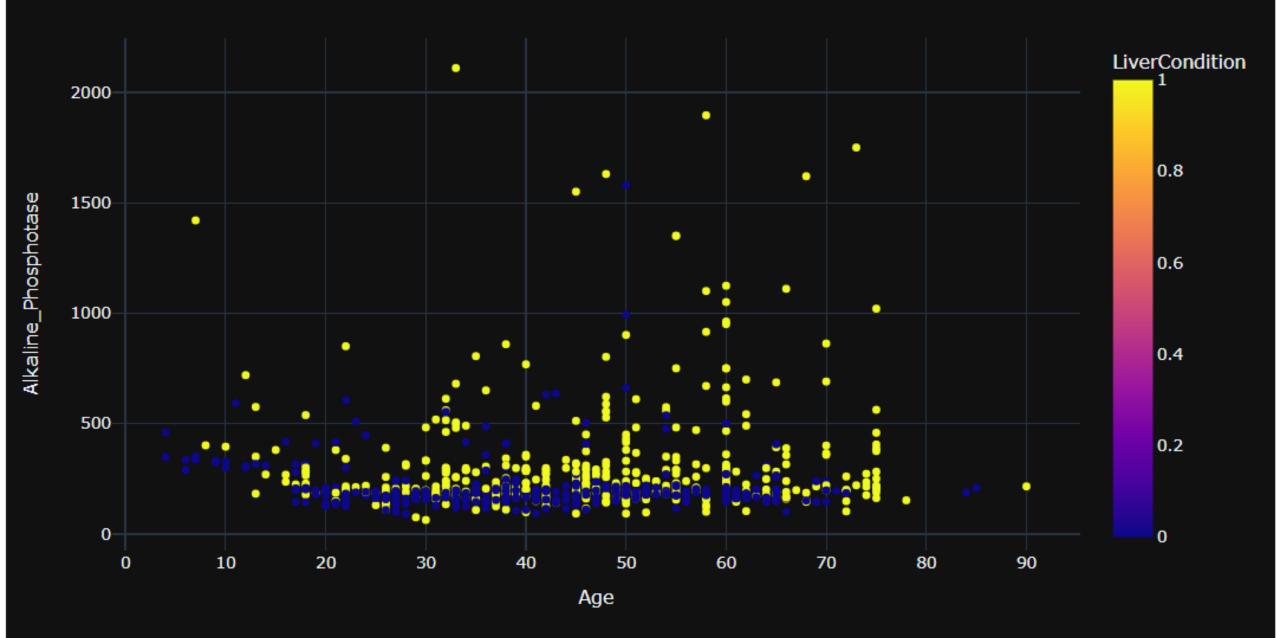


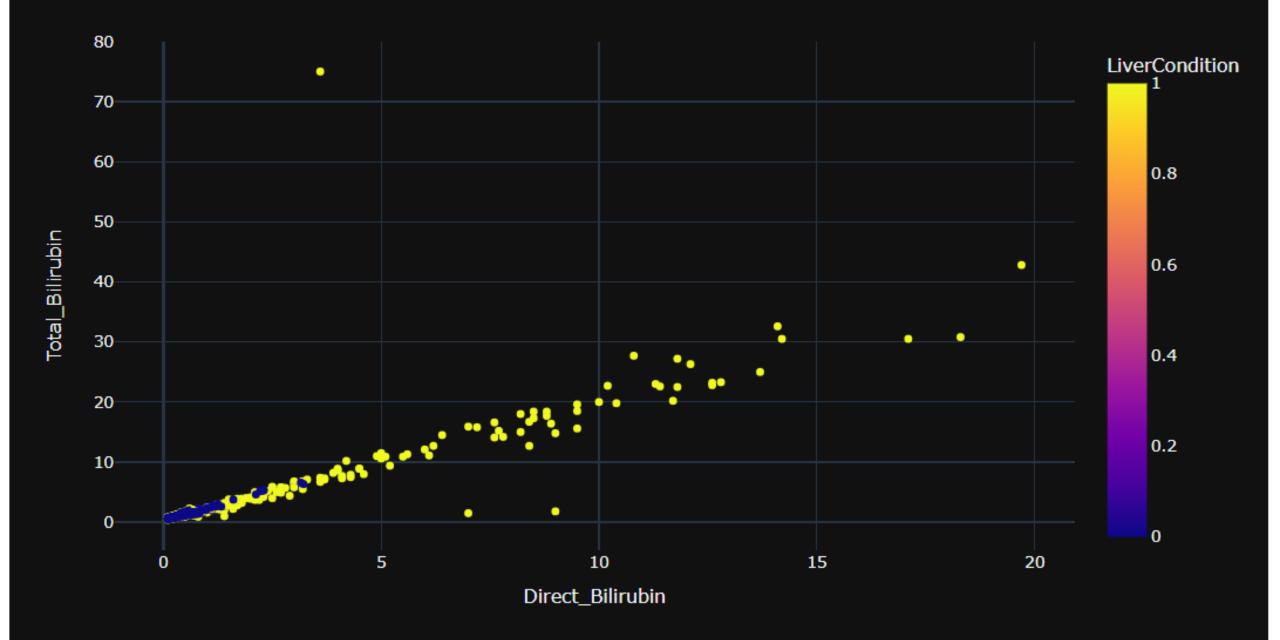


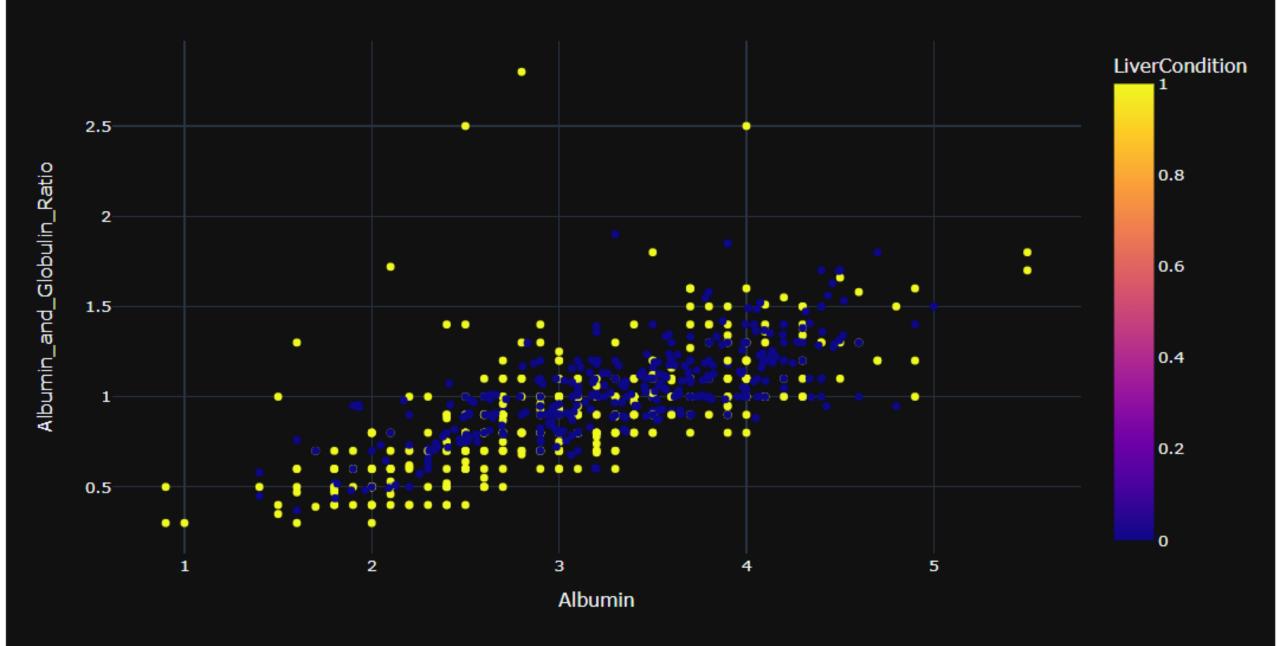


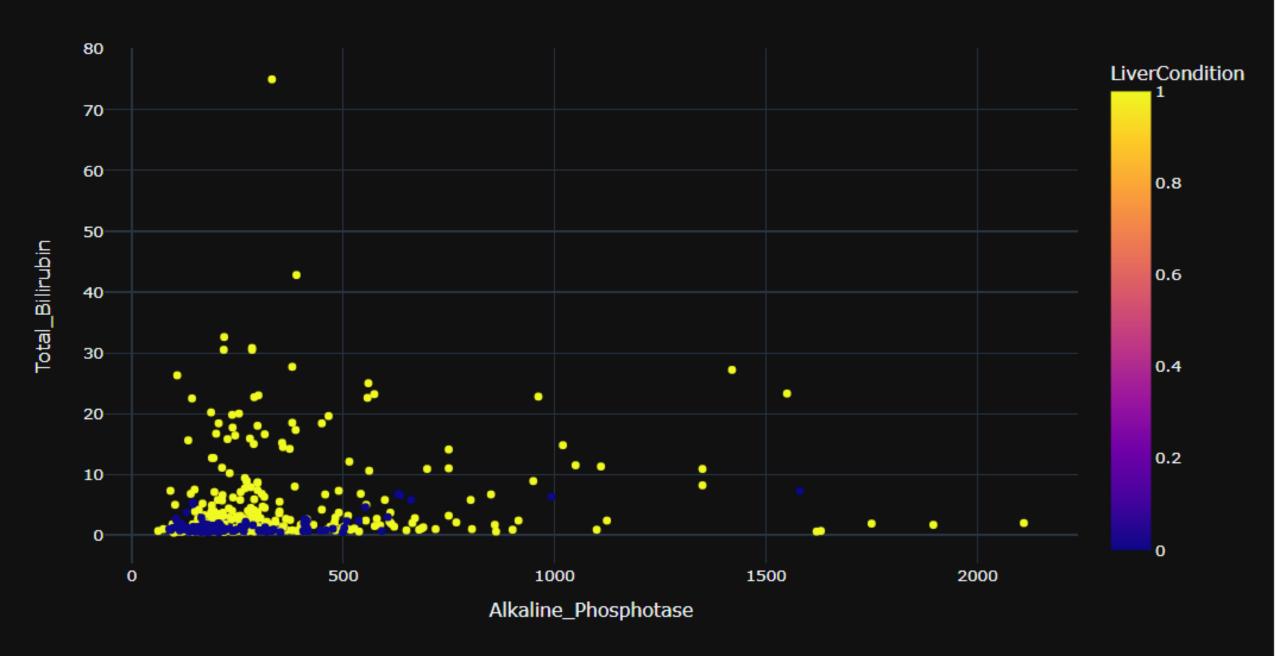


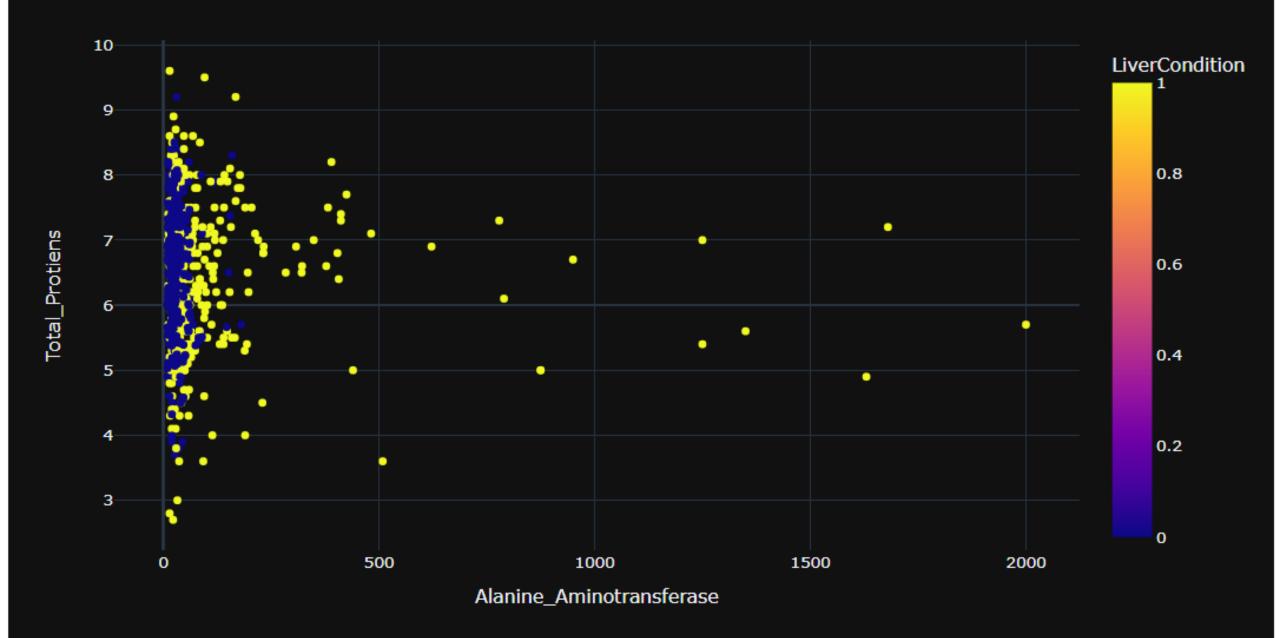












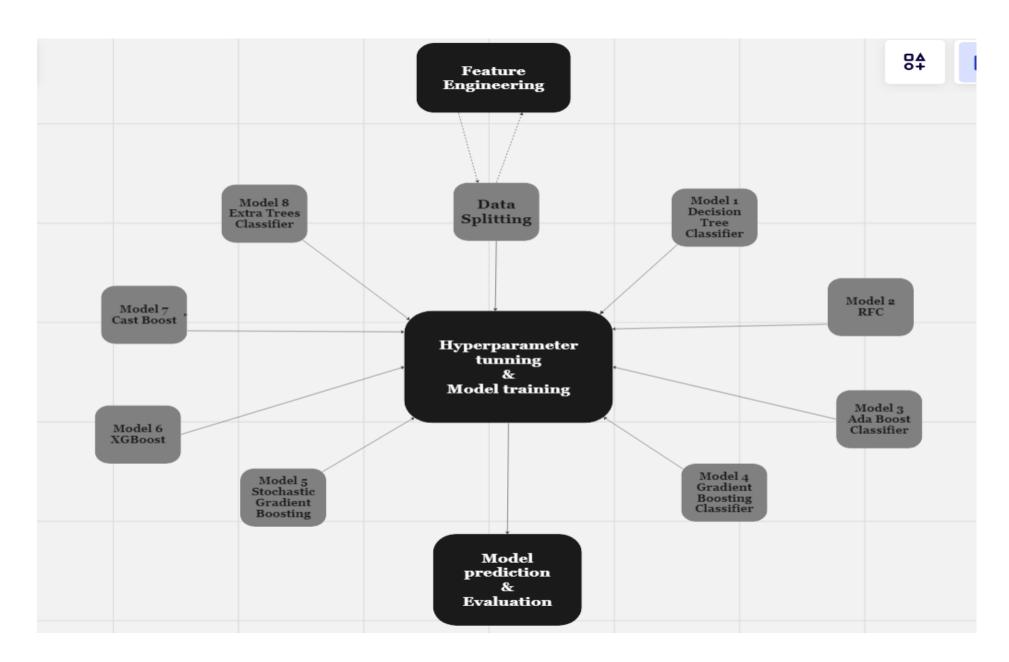
T-test Results

S/N	Features	Liver conditionCategory	Result(P-value)	Conclusion
1	Alkaline_Phosphotase	Normal + Diseased	7.03e-06 (<0.05)	Reject null: Significant difference
2	Alanine_Aminotransfera	Normal + Diseased	7.37e-05 (<0.05)	Reject null: Significant difference
3	Aspartate_Aminotransfe	Normal + Diseased	7.80e-08 (<0.05)	Reject null: Significant difference
4	Total_Bilirubin	Normal + Diseased	9.07e-05 (<0.05)	Reject null: Significant difference
5	Albumin	Normal + Diseased	8.84e-04 (<0.05)	Reject null: Significant difference
6	Age	Normal + Diseased	2.31e-04 (<0.05)	Reject null: Significant difference

Observations:

- Certain features exhibit direct relationships among themselves, such as:
- Protein features Albumin and Total Proteins; Albumin and Albumin-to-Globulin Ratio.
- Enzyme features Aspartate Aminotransferase and Alanine Aminotransferase also show a direct relationship.
- Bilirubin features Direct Bilirubin and Total Bilirubin

Model Schemas



Model Name	Test Accuracy
XGBoost	83.2%
Extra Trees Classifier	83.2%
Random Forest Classifier	80.8%
Ada Boost Classifier	79.6%
Cat Boost	79.6%
Gradient Boosting Classifier	79.0%
Stochastic Gradient Boosting	79.0%
Decision Tree Classifier	73.7%



What Worked: Certain features, such as enzyme(Alkaline Phosphatase & Alanine Aminotransferase), Total Protein and Total Bilirubin levels, showed significant impact on liver disease prediction.

What Didn't: Some models performed better than others; Decision Tree had comparatively lower accuracy.

Benefit: Accurate prediction models can aid in timely diagnosis and treatment decisions, potentially improving patient outcomes.

Conclusion:

- Based on the analysis, healthcare providers should focus on certain features that play a significant role in diagnosing liver disease for early diagnosis and treatment.
- For accurate predictions, we recommend using machine learning models such as XGBoost or Extra Trees Classifier.



Next Steps:

• Deployment: Planned possible collaboration with diagnostic centers to deploy the best predictive model using their patient dataset.

Thank You