

Identifying Best Classification Algorithm

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Dataset Generation

employment_status:

1. Employed
2. Self Employed
3. Unemployed

education_level:

1. Bachelors
2. Masters
3. PhD

credit_score:

1. Bad
2. Fair
3. Good

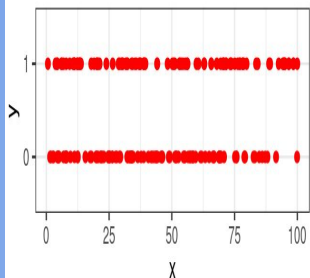
loan_eligibility:

0 -> not eligible
1 -> eligible

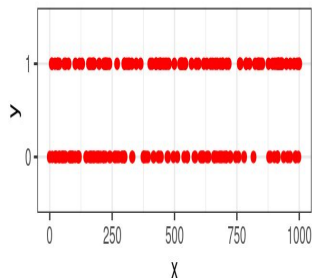
```
set.seed(22)
## Generating one binary/dichotomous dependent variable
loan_eligibility <- as.factor(sample(x = 0:1, size = 200, replace = TRUE))
## Generating two continuous independent variables
monthly_income <- runif(n = 200, min = 0, max = 100)
prev_loan_paid <- runif(n = 200, min = 0, max = 1000)
## Generate three categorical independent variables
employment_status <- as.factor(sample(x = c("Employed", "Self Employed", "Unemployed"), size = 200, replace = TRUE))
education_level <- as.factor(sample(x = c("Bachelors", "Masters", "PHD"), size = 200, replace = TRUE))
credit_score <- as.factor(sample(x = c("Bad", "Fair", "Good"), size = 200, replace = TRUE))
```

Bivariate Logistic Regression Analysis

monthly_income vs loan_eligibility



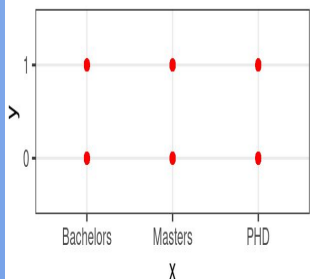
prev_loan_paid vs loan_eligibility



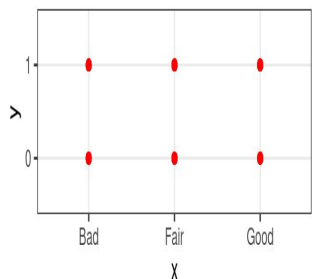
employment_status vs loan_eligibility



education_level vs loan_eligibility



credit_score vs loan_eligibility



Independent Variable	Accuracy	Sensitivity	Specificity	Confusion Matrix
monthly_income_in_K	0.5676	0.6500	0.4706	<div><div>01</div><div>0139</div><div>178</div></div>
prev_loan_paid_in_K	0.6216	0.7000	0.5294	<div><div>01</div><div>0148</div><div>169</div></div>
employment_status	0.3784	0.5500	0.1765	<div><div>01</div><div>01114</div><div>193</div></div>
education_level	0.5405	1.0000	0.000	<div><div>01</div><div>02017</div><div>100</div></div>
credit_score	0.5946	0.4500	0.7647	<div><div>01</div><div>094</div><div>11113</div></div>

Model Experimentation: Best ML Algorithm For Loan Eligibility Prediction

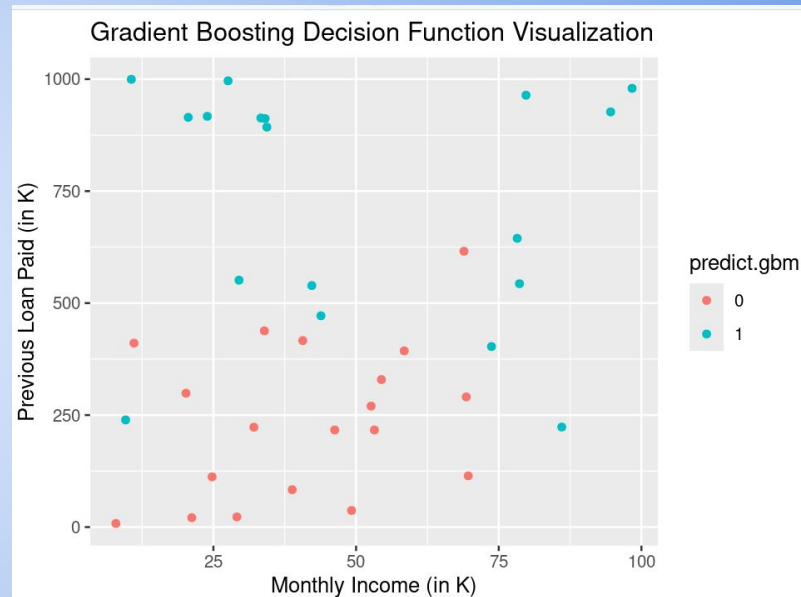
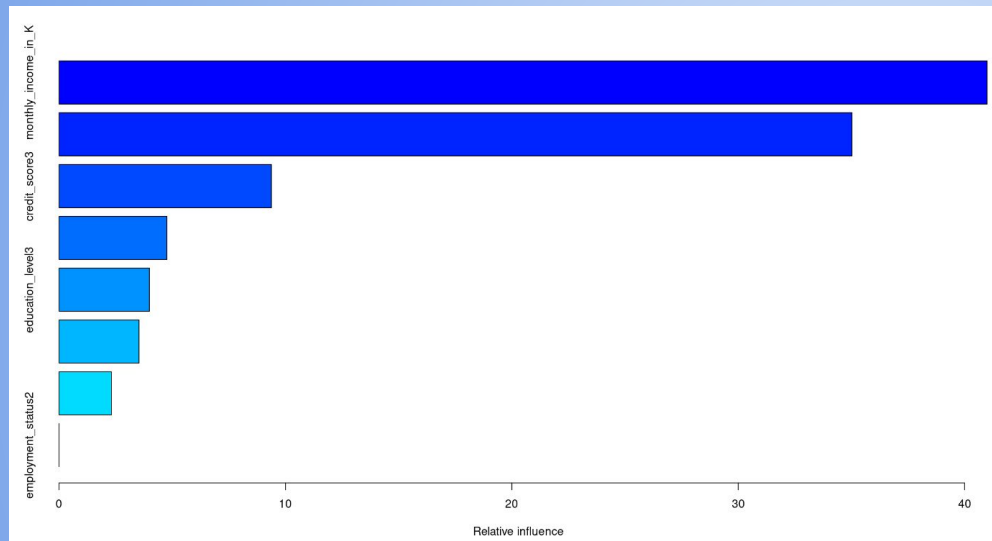
Algorithms	Accuracy	Sensitivity	Specificity	Confusion Matrix
Multivariate Logistic Regression	0.5676	0.6500	0.4706	<pre> 0 1 0 13 9 1 7 8 </pre>
Naive Baye's Algorithm	0.5405	0.6000	0.4706	<pre> 0 1 0 12 9 1 8 8 </pre>
Linear SVM	0.5676	0.6000	0.5294	<pre> 0 1 0 12 8 1 8 9 </pre>
Decision Tree	0.5405	1.0000	0.000	<pre> 0 1 0 20 17 1 0 0 </pre>
Bagging	0.3514	0.4000	0.2941	<pre> 0 1 0 8 12 1 12 5 </pre>
Random Forest	0.3784	0.4500	0.2941	<pre> 0 1 0 9 12 1 11 5 </pre>
Tuned Random Forest	0.4054	0.4000	0.4118	<pre> 0 1 0 8 10 1 12 7 </pre>
Gradient Boosting Algorithm	0.5946	0.6000	0.5882	<pre> 0 1 0 12 7 1 8 10 </pre>

```

var    rel.inf
monthly_income_in_K 40.994400
prev_loan_paid_in_K 35.024061
credit_score3       9.380856
credit_score2       4.763797
education_level3    3.989496
education_level2    3.532163
employment_status3  2.315227
employment_status2  0.000000

```

Visualizing Key Features Impacting Loan Approval



Conclusion

- **Bivariate analysis indicates that previous loan payment history is a significant predictor of loan eligibility.**
- **Multivariate model evaluation reveals Gradient Boosting outperforms other models with an accuracy of approximately 0.59.**
- **Monthly income and previous loan payments are pivotal factors in determining loan eligibility.**
- **A higher monthly income coupled with higher previous loan payments correlates with a greater likelihood of loan approval.**
- **Despite other features, previous loan payments alone can effectively model loan eligibility, yielding an accuracy of around 0.62.**

Any Questions?

Thank You