

(c) Summary of the logistic regression model built with 5 Lag and Volume input variables.

Call:

```
glm(formula = Direction ~ ., family = "binomial", data = train)
```

Deviance Residuals:

	Min	1Q	Median	3Q	Max
	-1.7186	-1.2498	0.9823	1.0841	1.4911

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	0.33258	0.09421	3.530	0.000415	***
Lag1	-0.06231	0.02935	-2.123	0.033762	*
Lag2	0.04468	0.02982	1.499	0.134002	
Lag3	-0.01546	0.02948	-0.524	0.599933	
Lag4	-0.03111	0.02924	-1.064	0.287241	
Lag5	-0.03775	0.02924	-1.291	0.196774	
Volume	-0.08972	0.05410	-1.658	0.097240	.

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1354.7 on 984 degrees of freedom

Residual deviance: 1342.3 on 978 degrees of freedom

AIC: 1356.3

Number of Fisher Scoring iterations: 4

Training data:

- Confusion matrix:

Confusion Matrix and Statistics

	Reference	
Prediction	Down	Up
Down	80	70
Up	361	474

Accuracy : 0.5624

95% CI : (0.5308, 0.5937)

No Information Rate : 0.5523

P-Value [Acc > NIR] : 0.2716

Kappa : 0.0562

Mcnemar's Test P-Value : <2e-16

Sensitivity : 0.8713

Specificity : 0.1814

Pos Pred Value : 0.5677

Neg Pred Value : 0.5333

Prevalence : 0.5523

Detection Rate : 0.4812

Detection Prevalence : 0.8477

Balanced Accuracy : 0.5264

'Positive' Class : Up

- Accuracy: 0.5624365

Test data:

- **Confusion Matrix:**

```
Confusion Matrix and Statistics

              Reference
Prediction Down Up
Down      31  44
Up       12  17

      Accuracy : 0.4615
      95% CI : (0.3633, 0.562)
No Information Rate : 0.5865
P-Value [Acc > NIR] : 0.9962

      Kappa : -3e-04

McNemar's Test P-Value : 3.435e-05

      Sensitivity : 0.2787
      Specificity : 0.7209
      Pos Pred Value : 0.5862
      Neg Pred Value : 0.4133
      Prevalence : 0.5865
      Detection Rate : 0.1635
      Detection Prevalence : 0.2788
      Balanced Accuracy : 0.4998

      'Positive' Class : Up
```

- Accuracy: 0.4615385

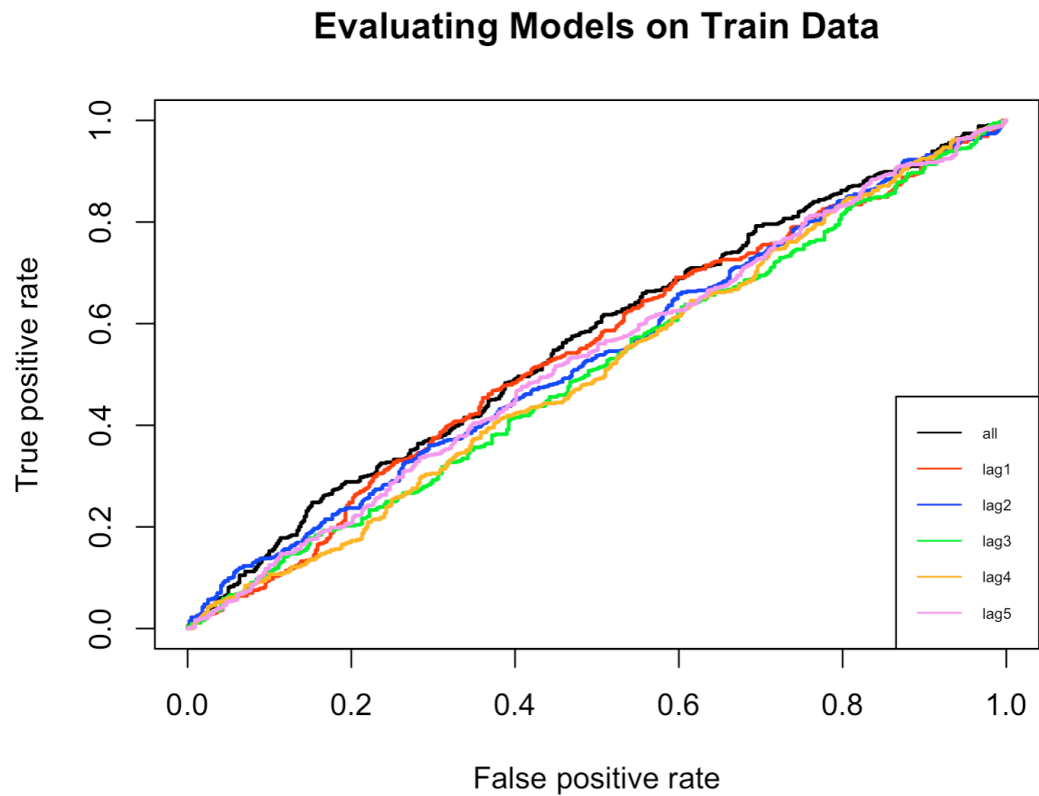
(d) Comparison table for all the evaluation metrics across the 6 models:

	TrainF1Score	TestF1Score	TrainAccuracy	TestAccuracy
Lag1	0.7031039	0.7096774	0.5532995	0.5673077
Lag2	0.7052490	0.7417219	0.5553299	0.6250000
Lag3	0.7115762	0.7393939	0.5522843	0.5865385
Lag4	0.7115762	0.7393939	0.5522843	0.5865385
lag5	0.7091267	0.7160494	0.5502538	0.5576923
all	0.6874547	0.3777778	0.5624365	0.4615385

The best model in terms of accuracy and F1 score is Lag2 (the model with only Lag2 as independent variable), based on the test data.

No, Lag2 model is not the best model on training data as it does not achieve the highest accuracy and F1 Score. On training data, the model with all variables achieves the highest accuracy and the model with Lag 3/ Lag 4 have the highest F1-Score on train data.

(e) Evaluating models on Train data:



Area under the curve for training data for each model:

```
> c(all_auc, lag1_auc, lag2_auc, lag3_auc, lag4_auc, lag5_auc)
[[1]]
[1] 0.5655179

[[2]]
[1] 0.5432506

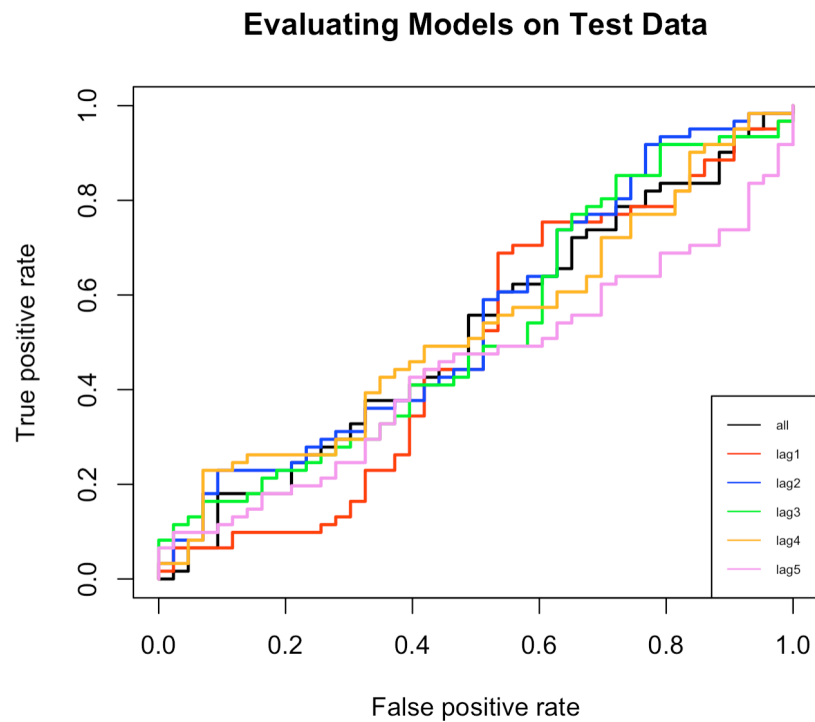
[[3]]
[1] 0.5349473

[[4]]
[1] 0.5067423

[[5]]
[1] 0.5084138

[[6]]
[1] 0.5302204
```

Evaluating models on Test data:



Area under the curve for test data for each model:

```
> c(all_auc, lag1_auc, lag2_auc, lag3_auc, lag4_auc, lag5_auc)
[[1]]
[1] 0.5177278

[[2]]
[1] 0.4864659

[[3]]
[1] 0.546321

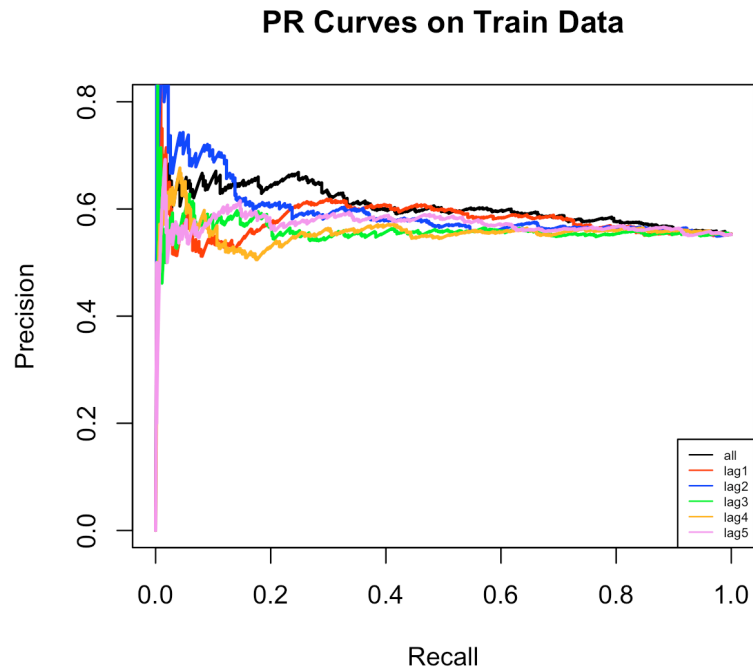
[[4]]
[1] 0.5242089

[[5]]
[1] 0.5257339

[[6]]
[1] 0.4422417
```

The best model on the basis of AUC is Lag2 on test data.
(Whereas on train data, best AUC is of the model with all the variables)

(f) Precision Recall curves on Train data:



Precision Recall curves on Test data:



The best model according to the PR curve is Lag2.

(g) Accuracy on the train data is maximum for the model with all variables due to more data capturing. But on the other hand, this model's accuracy is the worst on the test data. Overall, Lag2 model is the best based on a higher F1 Score on test data. Even though, it also has the highest accuracy, it does not matter for this case because we have imbalanced classes. On the same line, PR Curve's area under the curve (AUC) makes more sense than ROC's AUC in this case.