

Our results: Which customers will accept offers?

Below is the summary of our GLM predictive model.

What are the most influential variables that allow us to predict this response?

How well can we make predictions using this model?

A simple way to judge classification performance is the confusion matrix which breaks down the misclassification rate. Look at the confusion matrix in the validation sets result. The row is the true value of y, the column is the predicted value of y.

```
>summary(glm_model)
Model Details:
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H2OBinomialModel: glm
Model Key: GLM_model_R_1457656970445_5
GLM Model: summary
  family link      regularization number_of_predictors_total
number_of_active_predictors number_of_iterations training_frame
1 binomial logit Lasso (lambda = 1.127E-4
)                63                                57          101      RTM
P_5

H2OBinomialMetrics: glm
** Reported on training data. **

MSE: 0.07757775
R^2: 0.221004
LogLoss: 0.2747483
AUC: 0.7940001
Gini: 0.5880003
Null Deviance: 21611.94
Residual Deviance: 16915.7
AIC: 17031.7

Confusion Matrix for F1-optimal threshold:
      no  yes  Error  Rate
no    24816 2515 0.092020 =2515/27331
yes    1510 1943 0.437301 =1510/3453
Totals 26326 4458 0.130750 =4025/30784

Maximum Metrics: Maximum metrics at their respective thresholds
      metric threshold  value idx
1      max f1  0.206071 0.491215 211
2      max f2  0.149859 0.545716 243
3      max f0point5 0.413064 0.506541 118
4      max accuracy 0.521580 0.902157 85
5      max precision 0.873649 1.000000 0
6      max absolute_MCC 0.206071 0.422074 211
7 max min_per_class_accuracy 0.076867 0.722270 306

H2OBinomialMetrics: glm
```

** Reported on validation data. **

MSE: 0.08130666
 R²: 0.1955733
 LogLoss: 0.2845088
 AUC: 0.7889895
 Gini: 0.577979
 Null Deviance: 7386.88
 Residual Deviance: 5920.06
 AIC: 6036.06

Confusion Matrix for F1-optimal threshold:

	no	yes	Error	Rate
no	8400	817	0.088641	=817/9217
yes	549	638	0.462511	=549/1187
Totals	8949	1455	0.131296	=1366/10404

Maximum Metrics: Maximum metrics at their respective thresholds

	metric	threshold	value	idx
1	max f1	0.222369	0.482967	202
2	max f2	0.125944	0.535609	255
3	max f0point5	0.413343	0.479589	114
4	max accuracy	0.492893	0.897347	89
5	max precision	0.871053	1.000000	0
6	max absolute_MCC	0.222369	0.411435	202
7	max min_per_class_accuracy	0.077271	0.715249	302

Scoring History:

	timestamp	duration	iteration	log_likelihood	objective
1	2016-03-11 11:13:48	0.000 sec	0	10805.97080	0.35103
2	2016-03-11 11:13:49	0.205 sec	1	9114.07135	0.29647
3	2016-03-11 11:13:49	0.220 sec	2	8877.79091	0.28889
4	2016-03-11 11:13:49	0.237 sec	3	8714.62684	0.28365
5	2016-03-11 11:13:49	0.256 sec	4	8622.26183	0.28070

	timestamp	duration	iteration	log_likelihood	objective
16	2016-03-11 11:13:51	2.312 sec	64	8476.07325	0.27626
17	2016-03-11 11:13:51	2.562 sec	72	8470.66450	0.27610
18	2016-03-11 11:13:51	2.826 sec	80	8466.43715	0.27598
19	2016-03-11 11:13:51	3.044 sec	88	8463.21520	0.27589
20	2016-03-11 11:13:52	3.279 sec	96	8459.99344	0.27580
21	2016-03-11 11:13:52	3.560 sec	101	8457.85074	0.27546

Variable Importances: (Extract with `h2o.varimp`)

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Standardized Coefficient Magnitudes: standardized coefficient magnitudes

	names	coefficients	sign
1	month.may	0.628617	NEG
2	month.mar	0.615790	POS
3	emp.var.rate	0.579233	NEG
4	nr.employed	0.364999	NEG
5	cons.price.idx	0.319248	POS

	names	coefficients	sign
57	month.aug	0.000084	POS
58	job.entrepreneur	0.000000	POS
59	job.management	0.000000	POS
60	month.apr	0.000000	POS
61	month.oct	0.000000	POS
62	marital.married	0.000000	POS