Reading: GLM.Validation
Model: Source Text
Problem Type: Confusion matrix

Given An insurance company wants to make sure its litigation claims get assigned to a senior claims rep as soon as possible.

A logistic model was built to predict the likelihood of a claim going to litigation.

Claim	Claim went to	Predicted Probability
Number	Litigation	of going to Lit
1	Υ	96%
2	N	13%
3	Υ	37%
4	N	52%
5	N	96%
6	N	21%
7	Υ	50%
8	N	28%
9	N	79%
10	Υ	91%
11	N	17%
12	Υ	91%

**Find** Calculate confusion matrices for discrimination thresholds of 0.3 and 0.55.

#### Solution

Claim	Claim went	Predicted Probability	Discrimina	nt Threshold
Number	to Litigation	of going to Lit	0.30	0.55
1	Υ	0.96	TP	TP
2	N	0.13	TN	TN
3	Υ	0.37	TP	FN
4	N	0.52	FP	TN
5	N	0.96	FP	FP
6	N	0.21	TN	TN
7	Υ	0.50	TP	FN
8	N	0.28	TN	TN
9	N	0.79	FP	FP
10	Υ	0.91	TP	TP
11	N	0.17	TN	TN
12	Υ	0.91	TP	TP

Here, TP means True Positive, TN means True Negative, FP means False Positive, and FN means False Negative.

We assign these values as follows:

TP -> Claim went to Litigation = Y and Predicted Probability > Disciminant Threshold

FP -> Claim went to Litigation = N and Predicted Probability > Disciminant Threshold

FN -> Claim went to Litigation = Y and Predicted Probability < Disciminant Threshold

TN -> Claim went to Litigation = N and Predicted Probability < Disciminant Threshold

## **General Confusion Matrix**

#### Predicted

Actual

	Positive	negative
Positive	TP	FN
Negative	FP	TN

We assign the count of each type to the matrix.

Threshold = 0.3		
5	0	
3	4	

Threshold = 0.55		
3	2	
2	5	

Observe the lower threshold has less false negatives and more false positives than the higher threshold.

Whether this is a good or bad thing depends on the scenario.

If the price of a false positive is low in terms of money/time/resources and the cost of missing a true positive is high then this is good. If it is the reverse, i.e. little gain for a lot of cost then this is bad.

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Model: Source Text

Model: Source Text Confusion matrix

Given An insurance company wants to make sure its litigation claims get assigned to a senior claims rep as soon as possible.

A logistic model was built to predict the likelihood of a claim going to litigation.

Claim	Claim went to	Predicted Probability
Number	Litigation	of going to Lit
1	N	56%
2	Υ	38%
3	N	21%
4	N	48%
5	N	48%
6	Υ	73%
7	N	14%
8	N	91%
9	Υ	80%
10	N	1%
11	N	97%
12	N	97%

**Find** Calculate confusion matrices for discrimination thresholds of 0.26 and 0.89.

#### Solution

Claim	Claim went	Predicted Probability	Discrimina	int Threshold
Number	to Litigation	of going to Lit	0.26	0.89
1	N	0.56	FP	TN
2	Υ	0.38	TP	FN
3	N	0.21	TN	TN
4	N	0.48	FP	TN
5	N	0.48	FP	TN
6	Υ	0.73	TP	FN
7	N	0.14	TN	TN
8	N	0.91	FP	FP
9	Υ	0.80	TP	FN
10	N	0.01	TN	TN
11	N	0.97	FP	FP
12	N	0.97	FP	FP

Here, TP means True Positive, TN means True Negative, FP means False Positive, and FN means False Negative.

We assign these values as follows:

TP -> Claim went to Litigation = Y and Predicted Probability > Disciminant Threshold

FP -> Claim went to Litigation = N and Predicted Probability > Disciminant Threshold

FN -> Claim went to Litigation = Y and Predicted Probability < Disciminant Threshold

TN -> Claim went to Litigation = N and Predicted Probability < Disciminant Threshold

# General Confusion Matrix

### Predicted

	Positive	Negative
Positive	TP	FN
Negative	FP	TN

We assign the count of each type to the matrix.

Threshold = $0.26$		
3	0	
6	3	

Threshold = 0.89		
0	3	
3	6	