

# **IDS 702**

## **Logistic Regression Diagnostics for Predictions**

# Assessing predicted outcomes

Confusion matrix:

		Observed	
		Y=1	Y=0
Predicted	Y=1	TP (True Positives)	FP (False Positives)
	Y=0	FN (False Negatives)	TN (True Negatives)

# Confusion matrix

		Observed	
		Y=1	Y=0
Predicted	Y=1	TP (True Positives)	FP (False Positives)
	Y=0	FN (False Negatives)	TN (True Negatives)

- True positive rate (sensitivity) =  $TP / (TP + FN)$
- False positive rate (1-specificity) =  $FP / (FP + TN)$
- True negative rate (specificity) =  $TN / (FP + TN)$
- False negative rate =  $FN / (TP + FN)$

# Confusion matrix in R

```
> confusionMatrix(factor(ifelse(fitted(pumpmod)>0.5,"Ürgüp Sivrisi","Çerçvelik")),factor(pumpkin$Class),
positive="Ürgüp Sivrisi",mode="everything")
```

Confusion Matrix and Statistics

	Reference	
Prediction	Çerçvelik	Ürgüp Sivrisi
Çerçvelik	1163	196
Ürgüp Sivrisi	137	1004

Accuracy : 0.8668

95% CI : (0.8529, 0.8799)

No Information Rate : 0.52

P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.7327

Mcnemar's Test P-Value : 0.001481

Sensitivity : 0.8367

Specificity : 0.8946

Pos Pred Value : 0.8799

Neg Pred Value : 0.8558

Precision : 0.8799

Recall : 0.8367

F1 : 0.8578

Prevalence : 0.4800

Detection Rate : 0.4016

Detection Prevalence : 0.4564

Balanced Accuracy : 0.8656

'Positive' Class : Ürgüp Sivrisi

# ROC Curves

- Ideally, we'd like high values of both sensitivity and specificity (low values of 1-specificity)
- The receiver operating characteristic (ROC) curve plots:
  - Sensitivity on Y-axis
  - 1-specificity on X-axis
- Evaluated at lots of different probability cutoff values
- Area Under the Curve (AUC) calculated as area under the ROC curve (what value do we want?)

# ROC Curve in R

```
> library(pROC)
> roc(pumpkin$class.fac, fitted(pumpmod), print.thres=0.5, print.auc=T, plot=T, legacy.axes=T)
```

