Tutorial Business Analytics

Exercise 8 - Solution

Exercise 8.1

Create a 3-Fold Cross-validation for the specified instances

- Partition the data set into complementary subsets
- · Decide which subsets will be used for training and which for testing

Additionally, create a stratified 3-Fold Cross-validation for specified instances

| Instance | Class |
|----------|-------|
| 1 | + |
| 2 | + |
| 3 | + |
| 4 | + |
| 5 | - |
| 6 | + |
| 7 | + |
| 8 | - |
| 9 | - |
| 10 | _ |
| 11 | + |
| 12 | + |
| 13 | + |
| 14 | _ |
| 15 | - |

3-Fold Cross-validation:

(Note: This is only one of many possible solutions!)

$$P1 = \{1, ..., 5\}, P2 = \{6, ..., 10\}, P3 = \{11, ..., 15\}$$

Fold 1: Train: P2 & P3, Test: P1, classes: [4,1] **Fold 2:** Train: P1 & P3, Test: P2, classes: [2,3] **Fold 3:** Train: P1 & P2, Test: P3, classes: [3,2]

Classes (initial data set): [9,6]

Stratified 3-Fold Cross-validation:

We need to keep the distribution of + and – classes in balance for each set, i.e., three + classes and two - classes in each set.

$$P1 = \{1,2,3,5,8\}, P2 = \{4,6,7,9,10\}, P3 = \{11,12,13,14,15\}$$

Fold 1: Train: P2 & P3, Test: P1, classes: [3,2] Fold 2: Train: P1 & P3, Test: P2, classes: [3,2] Fold 3: Train: P1 & P2, Test: P3, classes: [3,2]

Classes (initial data set): [9,6]

Exercise 8.2

| True Class | Predicted Class |
|------------|------------------------|
| 0 | 0 |
| 0 | 1 |
| 1 | 1 |
| 1 | 0 |
| 0 | 0 |
| 1 | 0 |
| 0 | 0 |
| 1 | 1 |
| 0 | 1 |
| 1 | 0 |

Calculate Recall, False Alarm Rate, Specificity and Accuracy.

| True Class | Predicted Class |] |
|------------|-----------------|-----------|
| 0 | 0 |] TN |
| 0 | 1 | FP |
| 1 | 1 | TP |
| 1 | 0 | FN |
| 0 | 0 | TN |
| 1 | 0 | FN |
| 0 | 0 | TN |
| 1 | 1 | TP |
| 0 | 1 | FP |
| 1 | 0 | FN |

Recall (True Positive Rate, Sensitivity, Hit Rate)

"How many positive instances have been predicted to be positive"

$$tpr = \frac{tp}{tp + fn} = \frac{2}{2+3} = 0.4$$

False Alarm Rate (False Positive Rate)

"How many negative instances have been predicted to be positive"

$$fpr = \frac{fp}{fp + tn} = \frac{2}{2+3} = 0.4$$

Specificity (True Negative Rate)

"How many negative instances have been predicted to be negative"

$$tnr = \frac{tn}{fp + tn} = \frac{3}{2+3} = 0.6$$

Accuracy

"How many instances have been predicted correctly"

$$acc = \frac{tp + tn}{tp + fp + tn + fn} = \frac{2+3}{2+2+3+3} = 0.5$$

Exercise 8.3

The table below contains 3 classifier's **accuracy values**. Evaluate whether the results obtained by the new classifiers 1 and 2 are significantly different from the baseline classifier 0 (two-sided test, significance level 5%).

| Classifier 0 | Classifier 1 | Δ | Classifier 2 | Δ |
|--------------|--------------|-------|--------------|-------|
| 0.67 | 0.98 | -0.31 | 0.67 | 0.00 |
| 0.63 | 0.91 | -0.28 | 0.69 | -0.06 |
| 0.95 | 0.93 | 0.02 | 0.90 | 0.05 |
| 0.75 | 0.86 | -0.11 | 0.91 | -0.16 |
| 0.75 | 0.95 | -0.20 | 0.86 | -0.11 |
| 0.79 | 0.85 | -0.06 | 0.75 | 0.04 |
| 0.79 | 0.90 | -0.11 | 0.68 | 0.11 |
| 0.82 | 0.87 | -0.05 | 0.83 | -0.01 |
| 0.83 | 0.98 | -0.15 | 0.82 | 0.01 |
| 0.78 | 0.91 | -0.13 | 0.80 | -0.02 |

$$t = \frac{\overline{d}}{s_d / \sqrt{k}} \sim t_{k-1}$$

$$\overline{d} = \frac{1}{k} \sum_{i} d_{i}$$

$$s_d = \sqrt{\frac{1}{k-1} \sum_{i} \left(d_i - \overline{d} \right)^2}$$

$$H_0$$
: $\mu_0 - \mu_1 = 0$ vs H_1 : $\mu_0 - \mu_1 \neq 0$

$$\bar{d}_1 = -0.138$$

$$s_d = 0.102$$

$$T_1 = -4.267$$

$$t_{1-\frac{\alpha}{2}; n-1}^c = t_{0.975; 9}^c = 2.262$$

$$|T_1| > t_{1-\frac{\alpha}{2}; n-1}^c$$

⇒ Reject H₀. Classifier 1 is significantly different from Classifier 0

$$H_0$$
: $\mu_0 - \mu_2 = 0$ vs H_1 : $\mu_0 - \mu_1 \neq 0$ $\bar{d}_2 = -0.015$ $s_d = 0.079$ $T_2 = -0.602$ $|T_2| < t_{1-\frac{\alpha}{2}}^c; n-1$

⇒ Do not reject H_{0.} Classifier 2 is not significantly different from Classifier 0

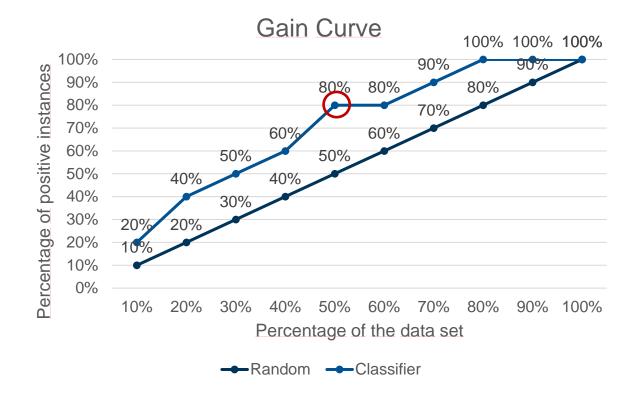
Exercise 8.4

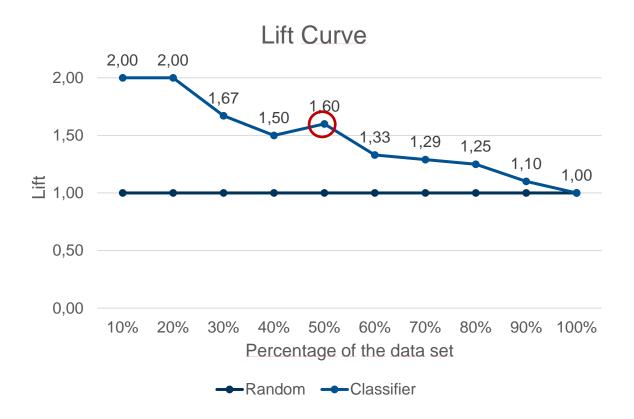
Use the given result of an evaluation (Cutoff = 0.87) to construct:

- a gain curve (10% steps)a lift curve
- an ROC curve

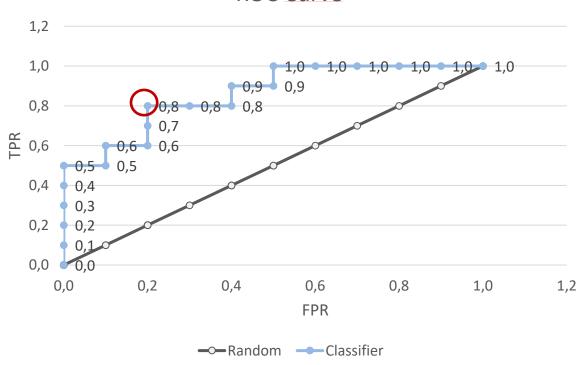
Remember: A Cutoff value of 0.87 means, we will classify an instance as positive until its probability falls below 0.87

| Number | Probability | Class |
|--------|-------------|-------|
| 1 | 0.991 | + |
| 2 | 0.977 | + |
| 3 | 0.973 | + |
| 4 | 0.945 | + |
| 5 | 0.918 | + |
| 6 | 0.915 | - |
| 7 | 0.906 | + |
| 8 | 0.889 | - |
| 9 | 0.873 | + |
| 10 | 0.871 | + |
| 11 | 0.869 | - |
| 12 | 0.866 | - |
| 13 | 0.862 | + |
| 14 | 0.852 | - |
| 15 | 0.837 | + |
| 16 | 0.831 | - |
| 17 | 0.829 | - |
| 18 | 0.811 | - |
| 19 | 0.787 | - |
| 20 | 0.779 | - |





ROC Curve



Annex

t-table

| df | α = 0.1 | $\alpha = 0.05$ | $\alpha = 0.025$ | α = 0.01 | $\alpha = 0.005$ |
|----|---------|-----------------|------------------|----------|------------------|
| 1 | 3.078 | 6.314 | 12.706 | 31.821 | 63.657 |
| 2 | 1.886 | 2.920 | 4.303 | 6.965 | 9.925 |
| 3 | 1.638 | 2.353 | 3.182 | 4.541 | 5.841 |
| 4 | 1.533 | 2.132 | 2.776 | 3.747 | 4.604 |
| 5 | 1.476 | 2.015 | 2.571 | 3.365 | 4.032 |
| 6 | 1.440 | 1.943 | 2.447 | 3.143 | 3.707 |
| 7 | 1.415 | 1.895 | 2.365 | 2.998 | 3.499 |
| 8 | 1.397 | 1.860 | 2.306 | 2.896 | 3.355 |
| 9 | 1.383 | 1.833 | 2.262 | 2.821 | 3.250 |
| 10 | 1.372 | 1.812 | 2.228 | 2.764 | 3.169 |
| 11 | 1.363 | 1.796 | 2.201 | 2.718 | 3.106 |
| 12 | 1.356 | 1.782 | 2.179 | 2.681 | 3.055 |
| 13 | 1.350 | 1.771 | 2.160 | 2.650 | 3.012 |
| 14 | 1.345 | 1.761 | 2.145 | 2.624 | 2.977 |
| 15 | 1.341 | 1.753 | 2.131 | 2.602 | 2.947 |
| 16 | 1.337 | 1.746 | 2.120 | 2.583 | 2.921 |
| 17 | 1.333 | 1.740 | 2.110 | 2.567 | 2.898 |
| 18 | 1.330 | 1.734 | 2.101 | 2.552 | 2.878 |
| 19 | 1.328 | 1.729 | 2.093 | 2.539 | 2.861 |
| 20 | 1.325 | 1.725 | 2.086 | 2.528 | 2.845 |
| 21 | 1.323 | 1.721 | 2.080 | 2.518 | 2.831 |
| 22 | 1.321 | 1.717 | 2.074 | 2.508 | 2.819 |
| 23 | 1.319 | 1.714 | 2.069 | 2.500 | 2.807 |