Image Analysis Project 8QA01

Part 4 – Evaluating your method

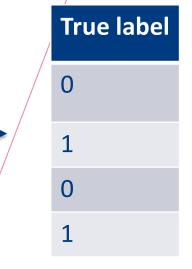
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Goal: Evaluate if your method is suitable for assessing skin lesions

- You developed some features and classifiers, now what?
- Evaluate features and/or features + classifier

| Q1: Age | ••• | Q7: Color | New feature |
|---------|-----|-----------|-------------|
| 20 | | 1 | 1.7 |
| 25 | | 1 | 2.5 |
| 40 | | 2 | 1.3 |
| 70 | | 3 | 0.1 |



| Pre | edicted label |
|-----|---------------|
| 0 | |
| 1 | |
| 0 | |
| 0 | |



Evaluating features

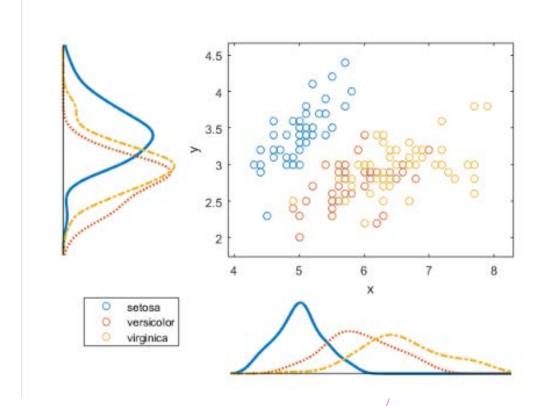
- Calculate features for artificial images (e.g. a circle) are the values logical?
- Compare features to measuring properties by hand
 - Correlation, inter-observer agreement, ...

| Asymmetry (algorithm) | Asymmetry (observer 1) |
|-----------------------|------------------------|
| 0 | 0 |
| 1 | 1 |
| 2 | 4 |
| 100 | 3 |



Evaluating features

Look at plots of your features per class – are there clusters?





- Compare predicted label to true label
- Accuracy = correctly classified images / total images * 100 (here/75%)

| True label | Predicted label |
|------------|-----------------|
| 0 | 0 |
| 1 | 1 |
| 0 | 0 |
| 1 | 0 |



- If classifier predicts a probability, convert to a label by thresholding
- The threshold is a hyperparameter
 - Use 0.5 or select a different one based on training/validation sets

| True label | Predicted probability | | Predicted label |
|------------|-----------------------|-------------|-----------------|
| 0 | 0.2 | > 0.5 ? | 0 |
| 1 | 0.9 | | 1 |
| 0 | 0.3 | | 0 |
| 1 | 0.4 | | 0 |



- Accuracy is not always a suitable metric
- If we have only 1% cancer in the data, predicting all images as healthy is 99% accurate!
- Consider other metrics, such as precision/recall, AUC (area under the receiveroperating curve) and others



- Our method has 75% performance on the test set
- Now what?

Table 1: Performance of our method

75%



Evaluating classifiers - wrongly classified examples

- Which images are incorrectly classified?
 - Confusion matrix

| Predicted → True ↓ | NC | С |
|-----------------------|----|----|
| Non-Cancer | 70 | 30 |
| Cancer | 5 | 95 |



Evaluating your method - wrongly classified examples

- Inspect the images visually
 - If your classifier outputs probabilities, look at "most incorrect" ones
- Are there any patterns?
 - Images too light/dark
 - Etc

| True label | P(class 1) |
|------------|------------|
| 0 | 0.9 |
| 0 | 0.6 |
| 1 | 0.1 |
| 1 | 0.4 |



Evaluating classifiers - different versions of your method

- What is the most important part of your method?
- Repeat your experiment, but remove 1 feature or parameter at a time
- Also referred to as "ablation experiments"

Evaluating classifiers – multiple runs

- Classifier A has 75% and classifier B has 76% is B better?
- Repeat the experiment 5 times, shuffling the data, to get mean + standard deviation
 - 75 \pm 0.1% and 76 \pm 0.1% vs
 - 75 \pm 5% and 76 \pm 5%
- You can shuffle the data, or use cross-validation



Evaluating algorithms

- We discuss evaluation from an algorithm's point of view,
- In practice, other factors should be considered
 - Do people trust this technology?
 - What are the rewards and risks to society?
 - Etc



Summary

- Goal of project
- Extracting features
- Training a classifier
- Evaluating your method
- Next: requirements for assignment