

Cost-sensitive Classification: Status and Beyond

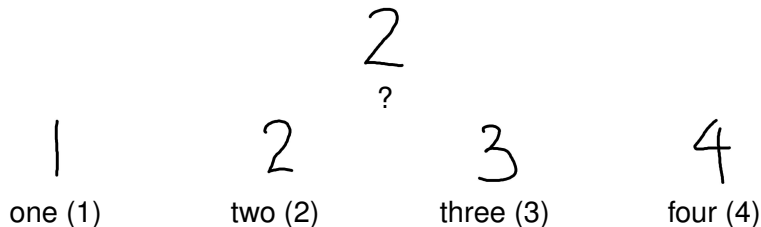
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Which Digit Did You Write?



- **classification**: a classic problem in machine learning

how to evaluate classification performance?



Mis-prediction Costs ($g(\mathbf{x}) \approx f(\mathbf{x})?$)

2
?

- ZIP code recognition:
1: **wrong**; 2: **right**; 3: **wrong**; 4: **wrong**
- check value recognition:
1: **one-dollar mistake**; 2: **no mistake**;
3: **one-dollar mistake**; 4: **two-dollar mistake**

different applications:
evaluate mis-predictions differently



Check Value Recognition

2
?

1: one-dollar mistake; 2: no mistake;

3: one-dollar mistake; 4: two-dollar mistake

- **cost-sensitive** classification problem:
different costs for different mis-predictions
- e.g. prediction error of g on some (\mathbf{x}, y) :

$$\text{absolute cost} = |y - g(\mathbf{x})|$$

cost-sensitive (as opposed to cost-less) classification:
relatively **new**, need more research



What is the Status of the Patient?



?



H1N1-infected



cold-infected



healthy

- another **classification** problem
—grouping “patients” into different “status”

are all mis-prediction costs equal?



Patient Status Prediction

error measure = society cost

| actual \ predicted | H1N1 | cold | healthy |
|--------------------|------|------|---------------|
| H1N1 | 0 | 1000 | 100000 |
| cold | 100 | 0 | 3000 |
| healthy | 100 | 30 | 0 |

- H1N1 mis-predicted as healthy: **very high cost**
- cold mis-predicted as healthy: **high cost**
- cold correctly predicted as cold: **no cost**

human doctors consider costs of decision;
can computer-aided diagnosis do the same?



Cost-sensitive Classification Setup

Given

N examples, each (input \mathbf{x}_n , label y_n) $\in \mathcal{X} \times \{1, 2, \dots, K\}$
and a K by K **cost matrix** \mathcal{C}

- $K = 2$: binary; $K > 2$: **multiclass**

Goal

a classifier $g(\mathbf{x})$ that pays a small cost $\mathcal{C}(y, g(\mathbf{x}))$ on future **unseen** example (\mathbf{x}, y)

cost-sensitive classification:
a **powerful** and general setup



A Quick Overview of Selected Algorithms

cost-sensitive classification via

- relabeling
- reweighting
- relabeling + reweighting (**our work, among others**)
- reducing to binary classification (**our work, among others**)
- reducing to regression (**our work**)



Cost-sensitive Classification via Relabeling

(Domingos, KDD, 1999)

key idea

cost-sensitive classification
= cost-less classification + relabeling some examples based on cost

- **general** and makes any cost-less approach cost-sensitive
- but **heuristic**: relabel using posterior probability estimate

theoretically sound approach?



Cost-sensitive Classification via Reweighting

(Elkan, IJCAI, 2001)

key idea

cost-sensitive classification
= cost-less classification + emphasizing some costly examples

- simple and **theoretically sound**
- but applies to **only binary** cost-sensitive classification
—multiclass case more complicated

theoretically sound approach
for **multiclass** cost-sensitive classification?



Cost-sensitive Classification via Relabeling + Reweighting (Abe et al., KDD, 2004; Lin, Caltech, 2008)

key idea

cost-sensitive classification
= cost-less classification + emphasizing and relabeling some examples

- **theoretically sound** for **multiclass**:
good cost-less classification \Rightarrow good cost-sensitive classification
- but introduces **relabeling noise** to the learning process
—bad practical performance

theoretically sound approach
for multiclass cost-sensitive classification
with **promising practical performance?**



Cost-sensitive Classification via Pairwise Binary Classification

(Beygelzimer et al, ICDM, 2003; Lin, NTU, 2010)

key idea

cost-sensitive classification
= binary classification + “Which of the two classes is of smaller cost?”

- **theoretically sound:**
good binary classification \Rightarrow good cost-sensitive classification
- **promising practical performance** (with a good binary classifier)
- does **not scale well with K** , the number of classes

theoretically sound approach
for **large- K** multiclass cost-sensitive classification
with promising practical performance?



Cost-sensitive Classification via Regression

(Tu and Lin, ICML, 2010)

key idea

cost-sensitive classification
= regression + “What is the estimated cost of each class?”

- **theoretically sound:**
good regression \Rightarrow good cost-sensitive classification
- **promising practical performance** (with a good regressor)
- **scales better with K**

what next?



Key Remaining Question: Application

theory: well-understood
algorithm: sufficiently many
application: **where? more?**

- **Where does cost come from?**
 - user-provided: but may not be feasible
 - consider cost intervals instead? (Liu and Zhou, KDD, 2010)
 - parameter-to-be-tuned: but currently lacks guidelines to users
 - link cost-sensitive to the true application needs?
- **What are important public benchmarks?**
 - semi-artificial (traditional): assigning arbitrary costs to existing sets
 - vision data with a class hierarchy?—ongoing but highly depends on feature extraction rather than costs
 - NELL data?—cost as soft-constraints
 - special types of learning problems (e.g. ranking)? others?

Assisting the users on **true application needs**
will drive future cost-sensitive classification research.



Thank you. Questions?

