### NON-DECOMPOSABLE PERFORMANCE MEASURES

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### Overview

- What we proposed?
- Existing Solutions
- Problems in Existing Solutions
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#### Problem Statement

Finding general optimization techniques which can work on all non-decomposable performance measures.

## Non-decomposable Performance Measures

- Performance Measures which do not decompose linearly.
- In classification problems, error rates are generally used.
- But not useful in case of class imbalance.
- So, measures like -
  - F-measure is used for text retrieval.
  - A continuous function of TPR and TNR which are used in class imbalanced classification settings.

# Problems in Non-decomposable Performance Measures

- Non-convex.
- Convex optimization techniques can not be applied on them.
- So, require an approximate upper bound on the raw performance measure curve given by some (surrogate) convex function, which can be optimized easily.

# SVM Approach to Optimize Non-Linear Performance Measures

#### Training Sample

$$S = ((x_1, y_1), \cdots, (x_n, y_n)) \in (\mathcal{X} \times \mathcal{Y})^n$$

drawn i.i.d. according to some unknown probability distribution.

• Find a rule  $h \in \mathcal{H}$  from hypothesis space  $\mathcal{H}$  such that it optimizes the expected prediction performance

$$R^{\Delta}(h) = \int \Delta((h(x_1), \cdots, h(x_n)), (y_1, \cdots, y_n)) dPr(S)$$

# SVM Approach to Optimize Non-Linear Performance Measures

- Loss function  $\Delta$  is not decomposable.
- ullet Optimize the empirical loss function  $\Delta$  over the entire sample set.

$$\hat{R}_{S}^{\Delta}(h) = \Delta((h(x_1), \cdots, h(x_n)), (y_1, \cdots, y_n))$$

• For such problems, Structured SVM is used.

$$\begin{split} \hat{R}_{S}^{\Delta}(h) &= \Delta((h(x_{1}), \cdots, h(x_{n})), (y_{1}, \cdots, y_{n})) \\ h(x, w) &= \underset{y \in \mathcal{Y}}{\text{arg max}} \, F(x, y; w) \\ F(x, y; w) &= \langle w, \psi(x, y) \rangle \\ h_{w}(\overline{x}, w) &= \underset{\overline{y} \in \mathcal{Y}}{\text{arg max}} (w^{T} \psi(\overline{x}, \overline{y})) \\ \min_{w, \zeta \geq 0} &= \frac{1}{2} \|w\|^{2} + C \zeta \\ s.t \quad \forall \overline{y}' \in \\ \overline{\mathcal{Y}} \setminus \overline{y} : w^{T} [\psi(\overline{x}, \overline{y}) - \psi(\overline{x}, \overline{y}')] \geq \Delta(\overline{y}', \overline{y}) - \zeta \end{split}$$

### References



Joachims, Thorsten

A Support Vector Method for Multivariate Performance Measures.

Proceedings of the 22Nd International Conference on Machine Learning



Narasimhan, Harikrishna and Vaish, Rohit and Agarwal, Shivani

On the Statistical Consistency of Plug-in Classifiers for Non-decomposable Performance Measures

Advances in Neural Information Processing Systems 27



Narasimhan, Harikrishna and Kar, Purushottam and Jain, Prateek

Optimizing Non-decomposable Performance Measures: A Tale of Two Classes. *ICML* 

# Questions?