

Information Retrieval (Preference Learning)

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Anno Accademico 2008/2009

Dip. di Matematica
Pura ed Applicata

F. Aiolli - Information Retrieval
2008/2009

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The structure of prediction

- **(Label) Ranking Tasks : $\text{FO}(\mathcal{Y})$**
 - Category Ranking
 - (Bipartite) Category Ranking
 - Q-label Classification
 - Single-label classification
- **(Instance) Ranking Tasks : $\text{FO}(\mathcal{X})$**
 - (Binary) Instance Ranking
- **Quantitative Tasks : $\{0, \dots, R-1\}^m$**
 - (Multivariate) Ordinal Regression
 - (Univariate) Ordinal Regression
 - Multi-label Classification
 - Binary Classification
 - One-class Classification

$\mathcal{X}$ = Users



$$f(x, y) \approx \hat{f}(x, y)$$

$$f: \mathcal{X} \times \mathcal{Y} \rightarrow \mathbb{R}$$

Types of predictions [Aiolli05]

□ Ordering Predictions

- Ordering of classes (or documents) on a relevance basis in such a way to be consistent with the supervision given as partial orders over the classes (or documents)
- Single-label classification, ranking

□ Rating Predictions

- Giving ranks from an ordinal scale to examples
- Binary classification, ordinal regression, and their multivariate extensions

Supervision

□ Supervision can be described as conjunctive sets of preferences of two types

■ Qualitative Preferences

- $(u(d_i, y_r), u(d_j, y_s))$

■ Quantitative Preferences ($\tau \in \mathcal{R}$)

- $(u(d, y), \tau)$
- $(\tau, u(d, y))$

Linear Preferences

- Now, consider linear expansion of the relevance function
 - $u(d,y) = w \cdot \phi(d,y)$
 - where $\phi(d,y) \in \mathcal{R}^d$ is a joint representation of document-class pairs and w a weight vector
- Qualitative preferences can be written as $w \cdot (\phi(d_i, y_r) - \phi(d_j, y_s)) > 0$
- Quantitative preferences $\delta(u(d,y) - \tau)$ can be written as $(w, \tau) \cdot (\delta \phi(d,y), -\delta) > 0$

Summarizing

- All the problem setting above can be seen as homogeneous linear problems in an opportune augmented space
- Any algorithm for linear optimization (e.g. perceptron, SVMs, or a linear programming package) can be used to solve them