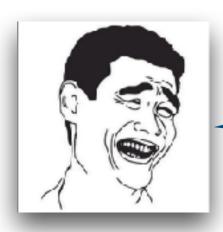
Learning to Rank Overview

Introduction of machine learning algorithms for ranking problem

Outline

- Introduction
- Ranking Methods
 - Supervised Methods
 - Numeric Methods
- Practical Issues



I want top php engineers



I want top php engineers









How to decide an order for these developers?





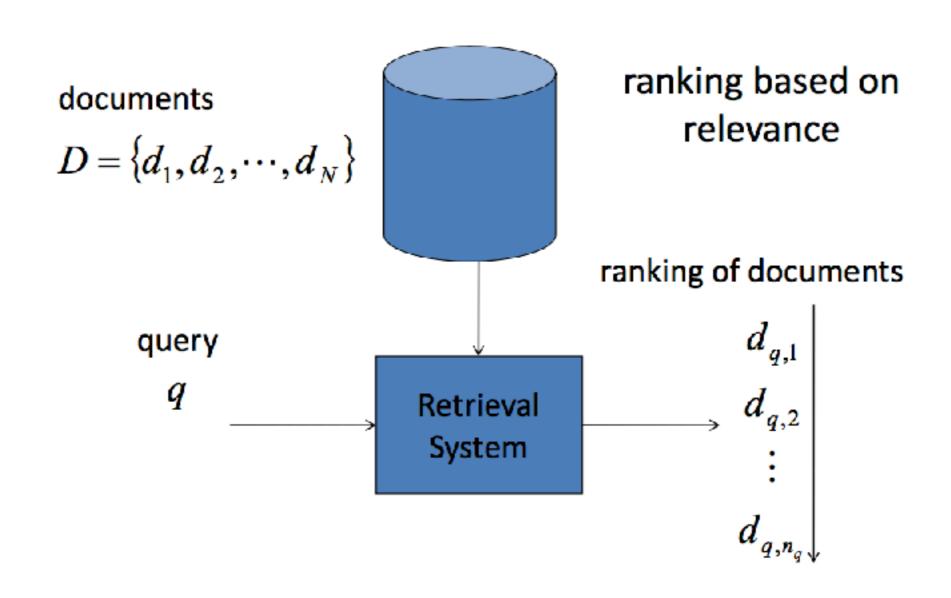


Ranking Everywhere

- Article ranking in community search page
- Article Infinite Scroll
 - Once the article opened, how to decide next 10 articles we recommend
- Applicant Quality Score
 - How to select top quality applicants to review first
- Codementor Request Matching

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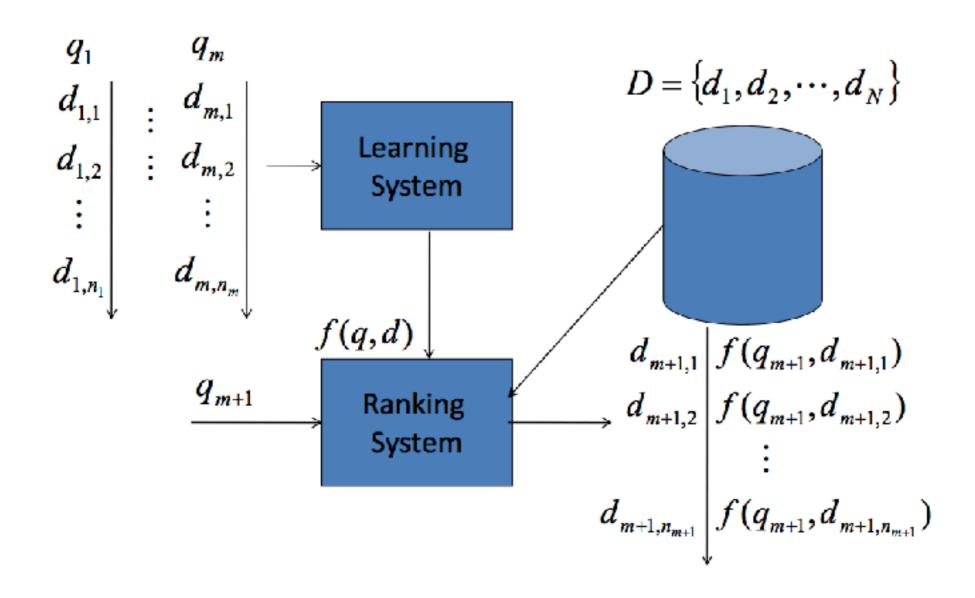
Native Ranking System



Native Ranking Methods

- The ranking expression is human experience
 - Need many time to tuning models
 - Hard to scale out if number of features increases
 - Overfitting problem
- Is it possible to use machine learning to solve ranking problem?

Learning to Rank System



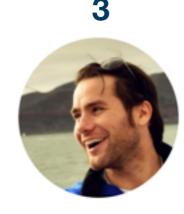
Supervised Methods

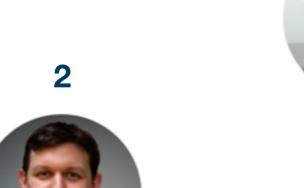
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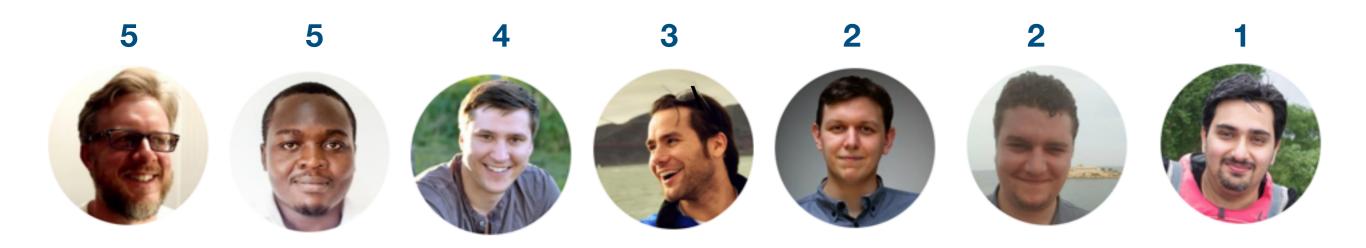








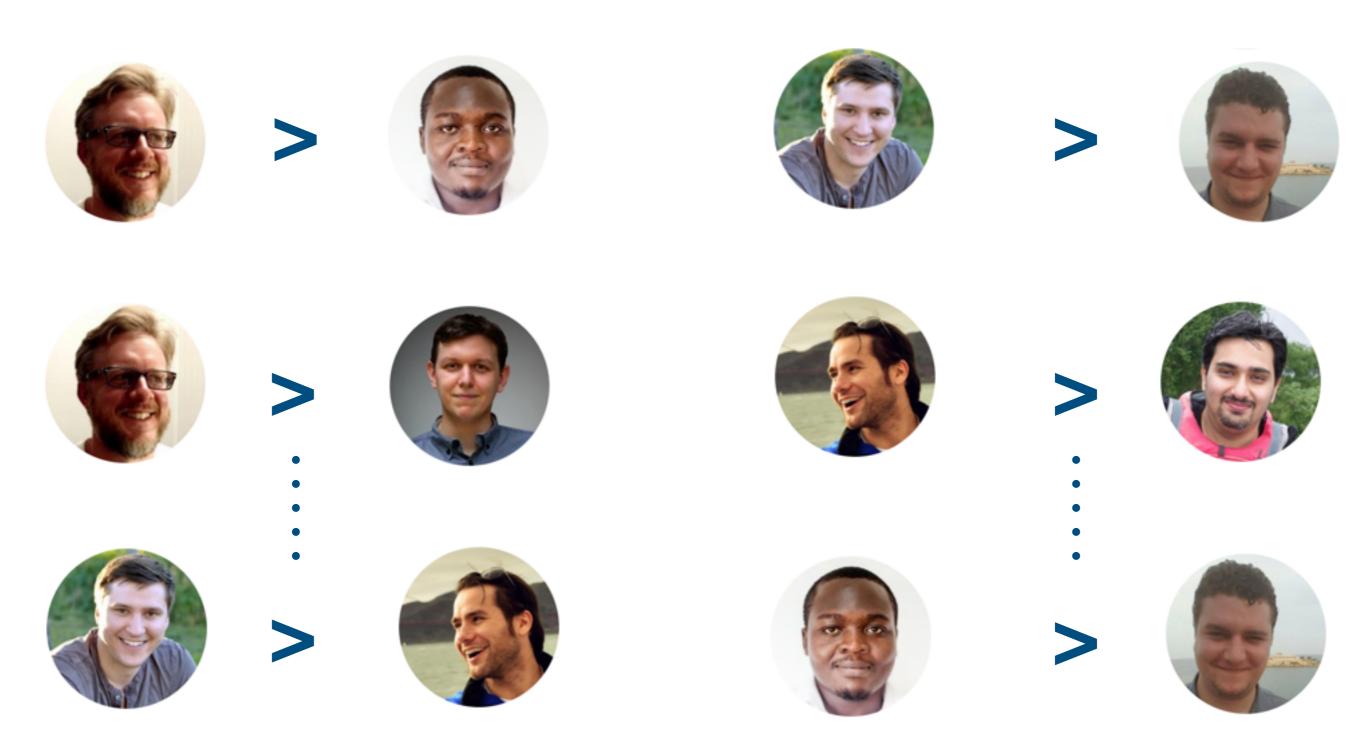


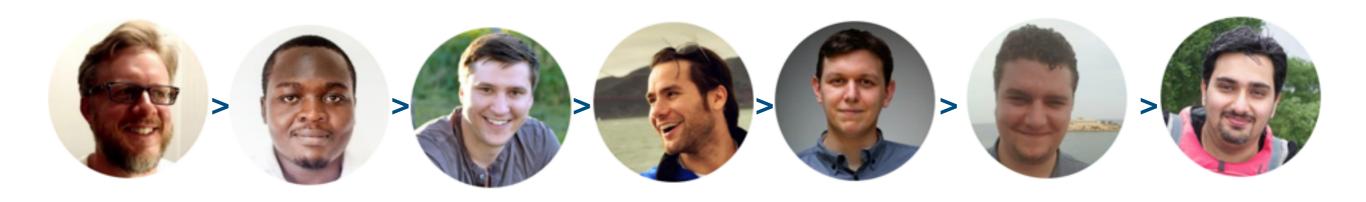


Top

- Model Complexity: O(|q|*n_q)
 - q: number of queries
 - n_q: number of matched documents of query q
- Algorithms
 - SVM, McRank, SLR ...

- Pros
 - Easy and intuitive to implement
 - Fast modeling
- Cons
 - Cannot consider ordering
 - Labeling is hard





Top

- Model Complexity: O(|q| * n^2)
- Algorithms
 - RankNet, IR-SVM, RankBoost

- Pros
 - Consider relative order
 - Most popular concept in real world
- Cons
 - Need to generate O(n^2) tags, high cost
 - Not consider the relevance between query and document

Numerical Methods

Top











































Top

















Which one is better ordering?

















Best Order

















Measure the difference with best orders















One of the orders

- Model Complexity: O(|q| * n!)
 - Many researches focus on reducing model complexity by numerical methods
- Algorithms
 - AdaRank, ListNet, SVM-MAP ...

- Pros
 - Consider all situations
 - Based on the origin problem definition
- Cons
 - O(n!) permutations (hard to implement and time bound)
 - Hard to optimize (objective functions not always continuous)

Practical Issues

- What is the objective metrics we pursue?
- How to generate labels?
- Cold-start problem

Objective Metrics

- Based on application
- Search
 - Most page views
 - Click Through Rate (CTR)
- CMX Job match
 - Contract Conversion Rate (CCR)
 - Interview Conversion Rate (ICR)

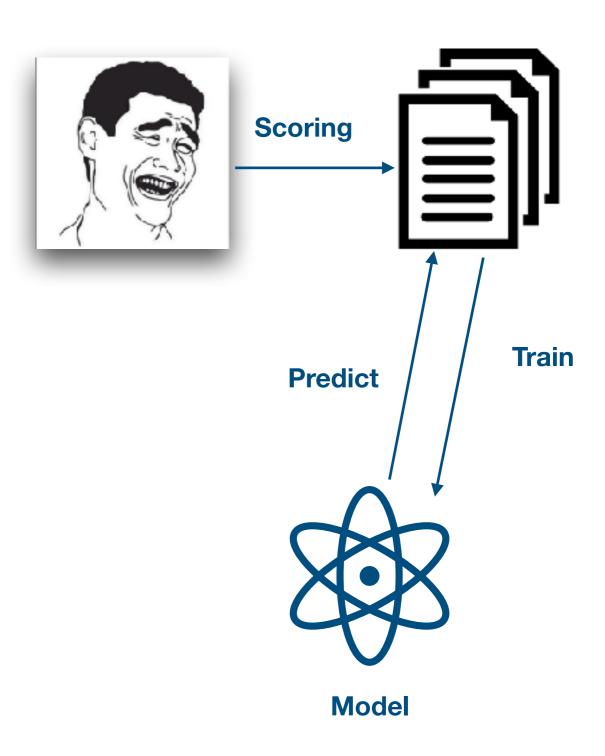
Labeling Problem

- How to label data if we have nothing
 - Small dataset human labeling is ok
 - Large dataset
 - Semi-supervised learning
 - Matrix factorization

Labeling Problem Semi-supervised

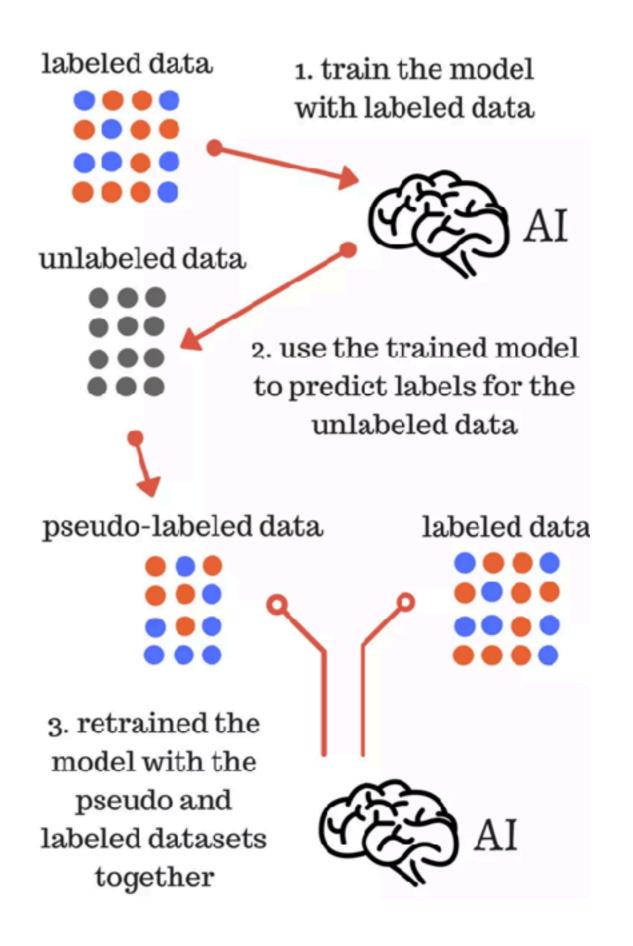
Iterative training with relevance Labels

- Rule-based models
- Human ranking expressions



Labeling Problem Semi-supervised

Pseudo Labeling



Labeling Problem - Matrix Factorization

When data size is large, we can label part of pairs

	P1	P2	Р3	P 4	P 5
Php	1	2	5	?	5
React	4	?	1	3	1
Javascript	3	5	?	1	?
Lavarel	?	1	4	?	5

Cold-start Problem

- Occurs in using some online metrics as features
 - Click count
 - View count
- How to handle new items?
 - Find similar items based on static features

Reference

Ranking

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