

# A Theoretical Analysis of NDCG Family Ranking Measures

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# Ranking: an example



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# Model of Ranking

- An instance space  $X$  ;
- A finite set of degree of relevancy  $Y$  ;
- A ranking (scoring) function  $f : X \rightarrow \mathfrak{R}$  ;
- Ranking a set  $S_n = \{(x_1, y_1), \dots, (x_n, y_n)\}$  with  $f$  :
  - Sorting  $x_1, \dots, x_n$  with  $f$ :

$$x_1, \dots, x_n \xrightarrow{f} x_{(1)}^f, \dots, x_{(n)}^f$$

$$f(x_{(1)}^f) \geq f(x_{(2)}^f) \dots \geq f(x_{(n)}^f)$$

$$y_1, \dots, y_n \longrightarrow y_{(1)}^f, \dots, y_{(n)}^f$$

# How to evaluate the ranking function?

- Normalized Discounted Cumulative Gain (**NDCG**): widely used in web search
  - Discounted Cumulative Gain (**DCG**)

Given a discount function  $D(r)$ ,  $r = 1, 2, \dots$ , the DCG of a ranking function  $f$  on a data set  $S_n = \{(x_1, y_1), \dots, (x_n, y_n)\}$  is defined as:

$$DCG_D(f, S_n) = \sum_{r=1}^n y_{(r)}^f D(r)$$

- **NDCG**:

$$NDCG_D(f, S_n) = \frac{DCG_D(f, S_n)}{IDCG_D(f, S_n)}, \quad IDCG_D(f, S_n) = \max_{f'} DCG_D(f', S_n)$$

- Commonly used discounts: logarithmic discount (**Standard NDCG**)

$$D(r) = \frac{1}{\log(1+r)}$$

# Some natural questions

- Are the NDCG family good evaluation measures?
- Why do we use the logarithmic discount? Any theoretical justifications?
- Is there any **criterion** to determine whether a ranking evaluation measure is good?

# A first observation (motivation)

- The standard NDCG (log discount) always converges to the **same** limit.
- **Theorem.** Let  $D(r) = \frac{1}{\log(1+r)}$ . Assume  $S_n$  consists of i.i.d. data. Then for **every** ranking function  $f$

$$NDCG_D(f, S_n) \xrightarrow{n \rightarrow \infty} 1, \quad \text{a.s.}$$

Standard NDCG is difficult to differentiate ranking functions on large data sets?

# Proposed criterion: Consistent Distinguishability

## ➤ Consistent Distinguishability (intuition):

For every pair of substantially different ranking functions, the ranking measure should be able to **consistently** distinguish the two functions on **almost all** data sets.

## ➤ Definition:

Let  $(x_1, y_1), (x_2, y_2), \dots$  be i.i.d. drawn from an underlying distribution. Let  $S_n = \{(x_1, y_1), \dots, (x_n, y_n)\}$ . A pair of ranking functions  $f_0, f_1$  is said to be **consistently distinguishable** by a ranking measure  $M$ , if there exists a negligible function  $neg(N)$  and  $b \in \{0,1\}$  so that for every sufficiently large  $N$ , with probability  $1 - neg(N)$

$$M(f_b, S_n) > M(f_{1-b}, S_n)$$

holds for all  $n > N$  simultaneously.

# Main results (informal)

## ➤ **Standard NDCG** (log discount):

Under natural conditions, every pair of substantially different ranking functions is consistently distinguishable by standard NDCG (although the measure converges to the same limits).

## ➤ **Characterization of feasible discounts of NDCG:**

- For polynomial discount  $D(r) = r^{-\beta}$ ,  $\beta \in (0,1)$ , under weak conditions every pair of substantially different ranking functions is consistently distinguishable by NDCG with this discount.
- For polynomial discount  $D(r) = r^{-\beta}$ ,  $\beta > 1$ , NDCG does not have consistent distinguishability. Moreover,  $NDCG_D(f, S_n)$  does not converge in general.



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Thanks!