

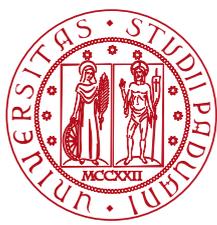
Visual Interactive Failure Analysis: Supporting Users in Information Retrieval Evaluation

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Marco Angelini, Giuseppe Santucci

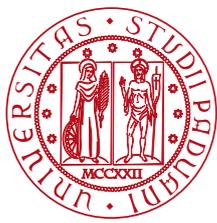
“La Sapienza” University of Rome, Italy



Outline



- Motivations
- Models for Interaction
- The Prototype and Application Examples
- The Domino Effect
- Final Remarks and On-Going Works



Motivation

Reduce the time needed to understand and analyze the behavior of an IR system providing a **visual interactive tool for evaluation**

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Failure Analysis



or “Understand What is Wrong”.

Is it better to re-rank or to re-query?

Motivation

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Failure Analysis

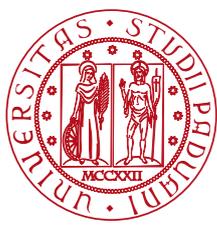
or “Understand What is Wrong”.

Is it better to re-rank or to re-query?



What-if Analysis

“Try to anticipate the effect of a modification of the system”.



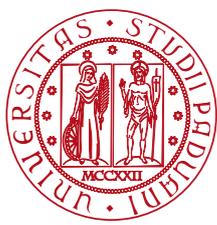
The Underlying Idea

IR Evaluation

+

Visual Analytics

Visual Analytics is not only a mean to improve the presentation of results, but also a mean allowing the users to analyze and interact with data



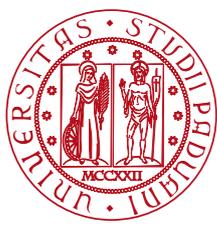
Models for Interaction



Rank Gain/Loss Model

Clustering via Learning to Rank

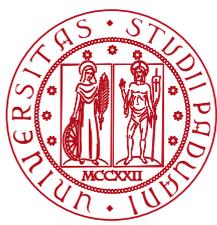
What-If Analysis Model



Models for Interaction

Rank Gain/Loss Model

DCG is good if we want to compare performances between different systems but **what if we want to look at misplaced documents?**



Models for Interaction

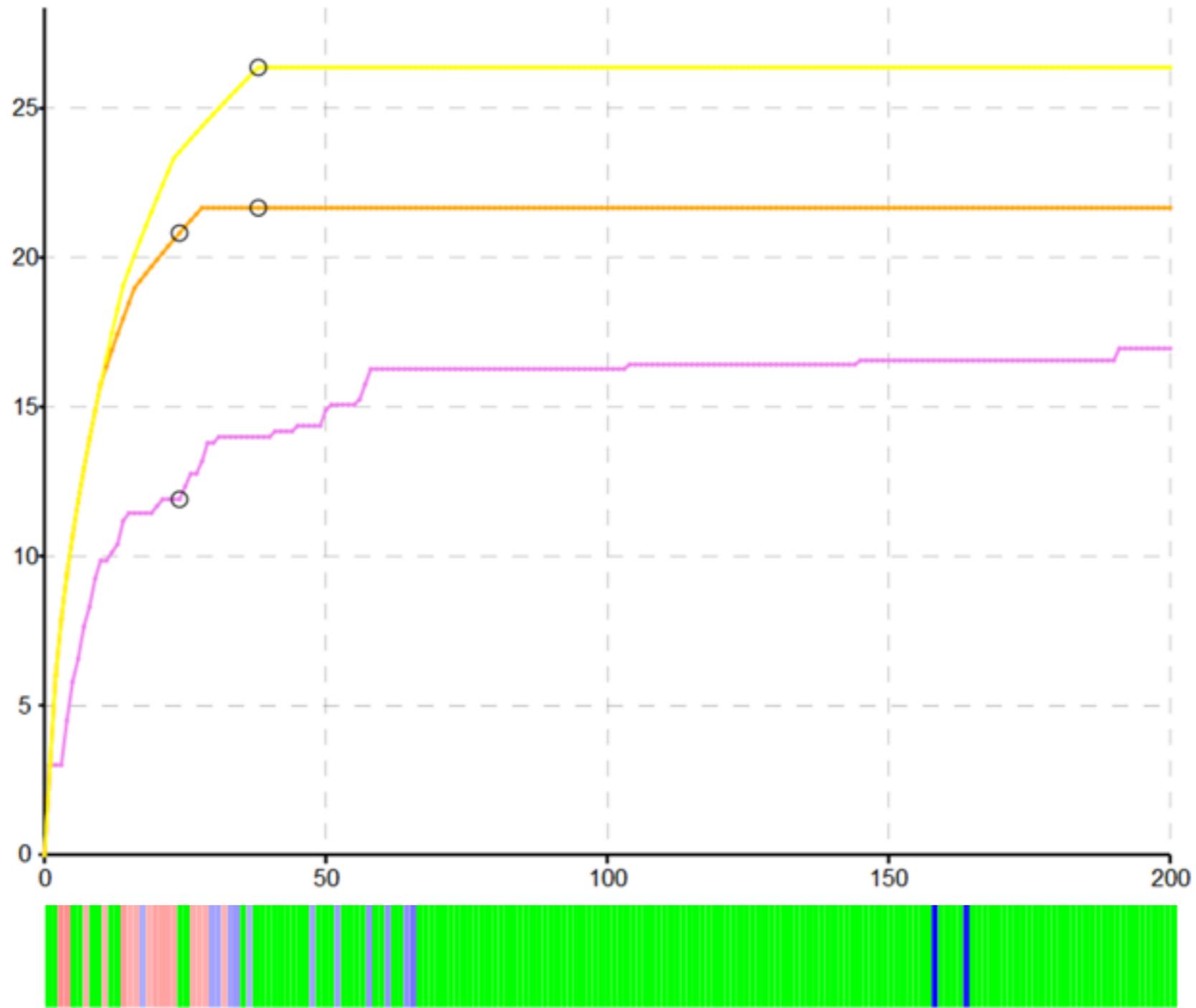
Rank Gain/Loss Model

DCG is good if we want to compare performances between different systems but **what if we want to look at misplaced documents?**

Relative Position = RPos let us determine how much a document is misplaced with respect to its ideal rank.

Models for Interaction

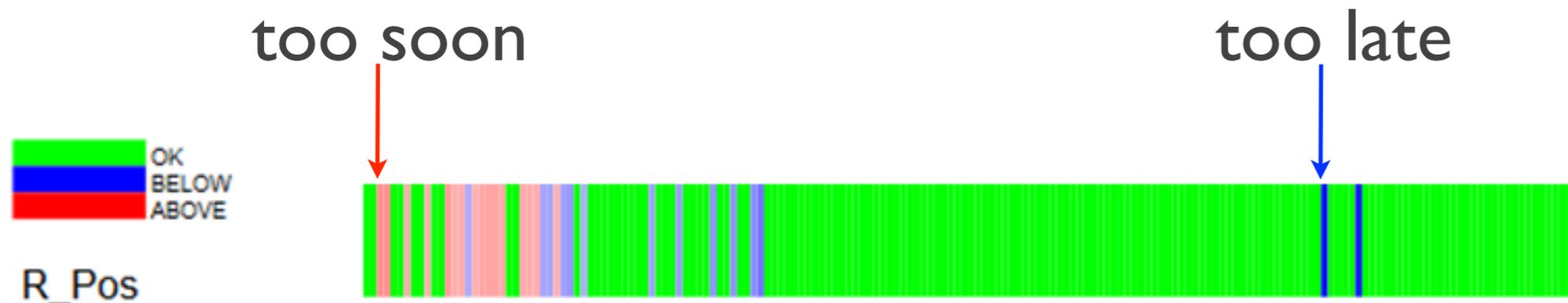
Rank Gain/Loss Model: R_Pos



Rank Gain/Loss Model: R_Pos

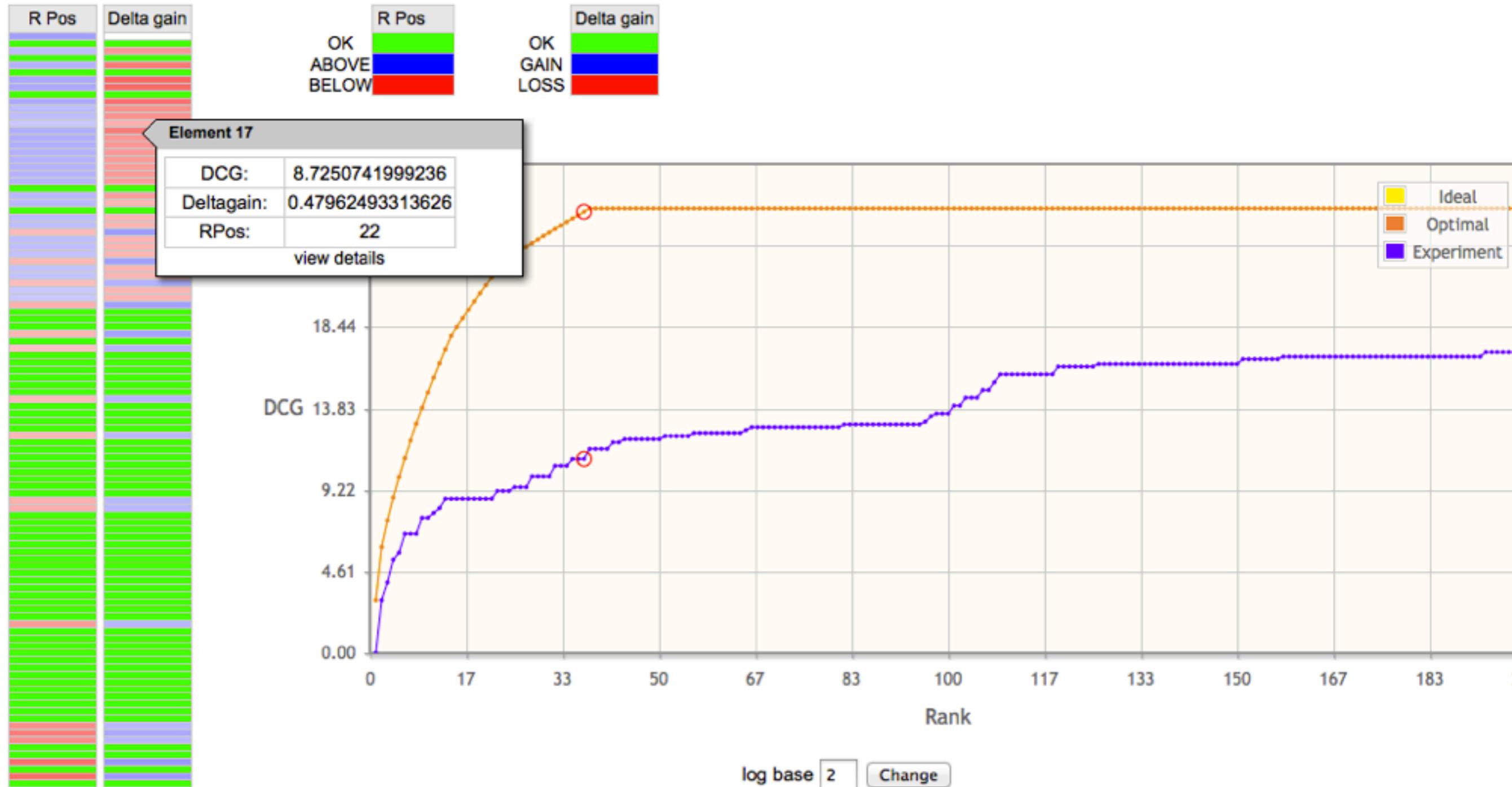
By means of R_Pos we can say if a document is ranked too soon or too late. But, **can we quantify “too soon” and “too late”?**

How much these misplaced documents impact on the **dcg?**



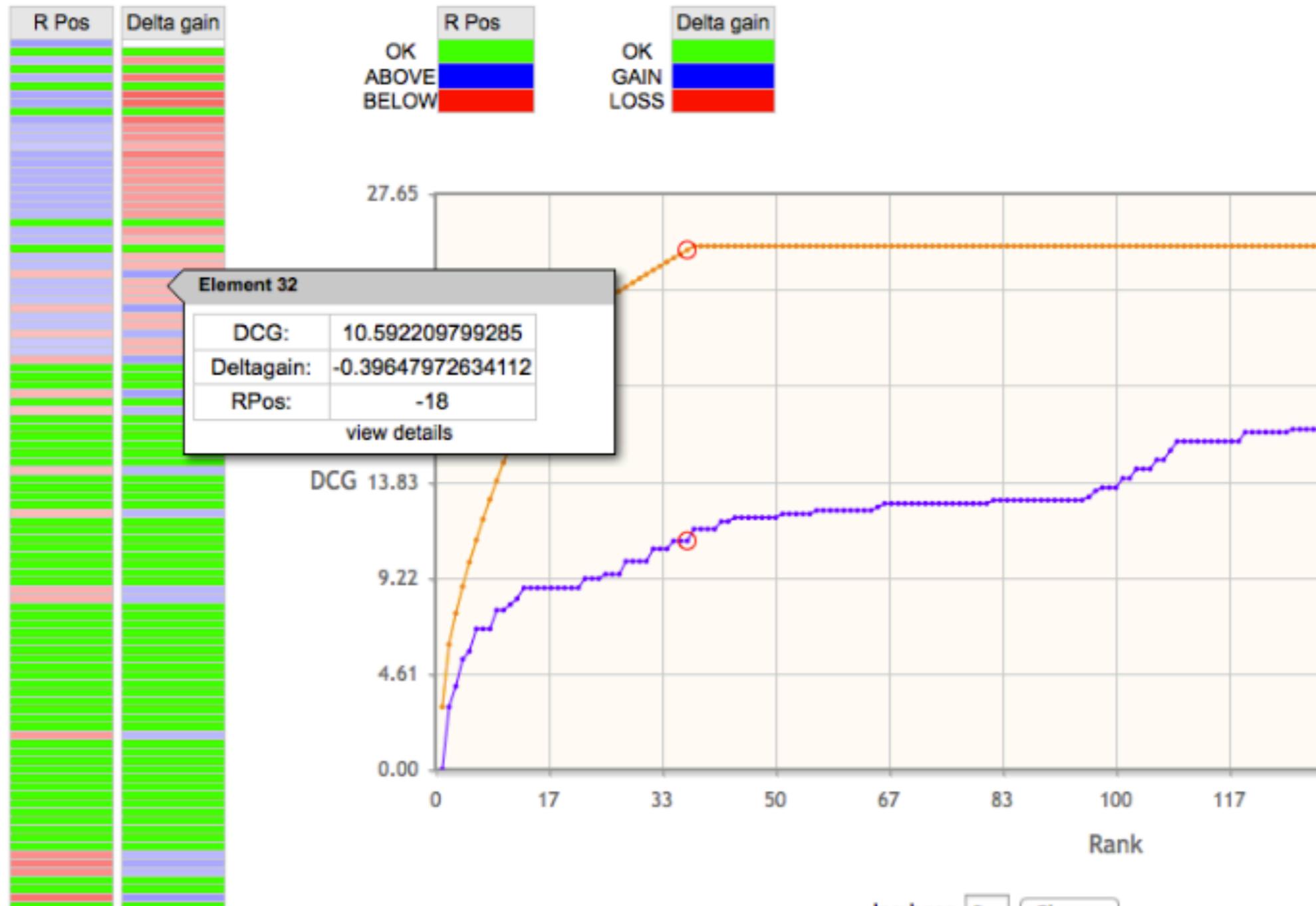
Models for Interaction

Rank Gain/Loss Model : Delta Gain

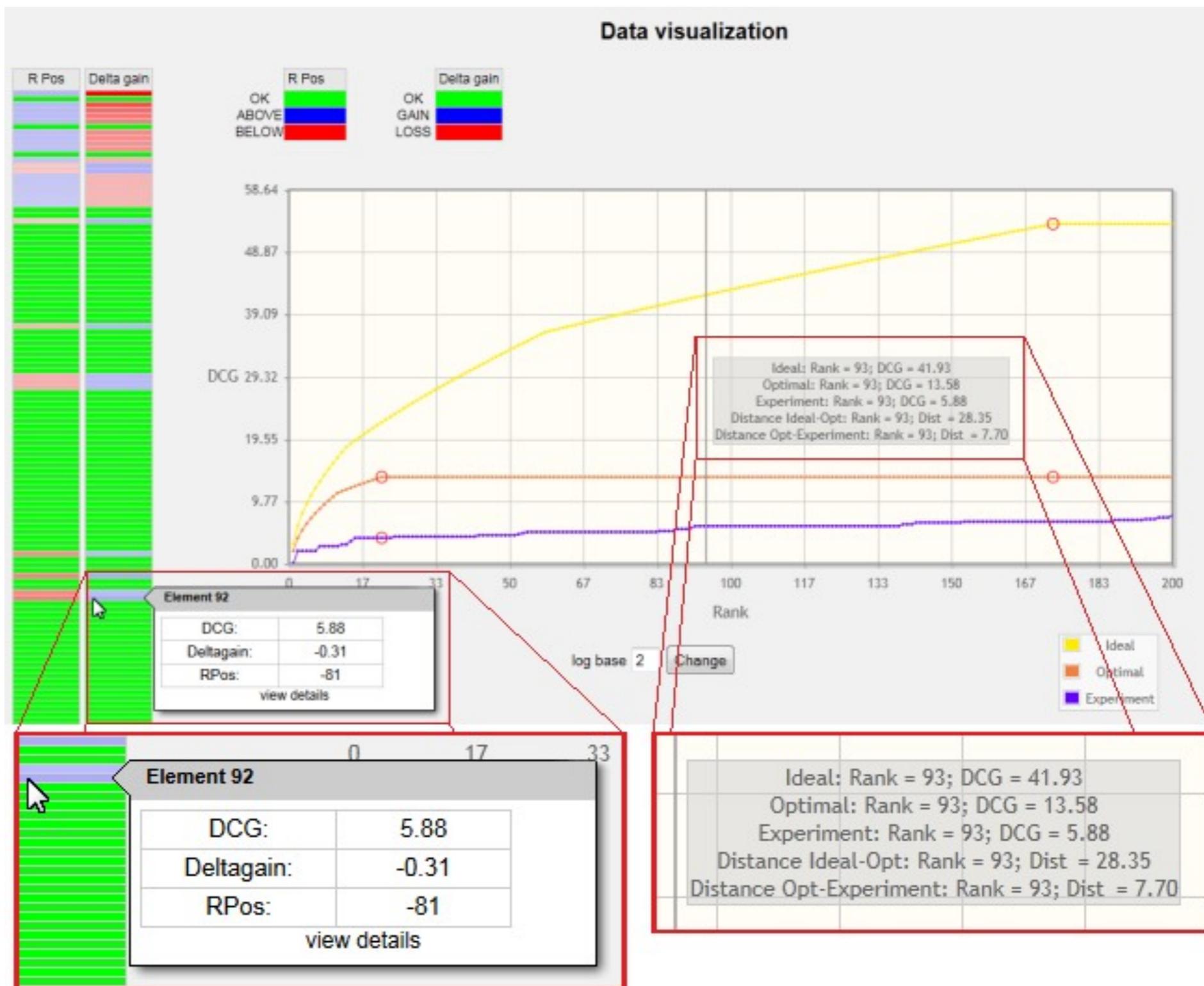


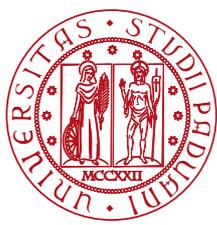
Models for Interaction

Rank Gain/Loss Model : Delta Gain



Failure Analysis



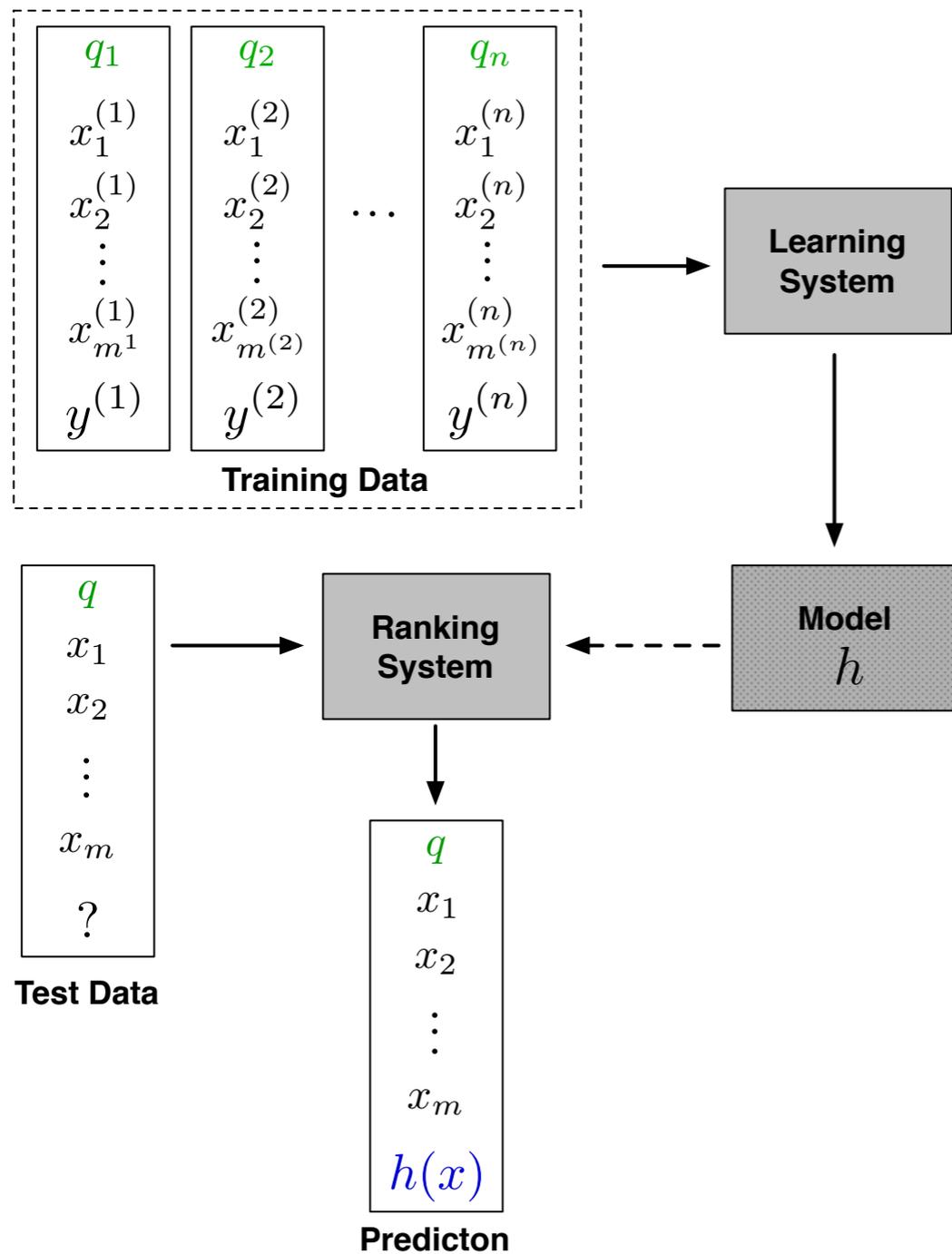


Models for Interaction

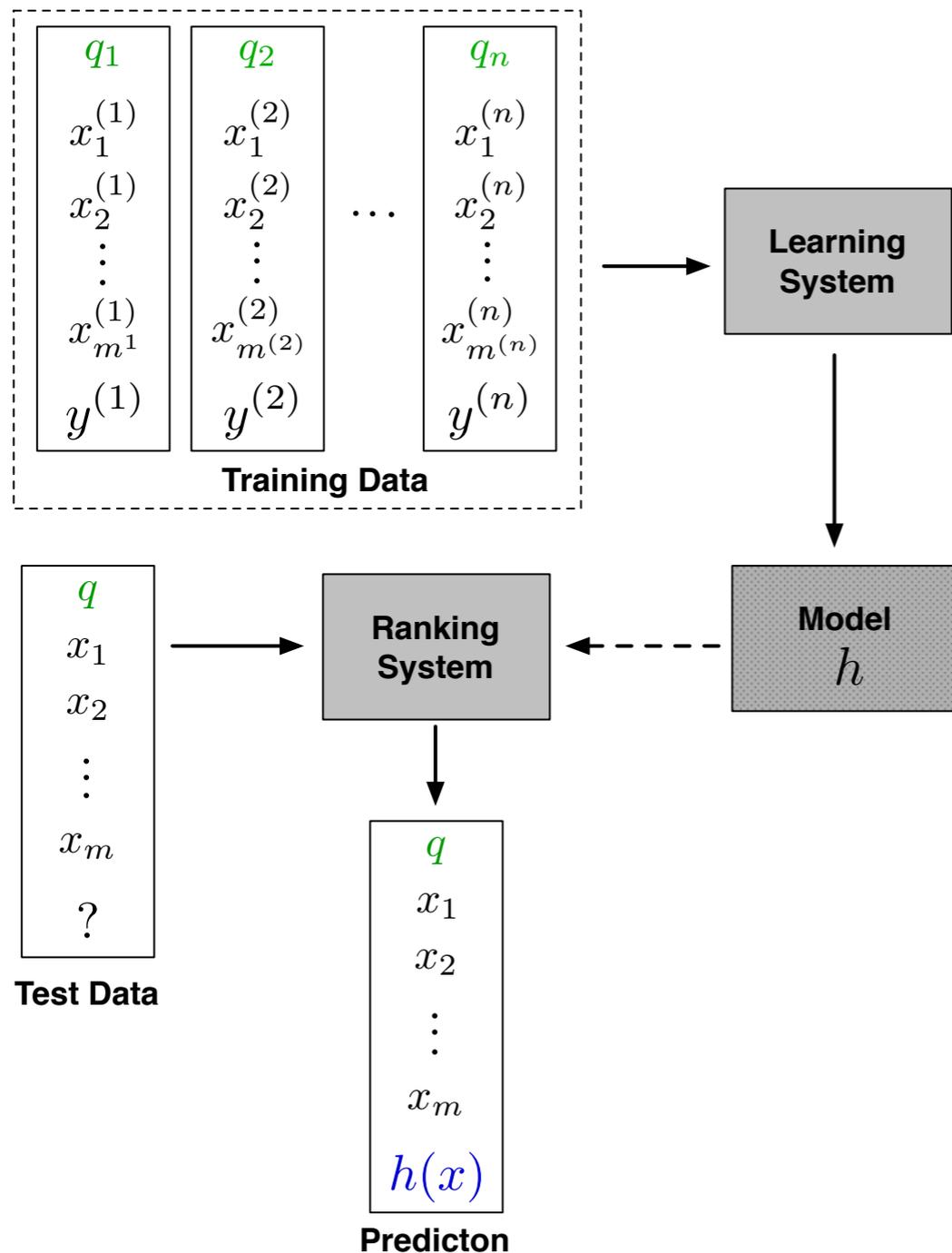
Clustering via Learning to Rank

Learn the ranking model of the IR system under investigation in order to simulate the way in which it ranks the documents

Clustering via Learning to Rank



Clustering via Learning to Rank



- Leverage on the clustering hypothesis
- Group together the documents which are similar from the considered ranking model point-of-view

Generating the clusters

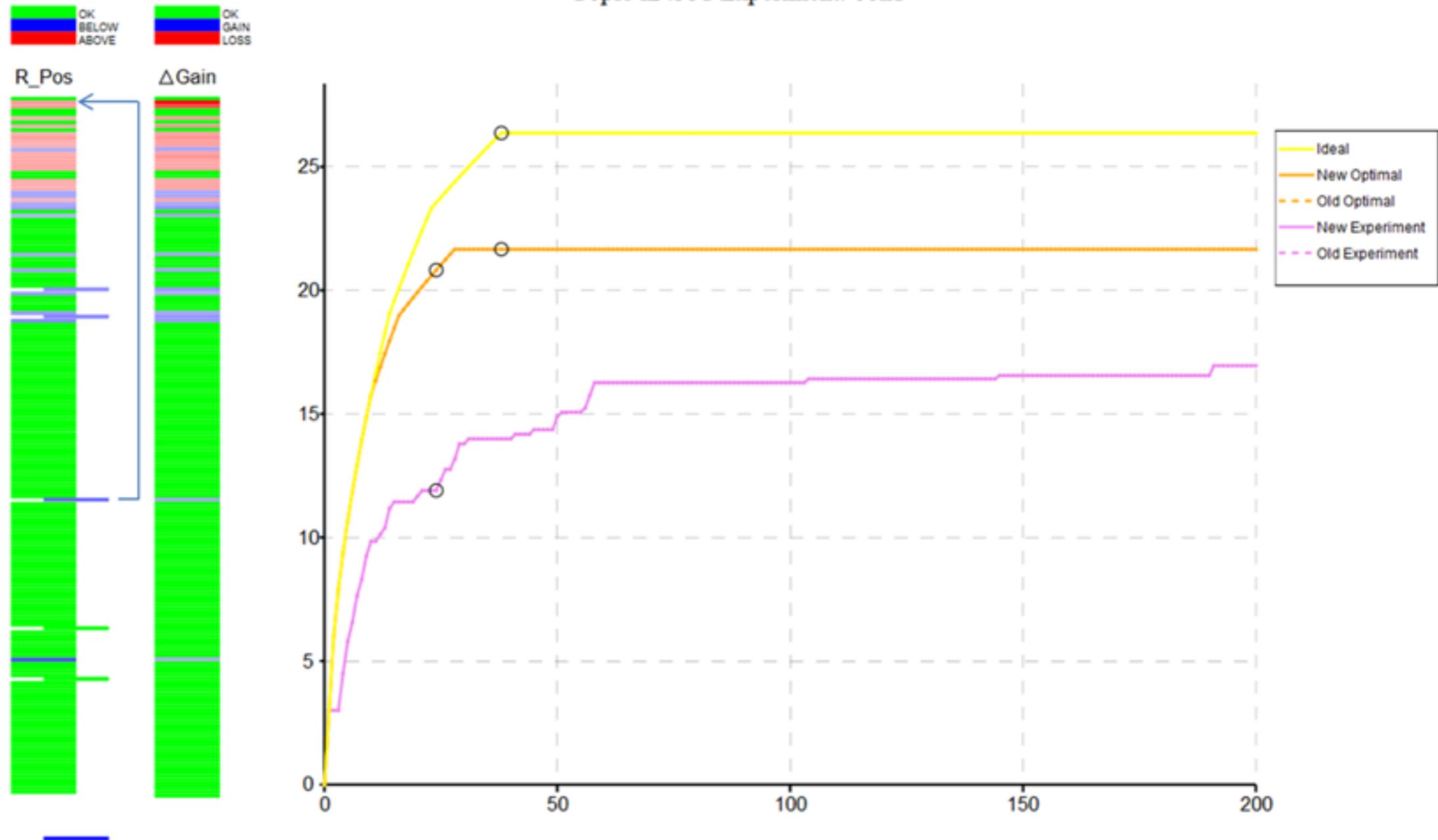
- (1) Submit each doc in D_j as a query and retrieve a set of docs D_i ;
- (2) determine $C_i = D_j \setminus D_i$;
- (3) ranking the documents in C_i by employing the learned ranking model

Models for Interaction

What-if Analysis

Visual comparison of Ranked Result Cumulated Gains

Topic ID:351 Experiment: bbn1

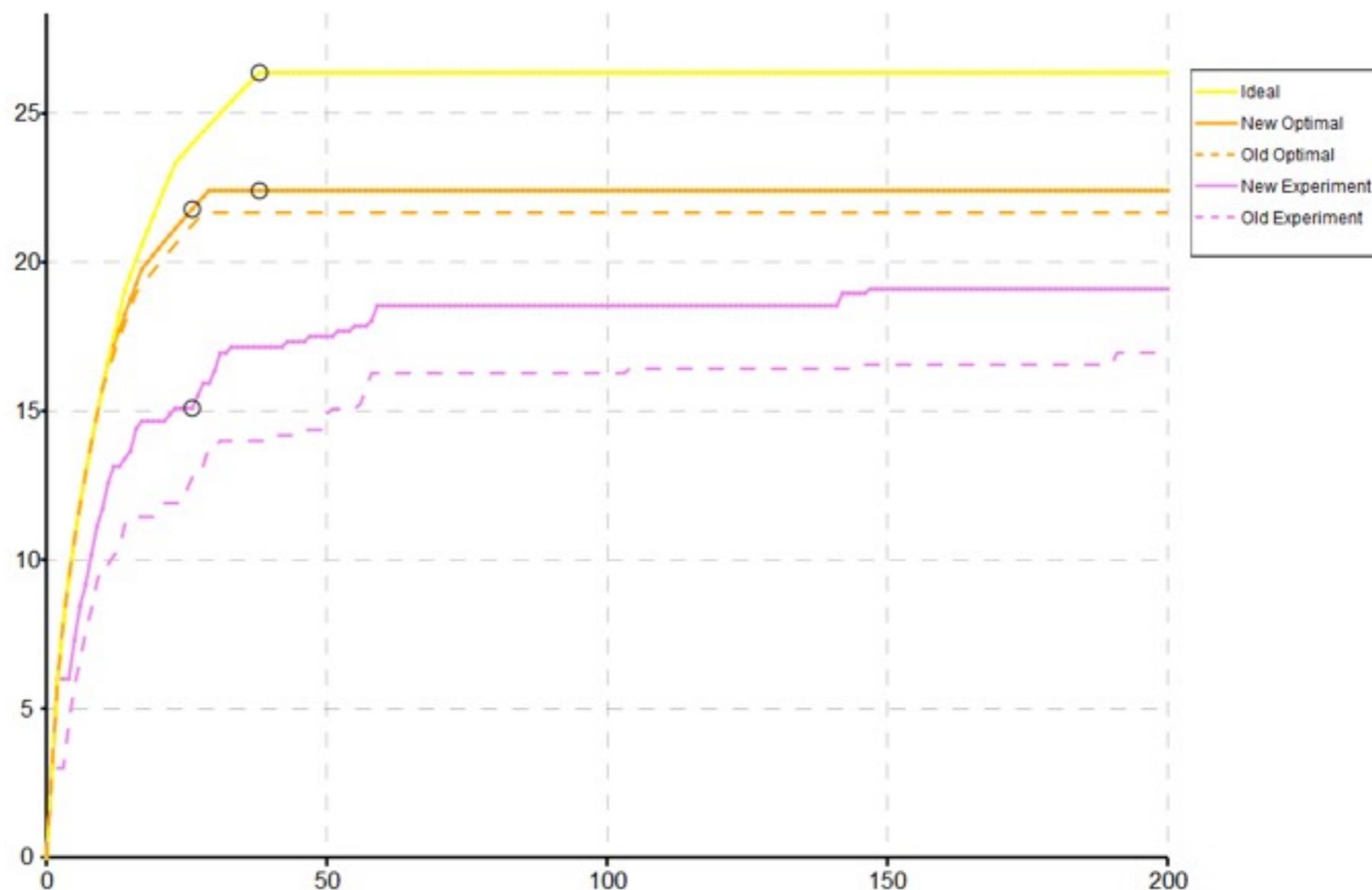
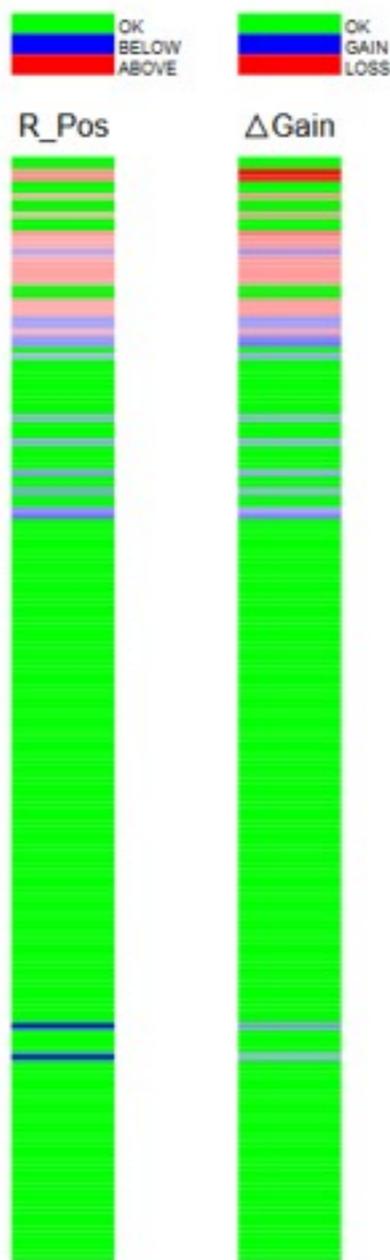


Models for Interaction

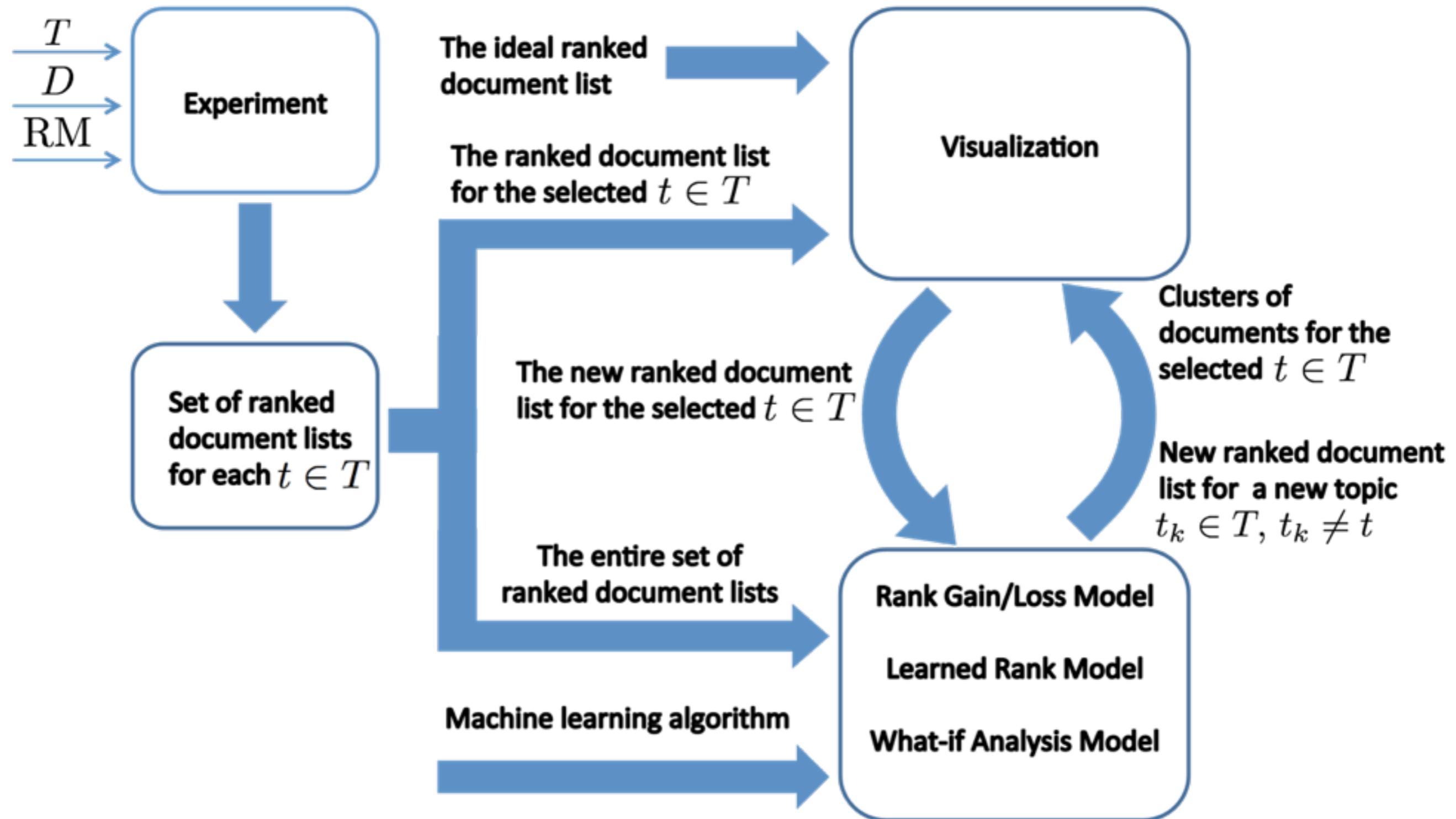
What-if Analysis

Visual comparison of Ranked Result Cumulated Gains

Topic ID:351 Experiment: bbn1

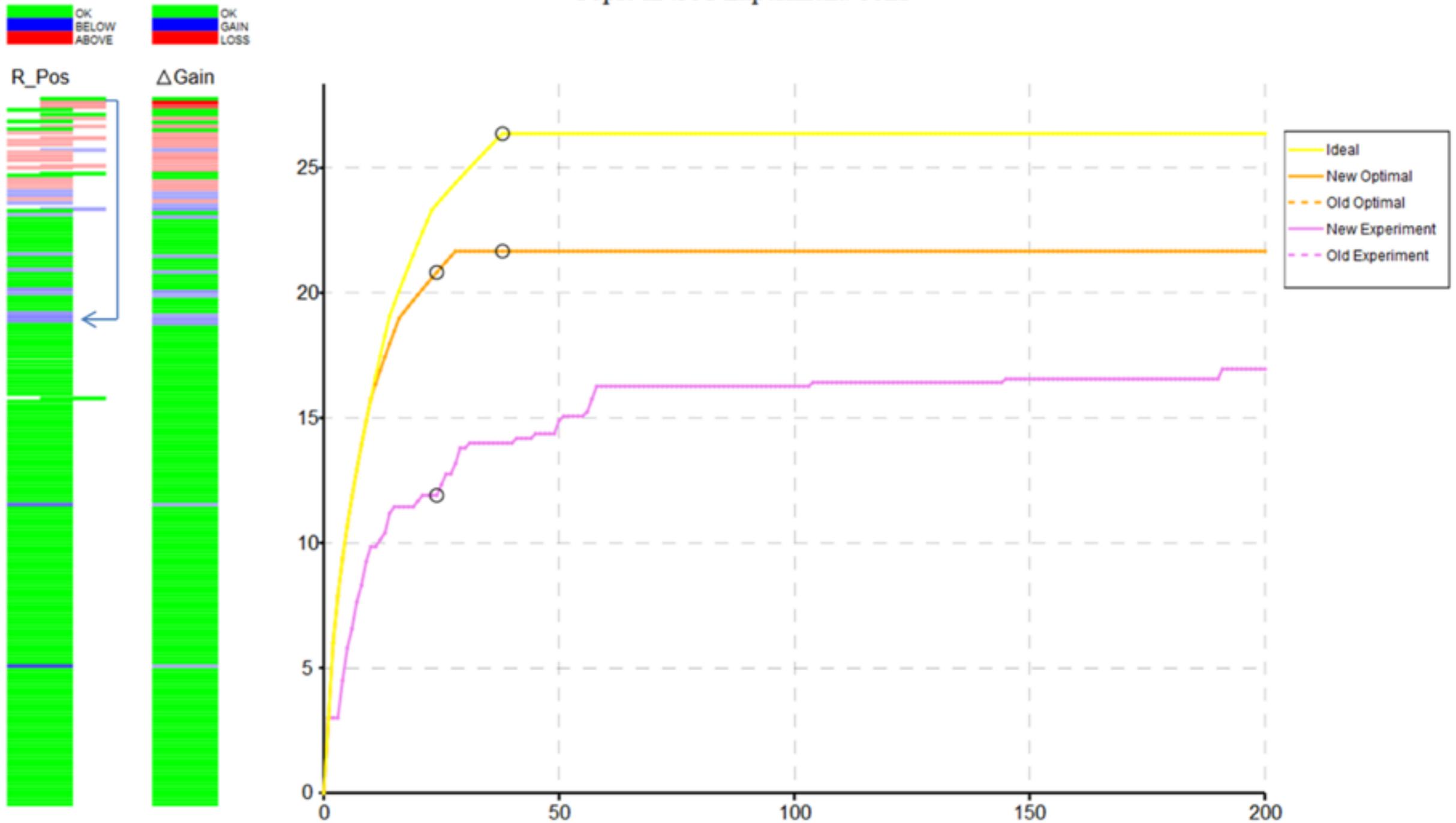


To Summarize: Data Pipeline



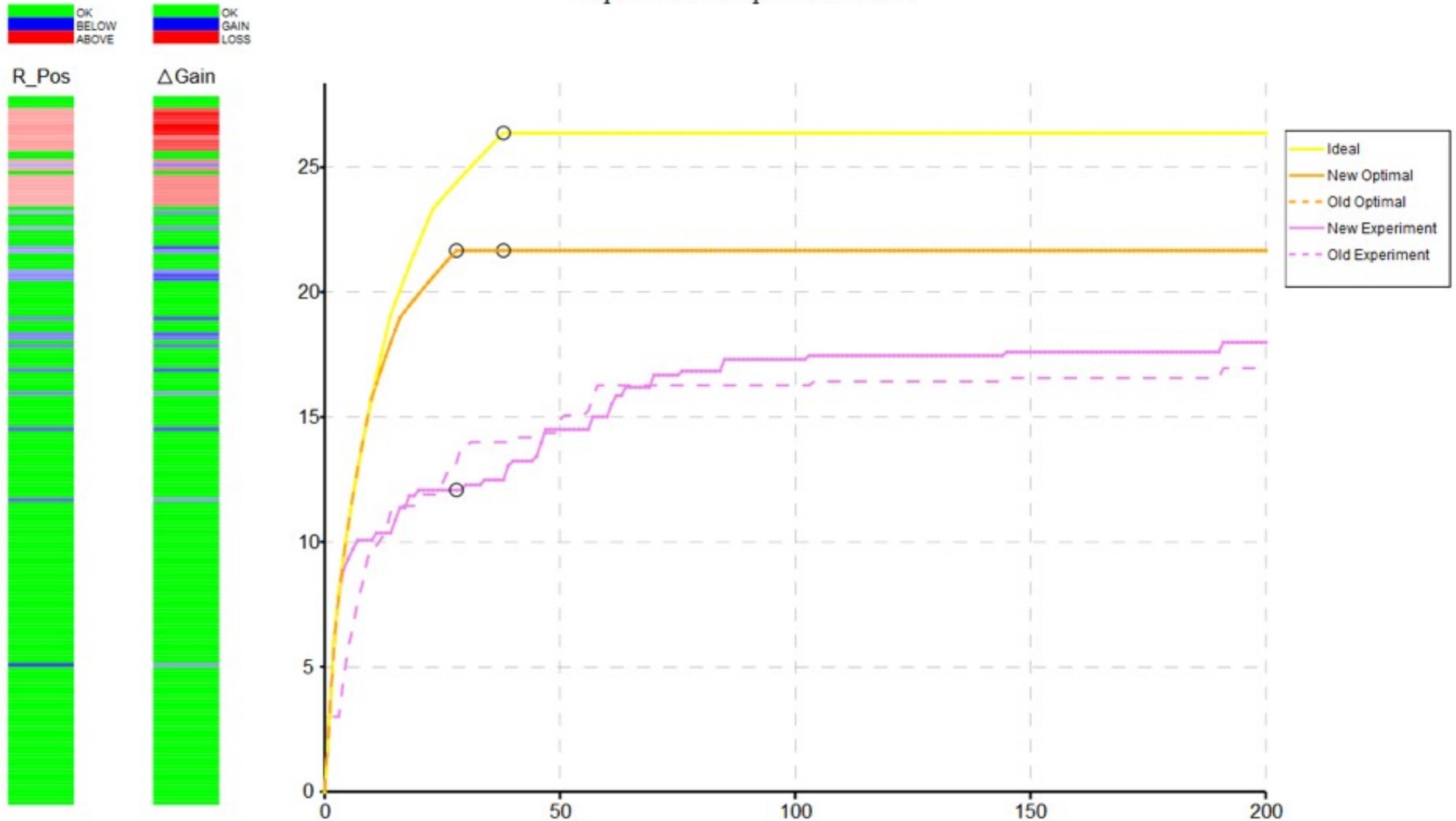
Visual comparison of Ranked Result Cumulated Gains

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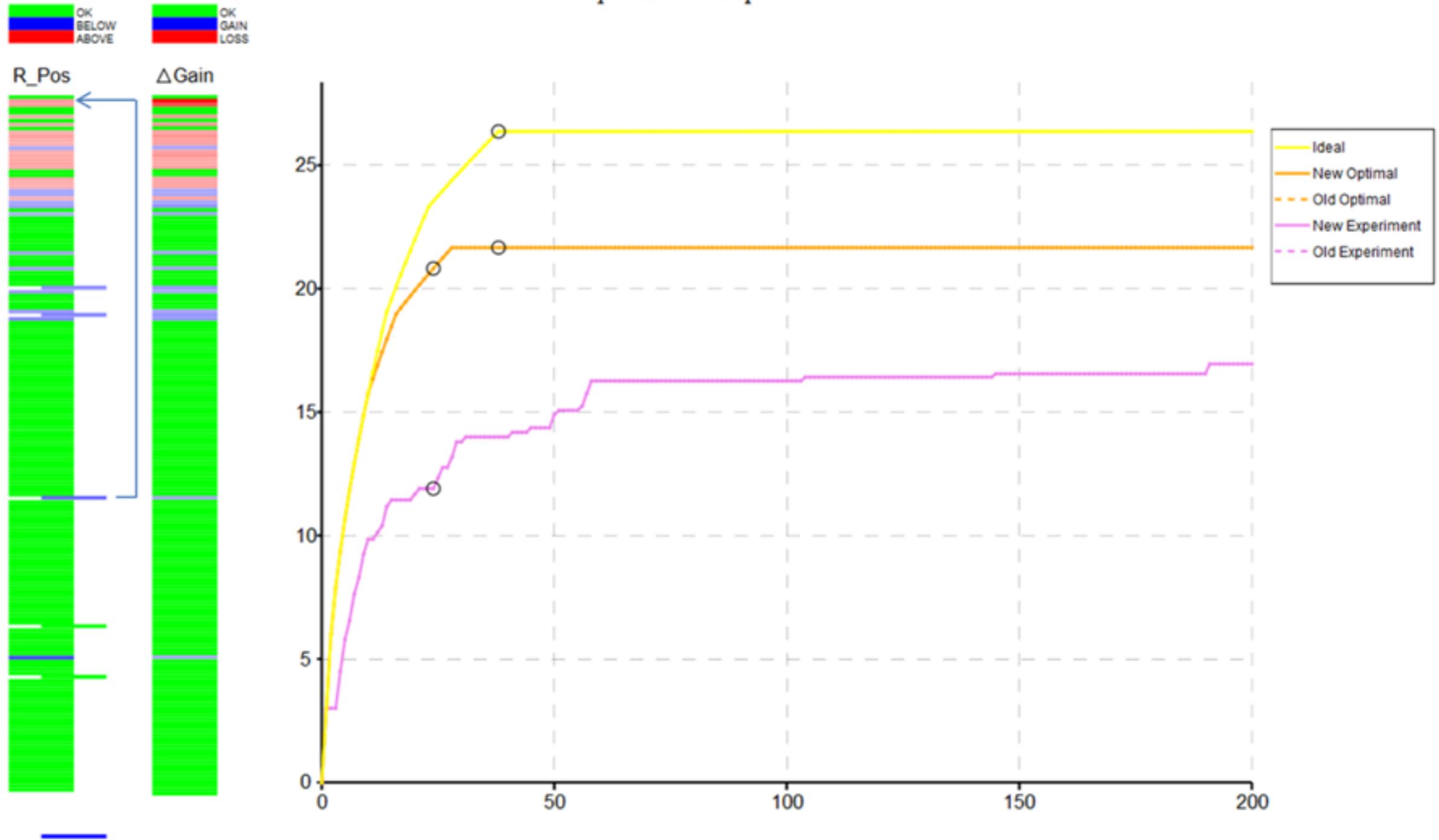
Visual comparison of Ranked Result Cumulated Gains

Topic ID:351 Experiment: bbn1



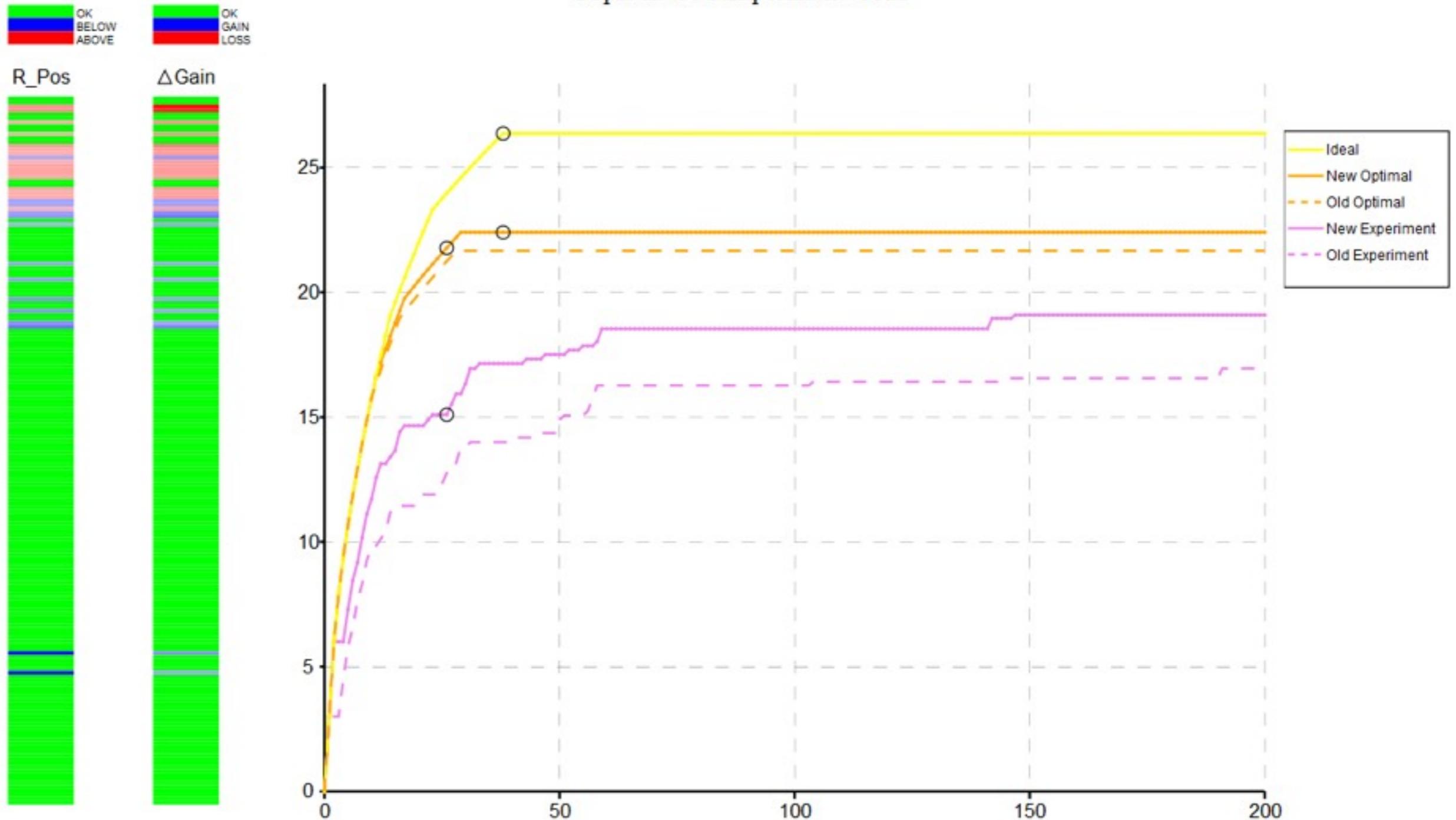
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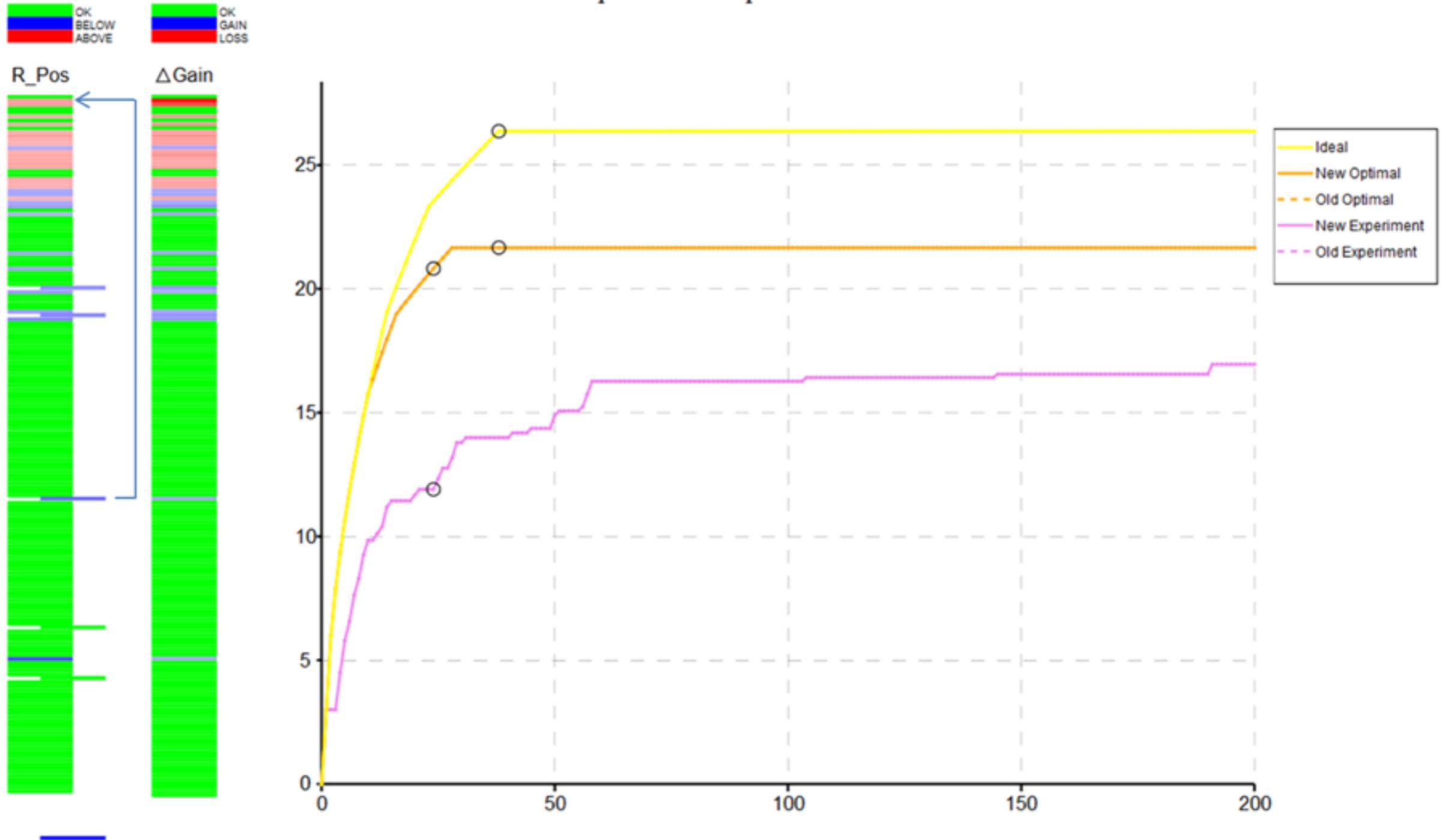
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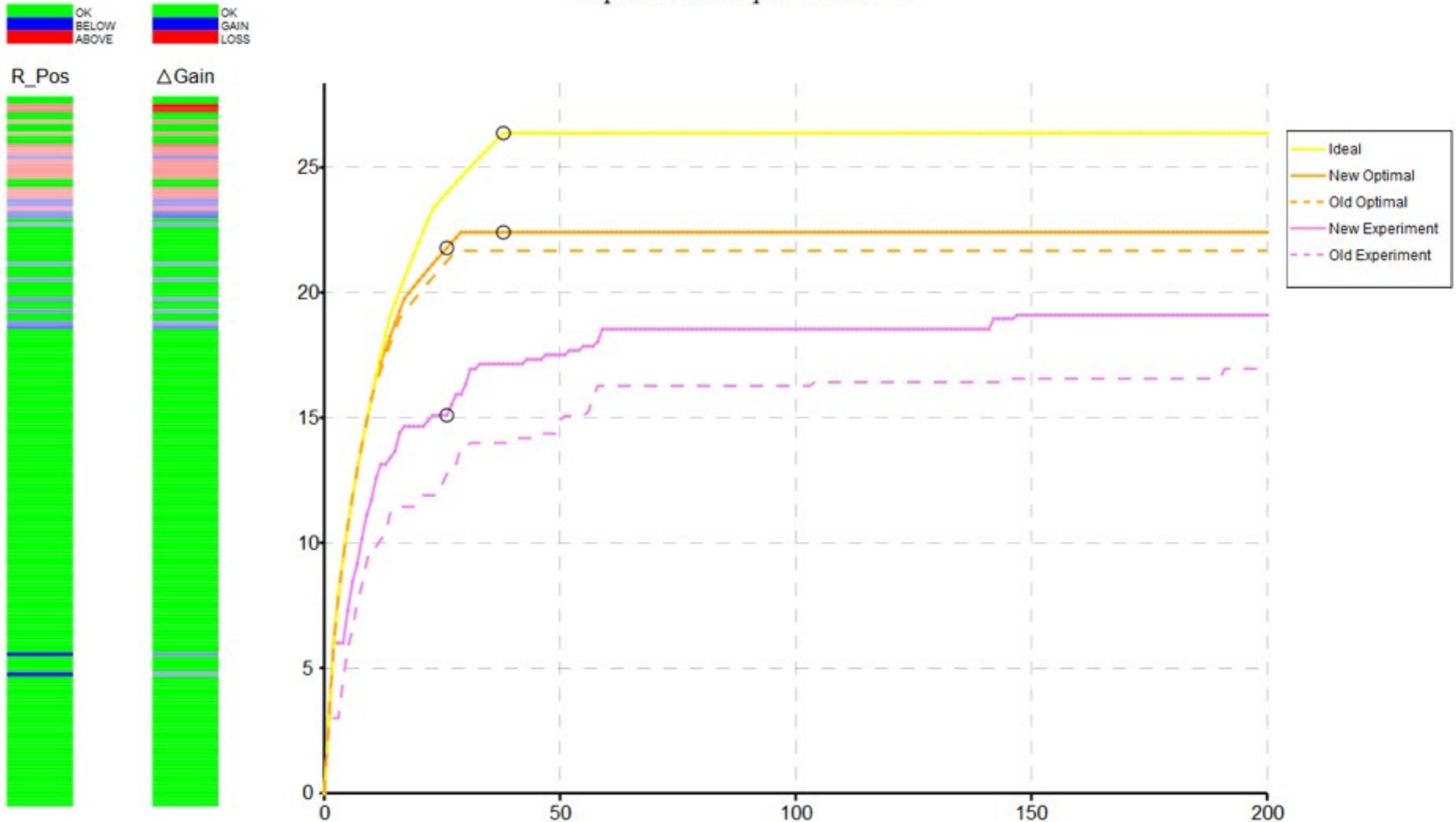
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Visual comparison of Ranked Result Cumulated Gains

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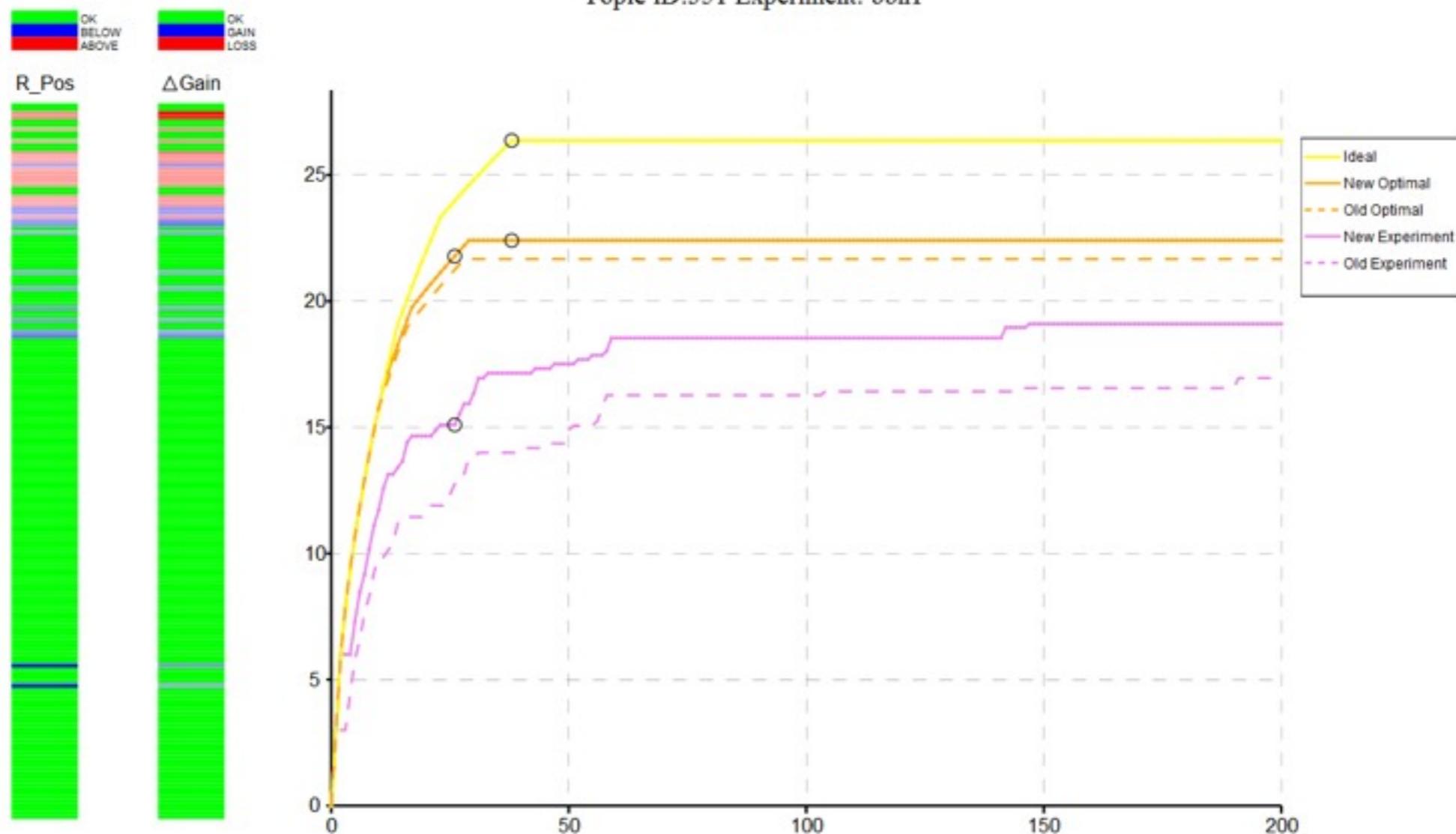


The Domino Effect

How does the change for topic 351 affect the other topics?

Visual comparison of Ranked Result Cumulated Gains

Topic ID:351 Experiment: bbn1

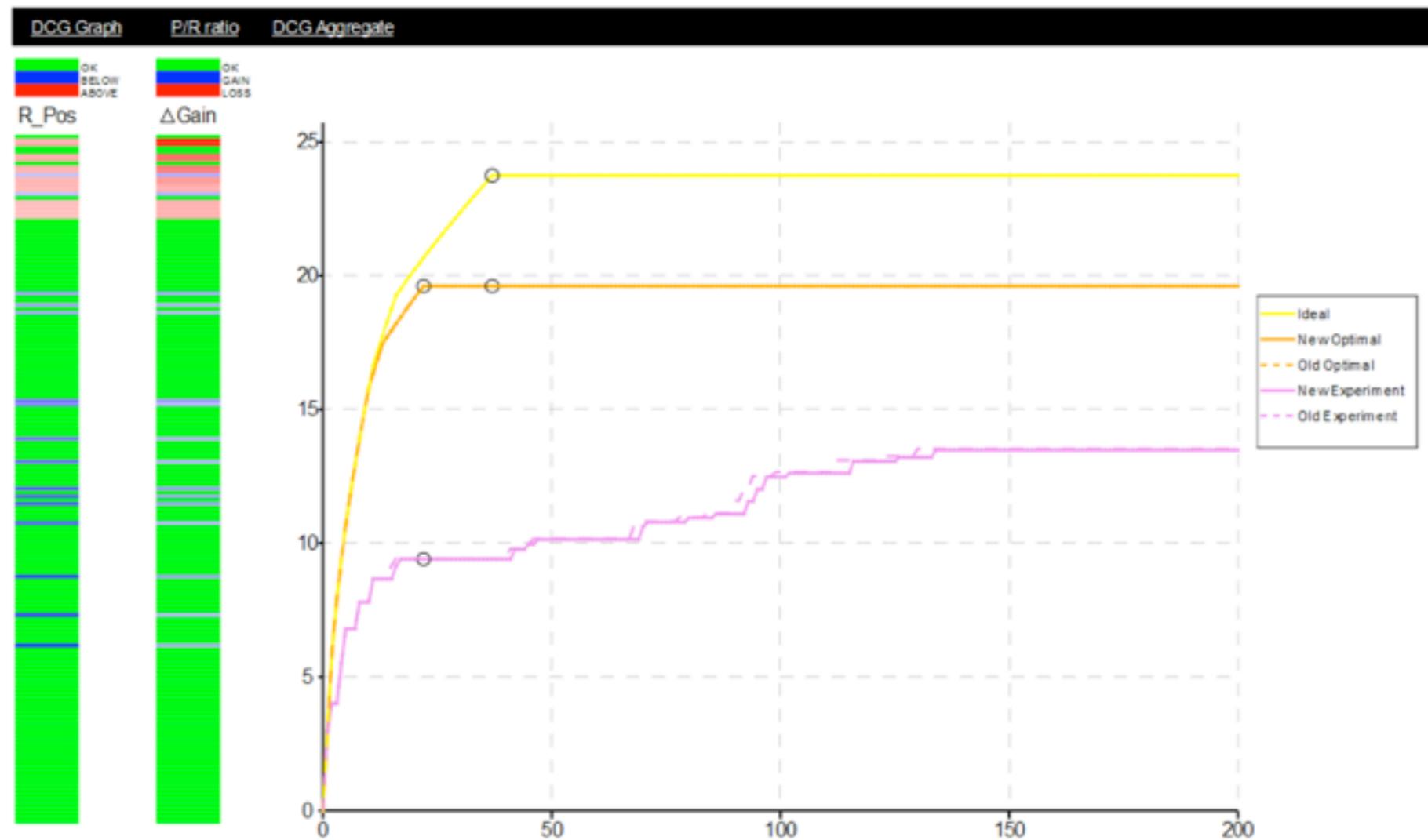


The Domino Effect

The change for 351 worsens the DCG curve of topic 355

Visual comparison of Ranked Result Cumulated Gains

Topic ID:355 Experiment: bbn1

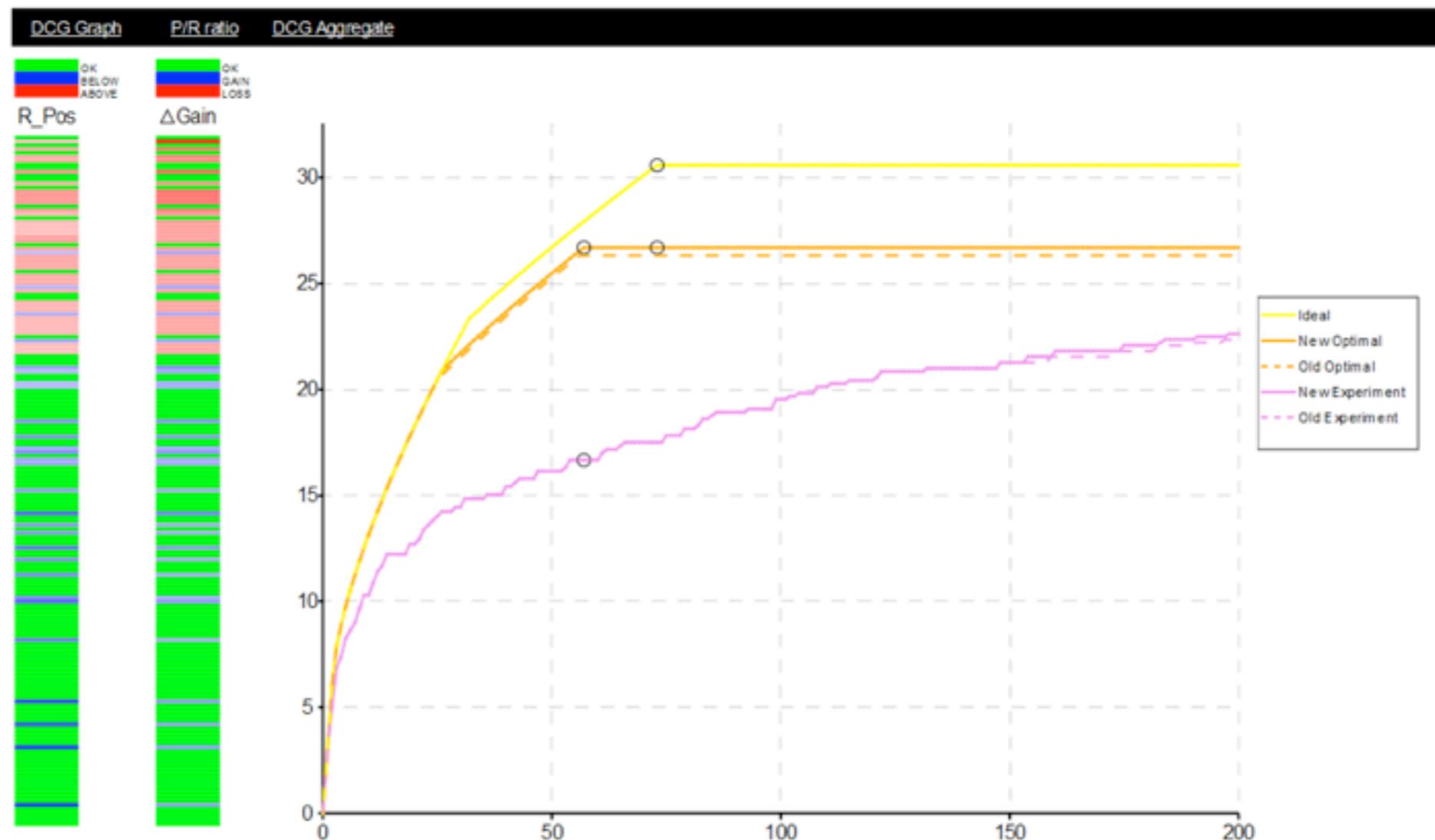


The Domino Effect

The change for 35 l improves the DCG curve of topic 400

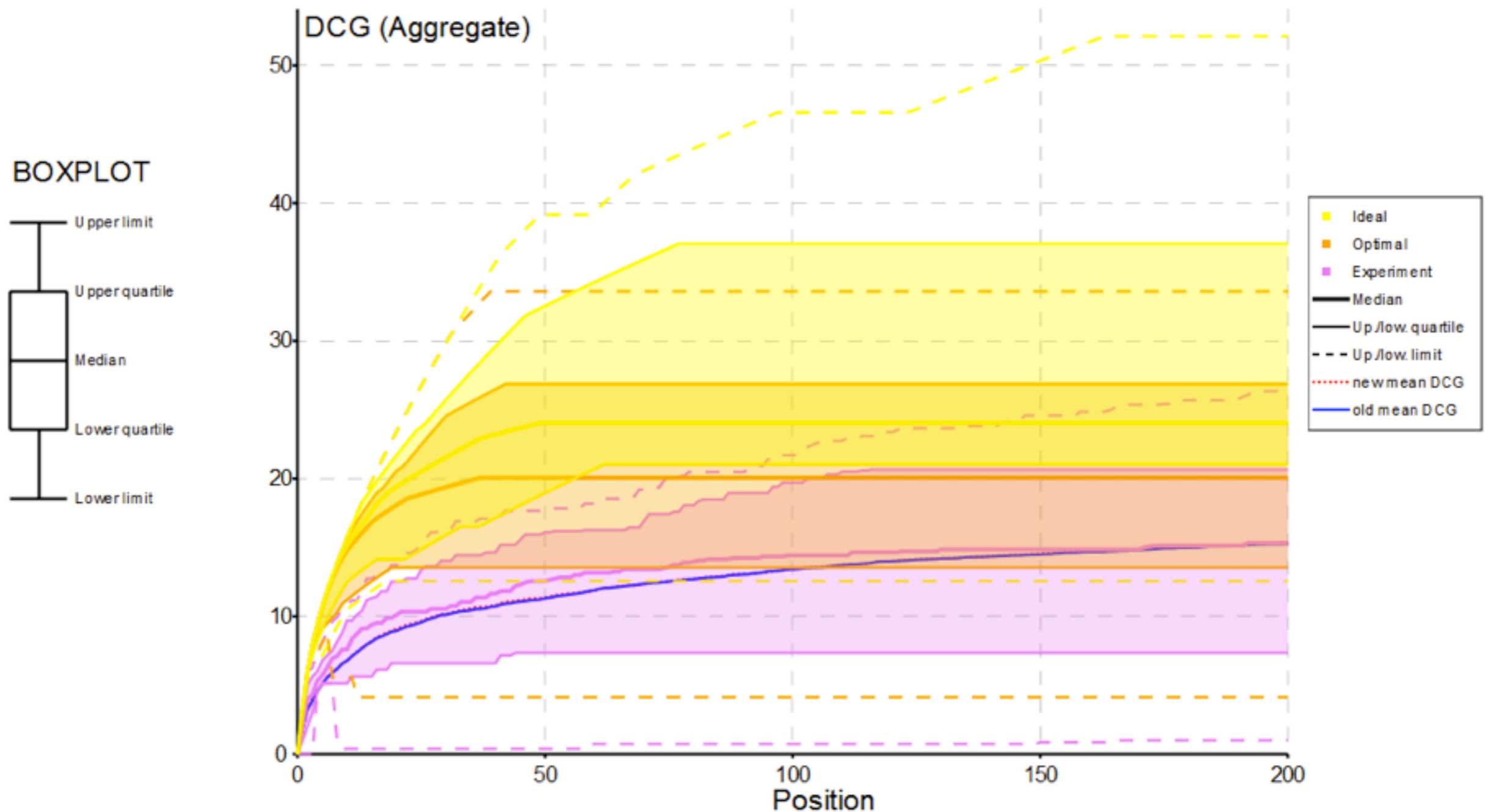
Visual comparison of Ranked Result Cumulated Gains

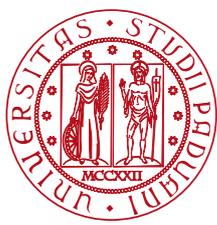
Topic ID:400 Experiment: bbn1



The Domino Effect

Aggregate view for the whole set of documents before and after the movement





Final Remarks and On-Going Work



- We presented the visual interactive tool allowing analysts to perform failure and what-if analyses
- We described the prototype implementing an actual fusion between IR evaluation and visual analytics tested on the TREC7 collection
- We are performing additional tests employing different learning to rank algorithms to construct the clusters
- We are investigating whether (and how) custom features extraction and selection may allow us to understand on which component of the IR system a change of ranking has an impact on