



### PAN 2010 Results

Uncovering Plagiarism, Authorship, and Social Software Misuse

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http://pan.webis.de



2nd International Competition on Plagiarism Detection, PAN 2010

These days, plagiarism and text reuse is rife on the Web.

#### Task:

Given a set of suspicious documents and a set of source documents, find all plagiarized sections in the suspicious documents and, if available, the corresponding source sections.

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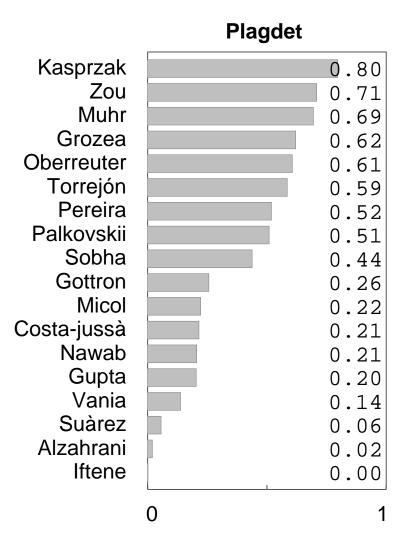
#### Corpus:

#### PAN-PC-10

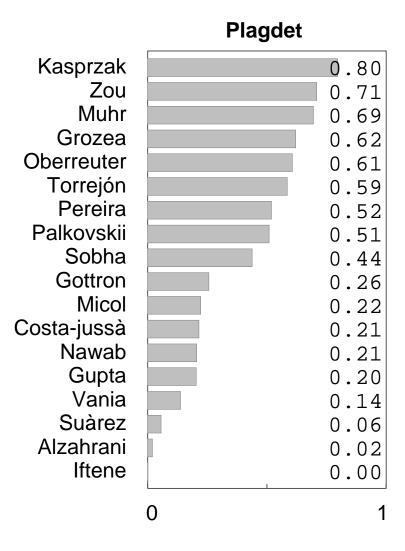
- □ 27 073 documents (obtained from 22 874 books from the Project Gutenberg)
- □ 68 558 plagiarism cases (about 0-10 cases per document)
- □ 6 plagiarism-relevant parameters (length, language, task, obfuscation, topic, fraction)

[Potthast et al., COLING 2010]

### Plagiarism Detection Results



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 Plagdet combines precision, recall, and granularity:

$$\textit{plagdet}(S,R) = \frac{F_1}{\log_2(1 + \textit{gran}(S,R))}$$

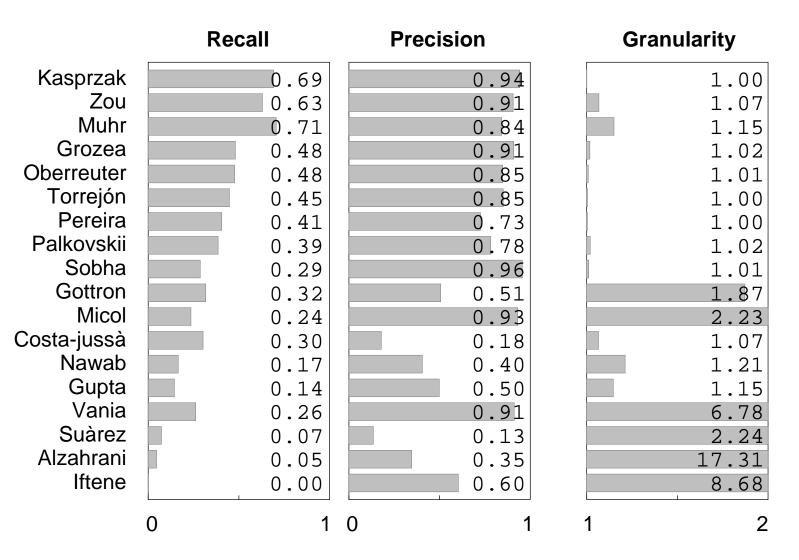
$$\textit{prec}(S,R) = \frac{1}{|R|} \sum_{r \in R} \frac{|\bigcup_{s \in S} (\mathbf{s} \sqcap \mathbf{r})|}{|\mathbf{r}|}$$

$$rec(S, R) = \frac{1}{|S|} \sum_{s \in S} \frac{|\bigcup_{r \in R} (\mathbf{s} \sqcap \mathbf{r})|}{|\mathbf{s}|}$$

□ The granularity *gran* measures the average number of times a plagiarism case is detected.

[Potthast et al., COLING 2010]

### Plagiarism Detection Results





1st International Competition on Wikipedia Vandalism Detection, PAN 2010

Every edit on Wikipedia has to be double-checked for integrity even if it affects just one char.

Task:

Given a set of edits on Wikipedia articles, distinguish ill-intentioned edits from well-intentioned edits.

1st International Competition on Wikipedia Vandalism Detection, PAN 2010

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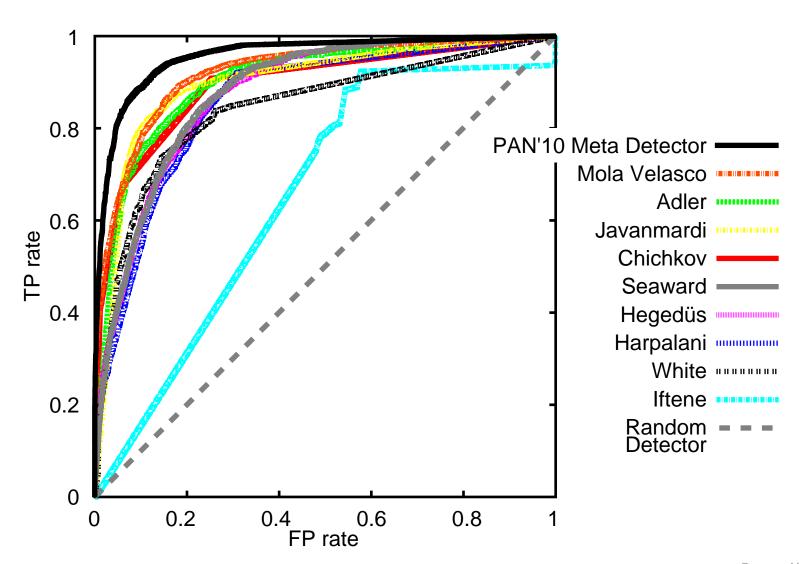
### Corpus:

#### PAN-WVC-10

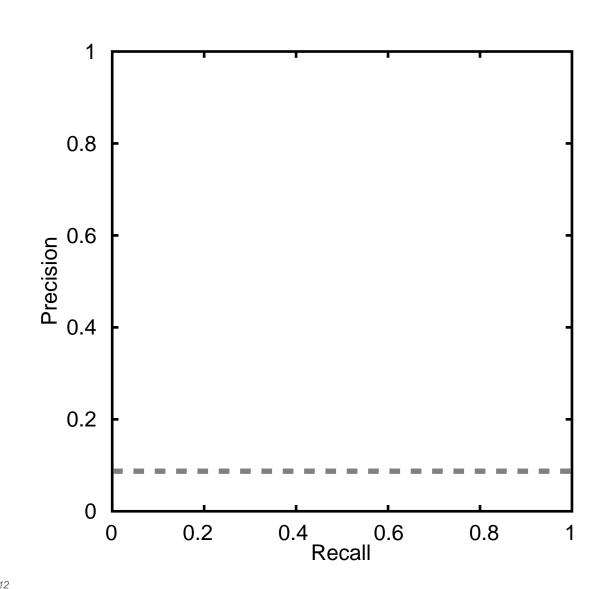
- □ 32 452 edits (sampled from a week's worth of Wikipedia edit logs)
- □ 28 468 different edited articles (edit frequency resembles article importance)
- □ 2391 edits are vandalism (a 7% ratio is in concordance with the literature)

[Potthast, SIGIR 2010]

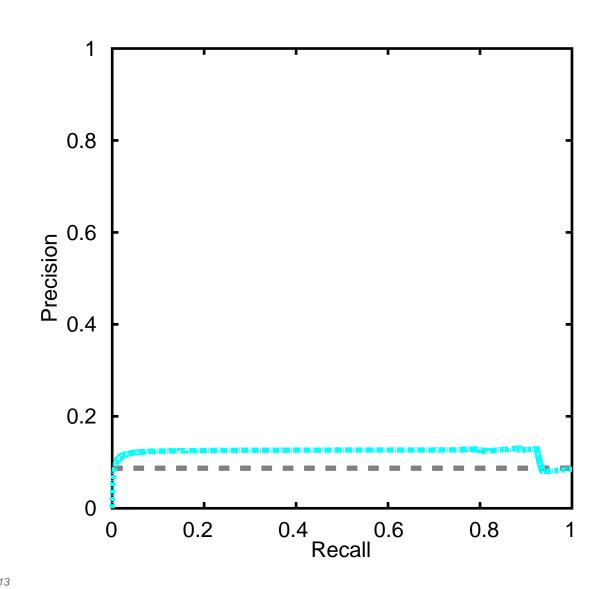
### Plagiarism Detection Results



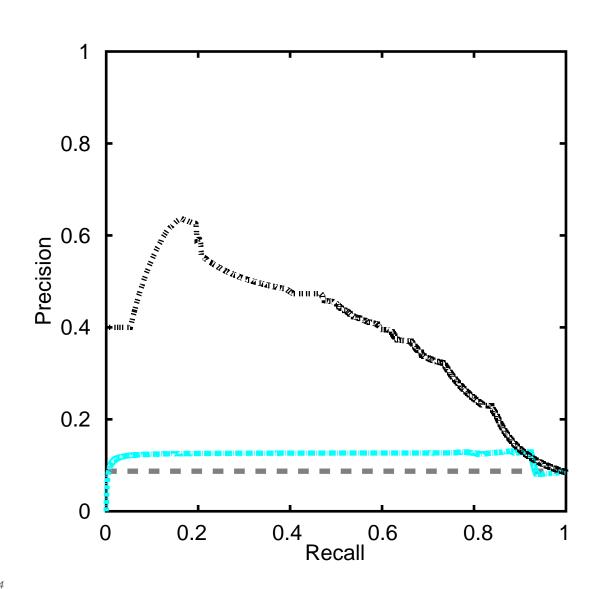
### Vandalism Detection Results

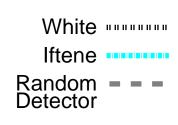


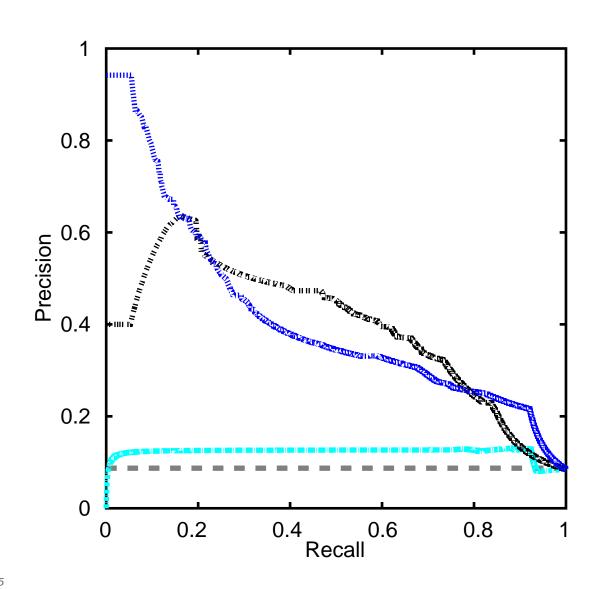
Random Detector

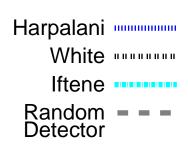


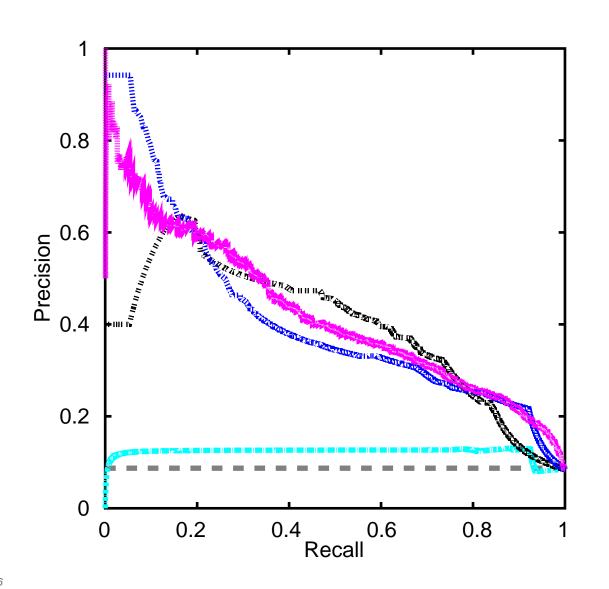


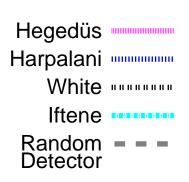


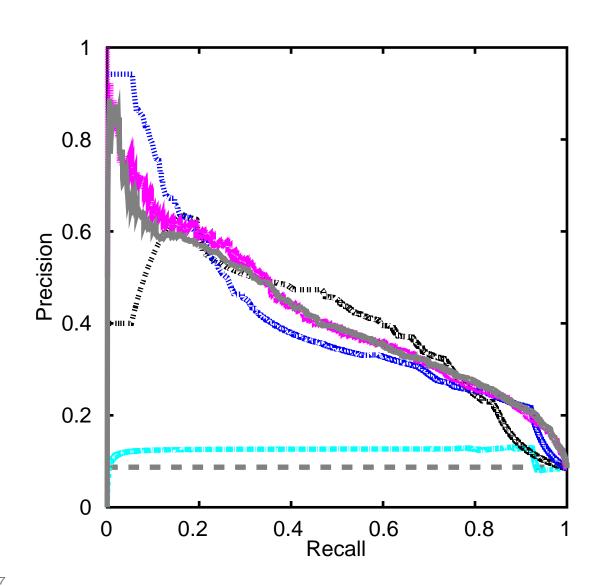


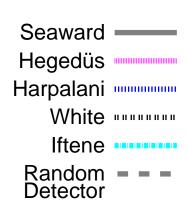


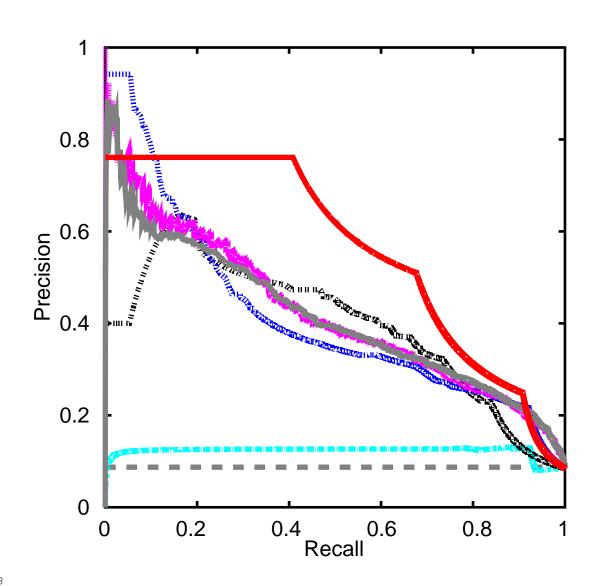


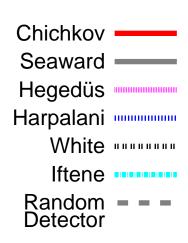


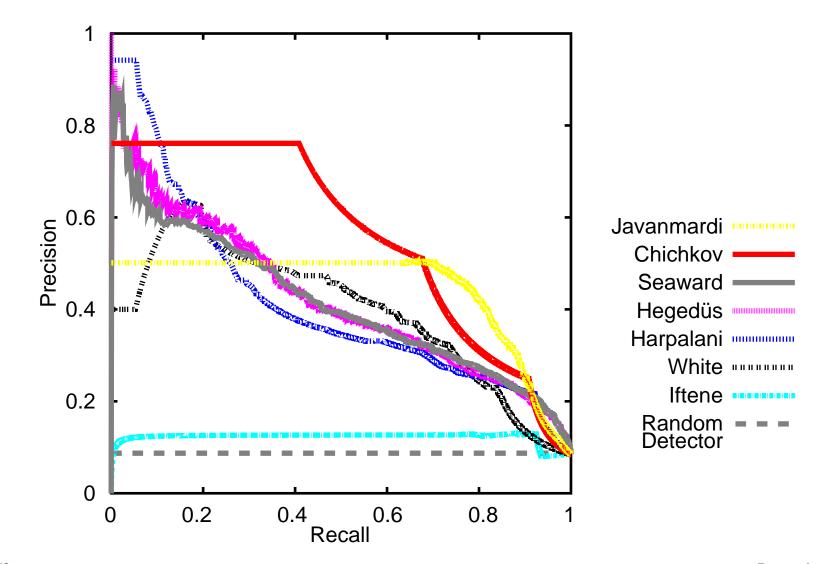


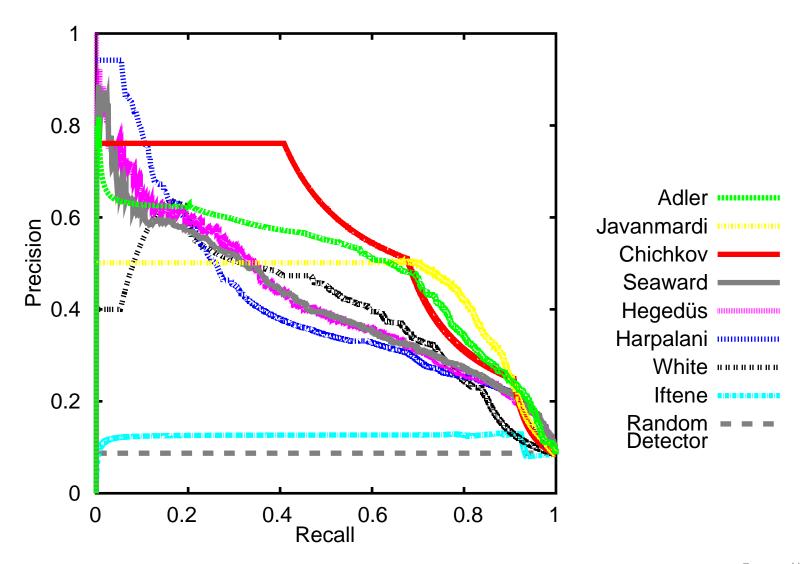


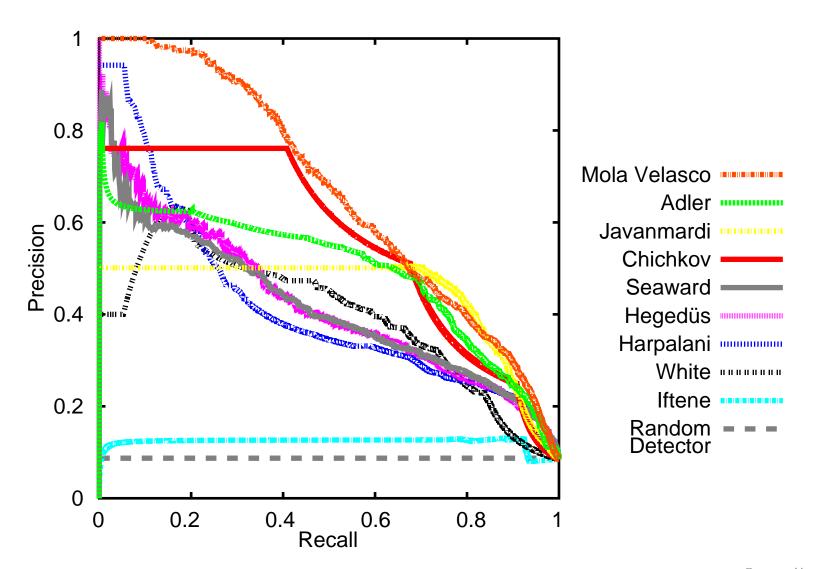


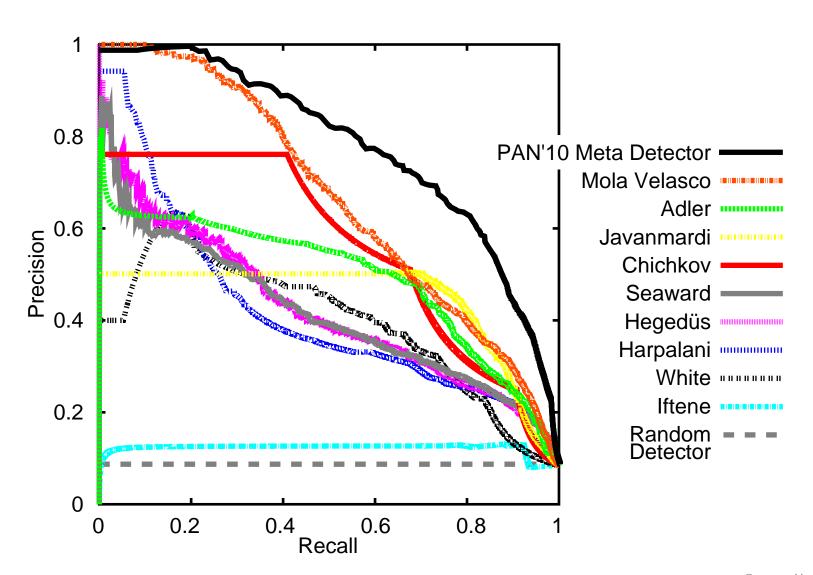




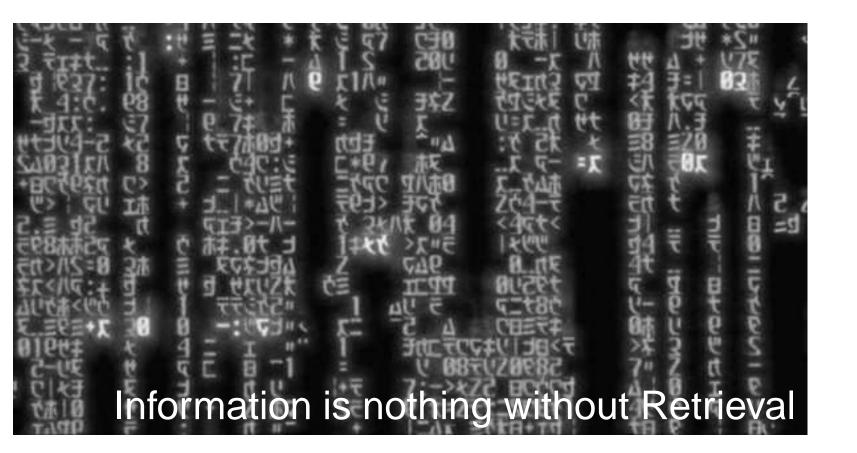


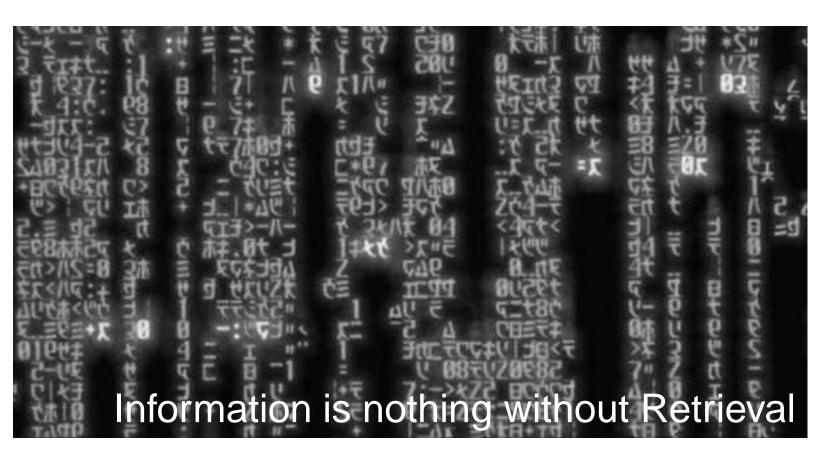






ROC-AUC	ROC rank	PR-AUC	PR rank		Detector
0.95690	_	0.77609	_	_	PAN'10 Meta Detector
0.92236	1	0.66522	1	_	Mola Velasco
0.90351	2	0.49263	3	$\downarrow$	Adler
0.89856	3	0.44756	4	$\downarrow$	Javanmardi
0.89377	4	0.56213	2	$\uparrow\uparrow$	Chichkov
0.87990	5	0.41365	7	$\downarrow \downarrow$	Seaward
0.87669	6	0.42203	5	$\uparrow$	Hegedus
0.85875	7	0.41498	6	$\uparrow$	Harpalani
0.84340	8	0.39341	8	_	White
0.65404	9	0.12235	9	_	Iftene
0.50000	10	0.08490	10	_	Random Detector





Retrieval is nothing without Evaluation