

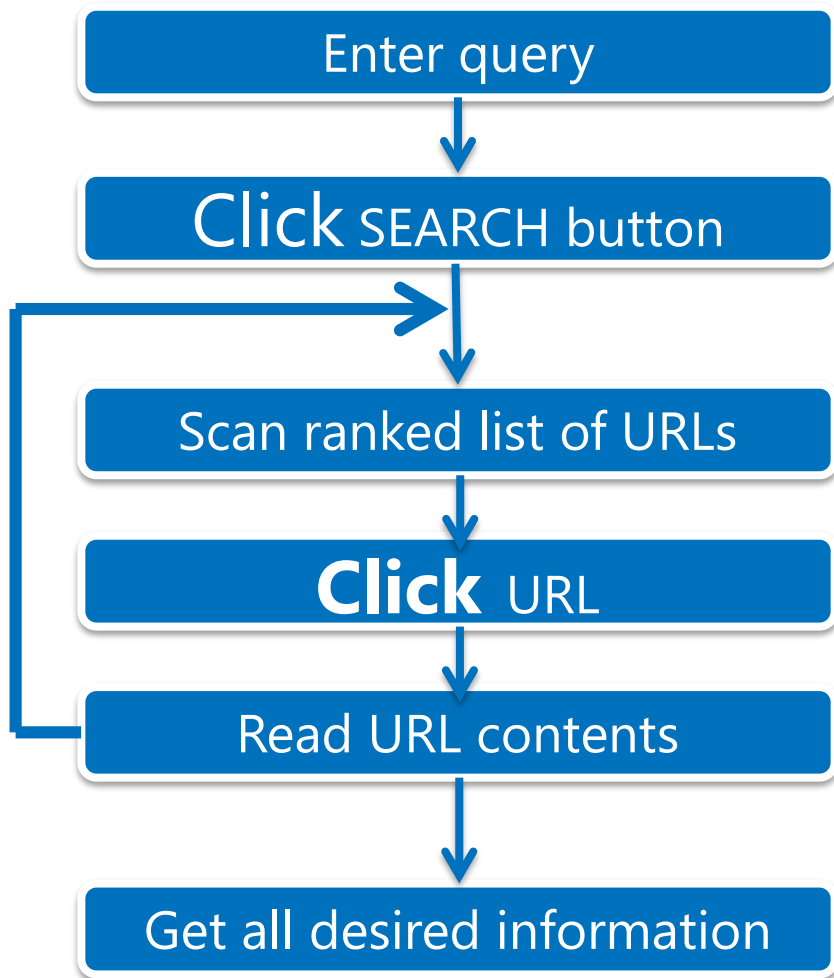
Two-layered Summaries for Mobile Search: Does the Evaluation Measure Reflect User Preferences?

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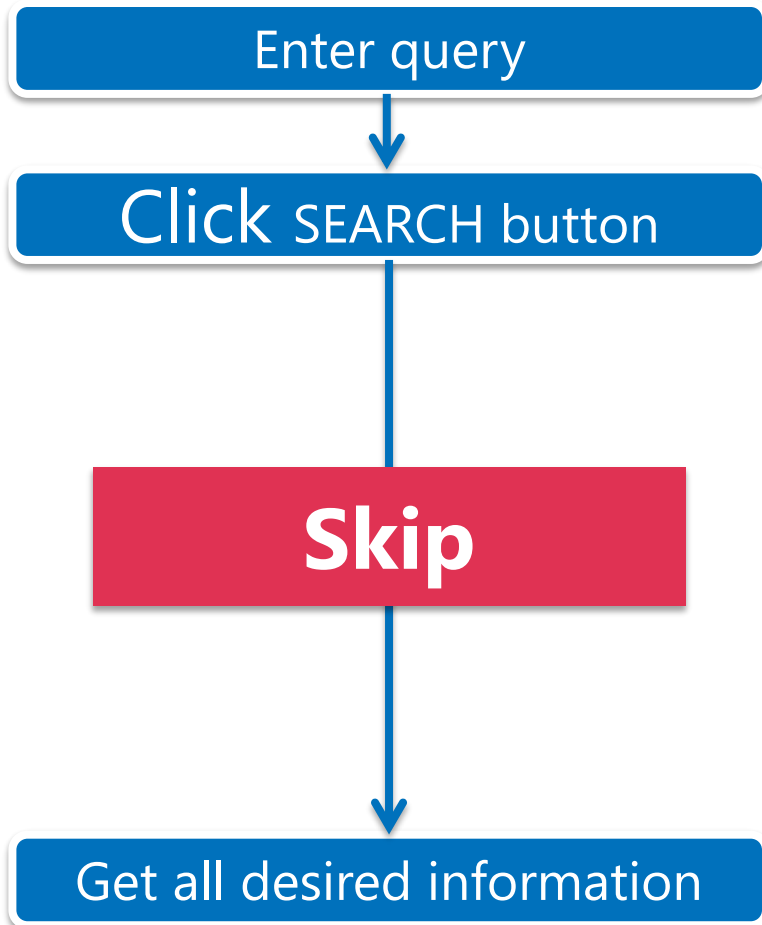
MOTIVATION AND TASK

IR Systems in *Ten-Blue-Link* Paradigm

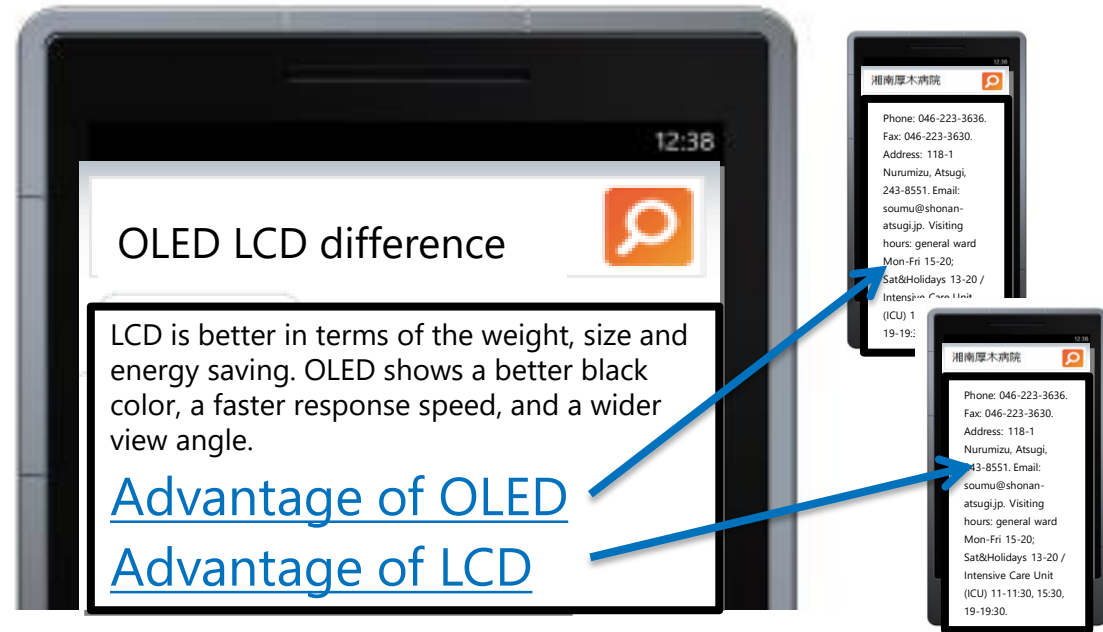


Long way to get all desired information

MobileClick System



System output



Task: Given a search query, return a **two-layered** textual output

Go beyond the "ten-blue-link" paradigm, and tackle **information** retrieval rather than document retrieval

iUnit Summarization Subtask at NTCIR-12

- Given a query, a set of iUnits, and a set of intents, generate a **two-layered summary**

Input: Query



Input: iUnit set

iUnit
A series of evaluation workshops
Designed to enhance IA research
...

Input: Intents



Output: **Two-layered summary**

The NTCIR Workshop is a series of evaluation workshops designed to enhance research in information access technologies including information retrieval, summarization, extraction, question answering, etc.

News

Schedule →

2nd layer

20/Jan./2016:	Task Registration Due
06/Jan./2016:	Document Set Release
Jan.-May/2016:	Dry Run
Mar.-July/2016:	Formal Run
01/Aug./2016:	Evaluation Results Due
01/Aug./2016:	Task overview release
15/Sep./2016:	Paper submission Due

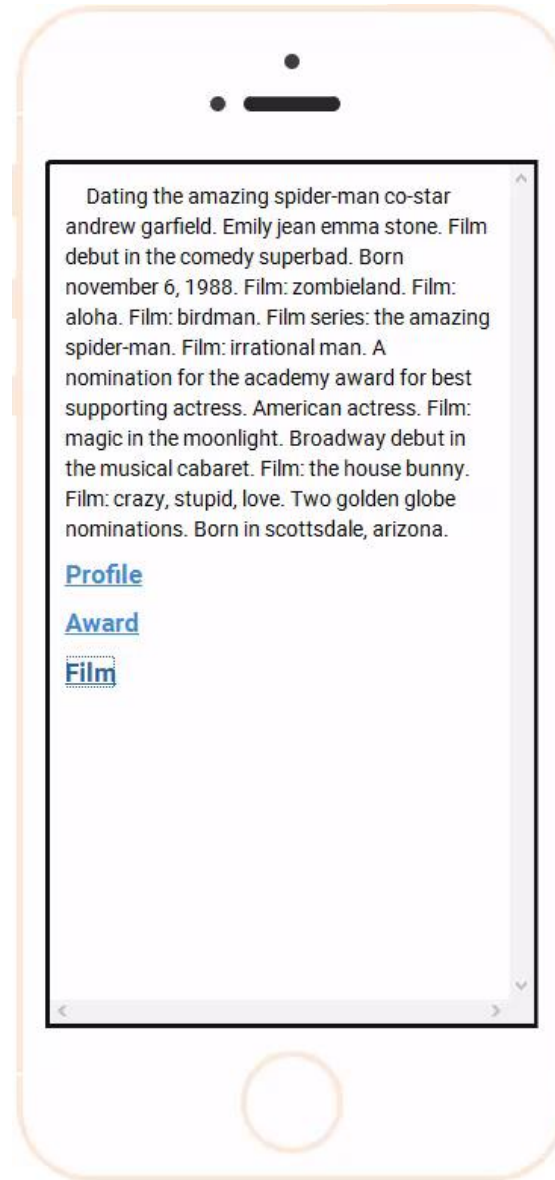
Evaluation metric designed for mobile information access

M-measure
0.5

Challenge

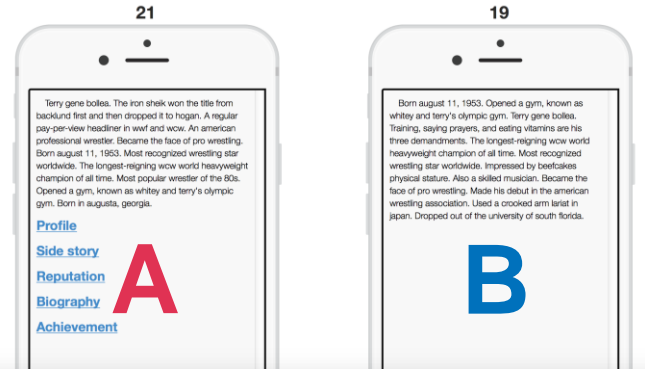
Lay out iUnits so that any types of users can be immediately satisfied

Two-layered Summary in Action

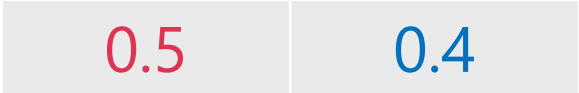


Research Question Addressed in This Work

Does the Evaluation Measure Reflect User Preferences?



M-measure



Which is higher?

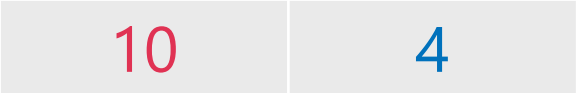


Which is better?



User preference

(# of users who prefer to A (B))

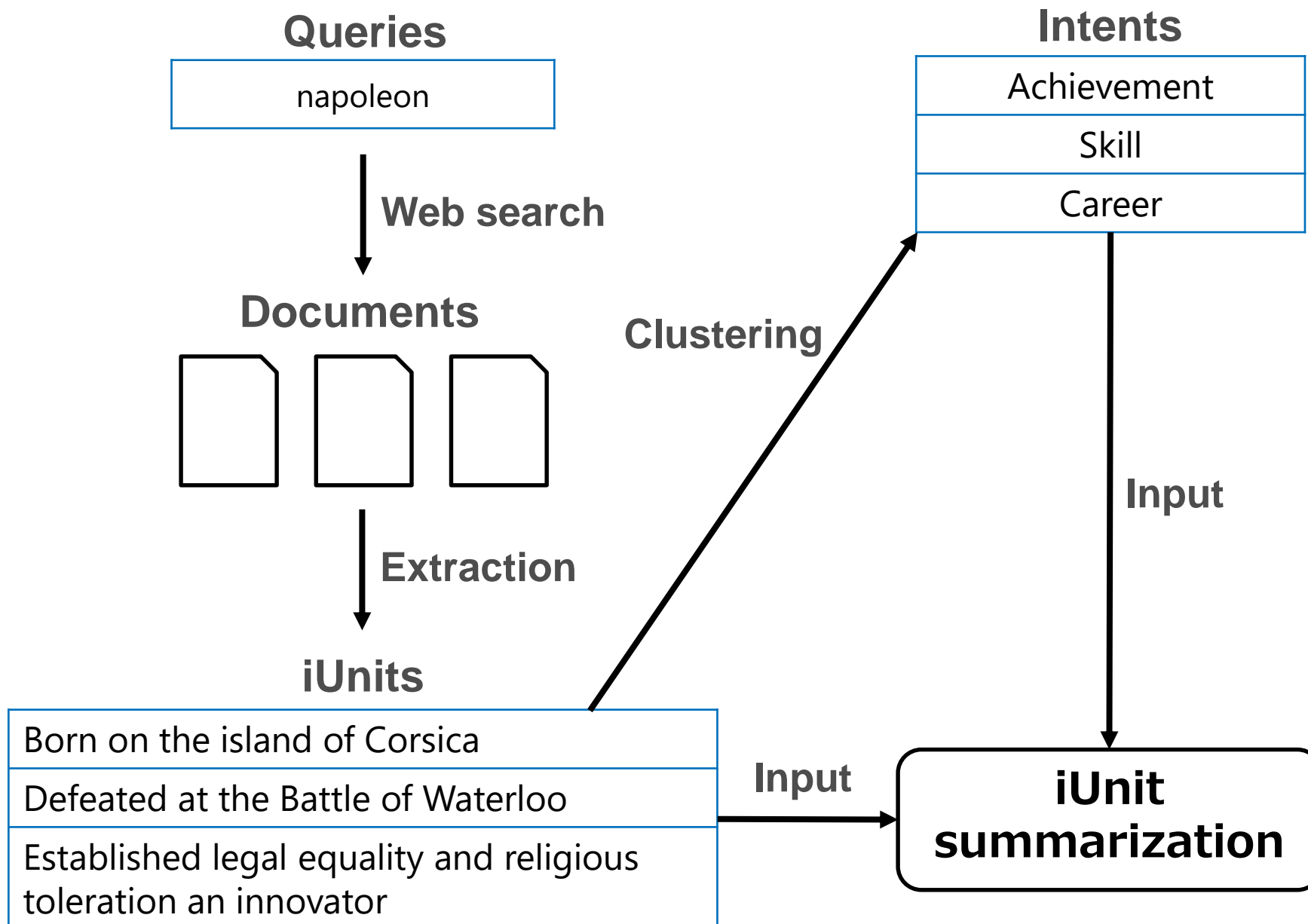


||| Same?



DATA

Overview of Data



Queries and Documents

- **Queries**

- 100 English/Japanese queries
- Most of which were ambiguous/underspecified
- **Selected from five categories:**
celebrity, location, definition, and QA (similar to NTCIR 1CLICK-2)

Examples

CELEBRITY	LOCATION	DEFINITION	QA
hulk hogan	bank adelanto	bitcoin	what is mirror made of
bruno mars	cafe killeen	divers disease	how to cook coleslaw
sharon stone	cincinnati art museum	windows 7	role of animal tail

- **Documents**

- 500 commercial search engine results for each query
from which iUnits were extracted

- **Definition**

- Atomic information pieces relevant to a given query

- The number of iUnits

- **2,317** (23.8 iUnits per query) for English

- **4,169** (41.7 iUnits per query) for Japanese

Examples of iUnits for query “Napoleon”

Born on the island of Corsica	General of the Army of Italy
Defeated at the Battle of Waterloo	One of the most controversial political figures won at the Battle of Wagram
Established legal equality and religious toleration an innovator	Baptised as a Catholic
Absent during Peninsular War	Cut off European trade with Britain

- **An intent can be defined as**
 - **A specific interpretation of an ambiguous query** (“Mac OS” and “car brand” for “jaguar”), or
 - **An aspect of a faceted query** (“windows 8” and “windows 10” for “windows”)

- **Obtained by clustering iUnits**

iUnits

Born on the island of Corsica
Defeated at the Battle of Waterloo
Established legal equality and religious toleration an innovator
Absent during Peninsular War

Clustering



Intents

Achievement
Skill
Career

EVALUATION

Per-intent iUnit Importance and Intent Probability

- **Importance of iUnits in terms of an intent**

In terms of intent “Definition”

iUnit	Importance
A series of evaluation workshops	5
Task Registration Due 20/Jun./2016	3

In terms of intent “Schedule”

iUnit	Importance
A series of evaluation workshops	2
Task Registration Due 20/Jun./2016	5

- **Intent probability $P(i|q)$**

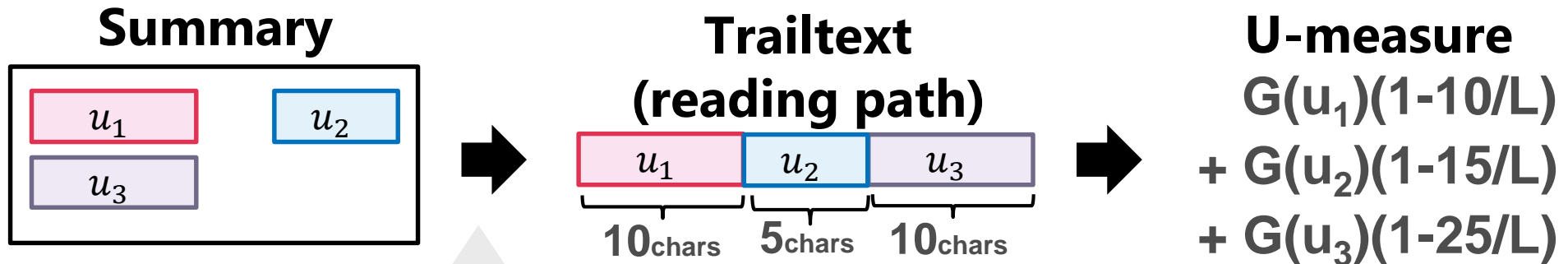
- Probability of having intent i for a given query q

Intent	Prob.
Definition	0.4
Schedule	0.3
Tasks	0.3

For details, see our MobileClick-2 overview paper

Evaluation of iUnit Summarization (Single-layer Case)

- Consider single-layered summary evaluation
- **U-measure** [Sakai and Dou. SIGIR2013]
 - Higher if more important iUnits appear earlier



Create a list of iUnits by assuming that users read text from left to right, from top to bottom

$$U = \sum_{r=1} G(u_r) \left(1 - \frac{\text{pos}(u_r)}{L} \right)$$

u_r : r-th iUnit

$G(u)$: importance of u

$\text{pos}(u)$: offset of u from the beginning

L : patience parameter

- **M-measure**

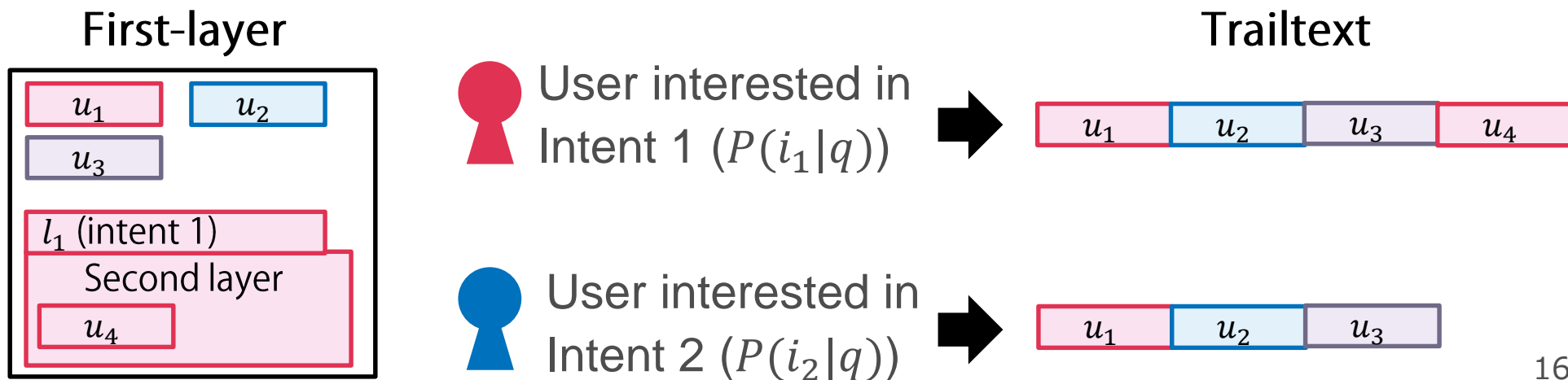
- **Expectation of U-measure over multiple *trailtexts***

$$M = \sum_{\mathbf{t}} P(\mathbf{t})U(\mathbf{t})$$

$P(\mathbf{t})$: probability of trailtext \mathbf{t}
 $U(\mathbf{t})$: U-measure of trailtext \mathbf{t}

1. Generate trailtexts by assuming that

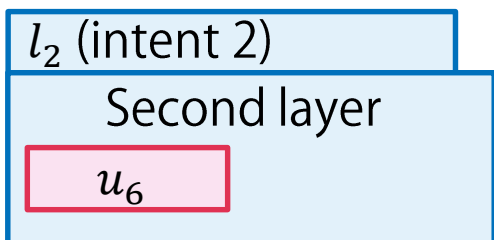
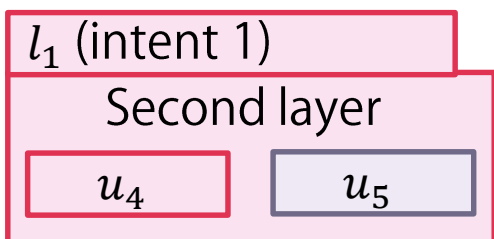
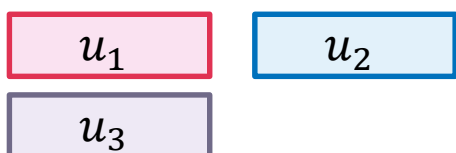
- Users read a summary from the top of the first layer
- Users click on an intent if they are interested in it



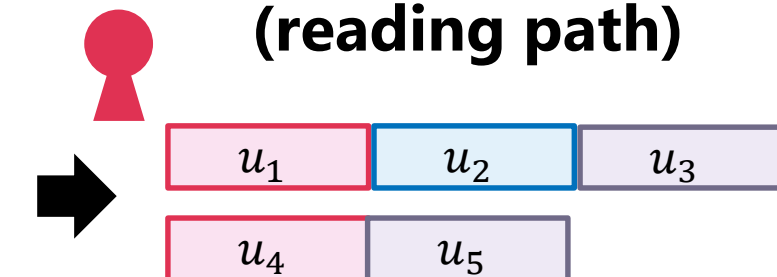
Evaluation of iUnit Summarization (Two-layer Case)

2. Compute the expectation of U-measure

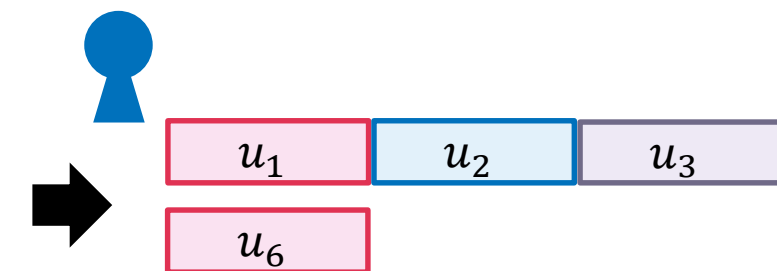
First layer



Trailtext (\mathbf{t})
(reading path)



$$P(\mathbf{t}_1) = P(i_1|q) = 0.75$$



$$P(\mathbf{t}_2) = P(i_2|q) = 0.25$$

U M-measure

0.44

0.12

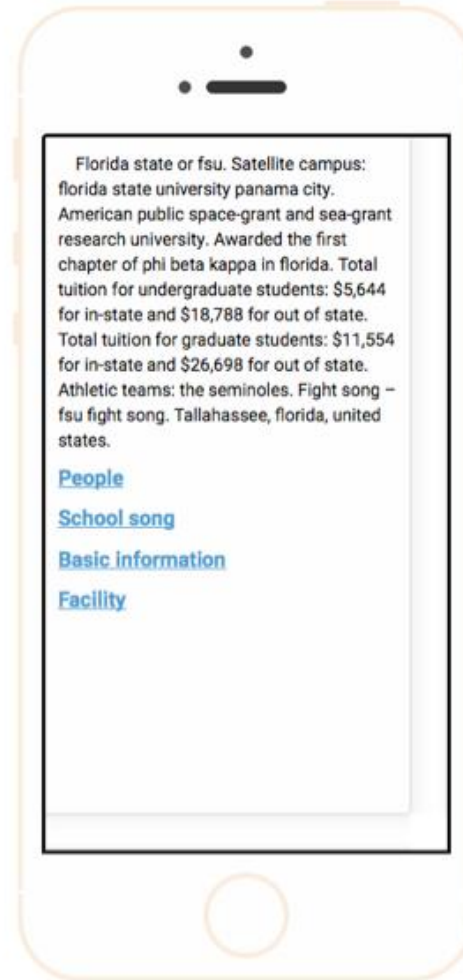
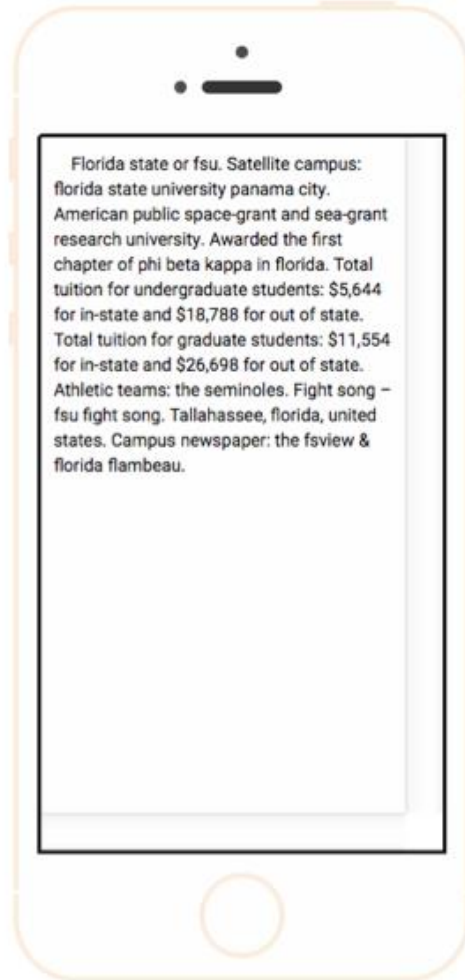
0.36

$$M = \sum_{\mathbf{t}} P(\mathbf{t})U(\mathbf{t})$$

Because trailtext \mathbf{t}_2 is read by users interested in i_2

EXPERIMENT

Pairwise Comparison



All possible pairs of 7 summaries for 25 queries were presented to about 14 users

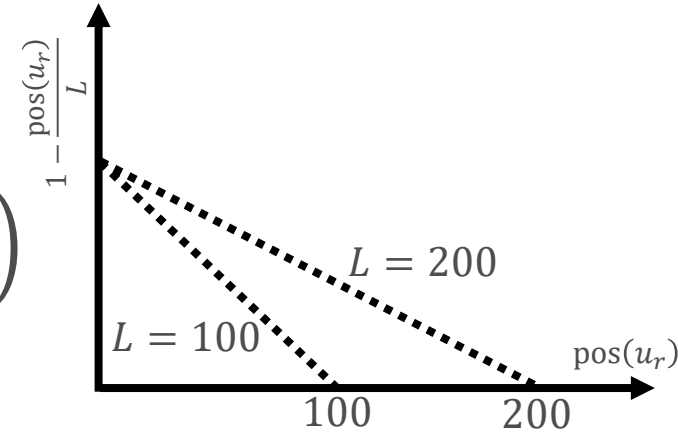
Instruction in Pairwise Comparison

- **Users were asked to select either**
the left one is better,
the right one is better,
equally good, or
equally bad
- **Criteria:**
 - (1) **How much useful information you can get from the summary, and**
 - (2) **How quickly you can get useful information from the summary**

Settings of M-measure

- **L of U-measure in M-measure**

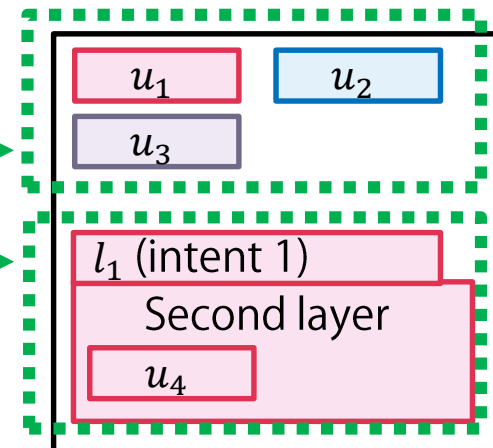
- $U = \sum_{r=1} G(u_r) \max\left(0, 1 - \frac{\text{pos}(u_r)}{L}\right)$



- L is a patience parameter that controls how the gain of iUnits decreases as the user reads the text

- **Simple variants of M-measure**

- Use only first layer
 - Use only second layer
 - Use a uniform distribution for $P(i|q)$

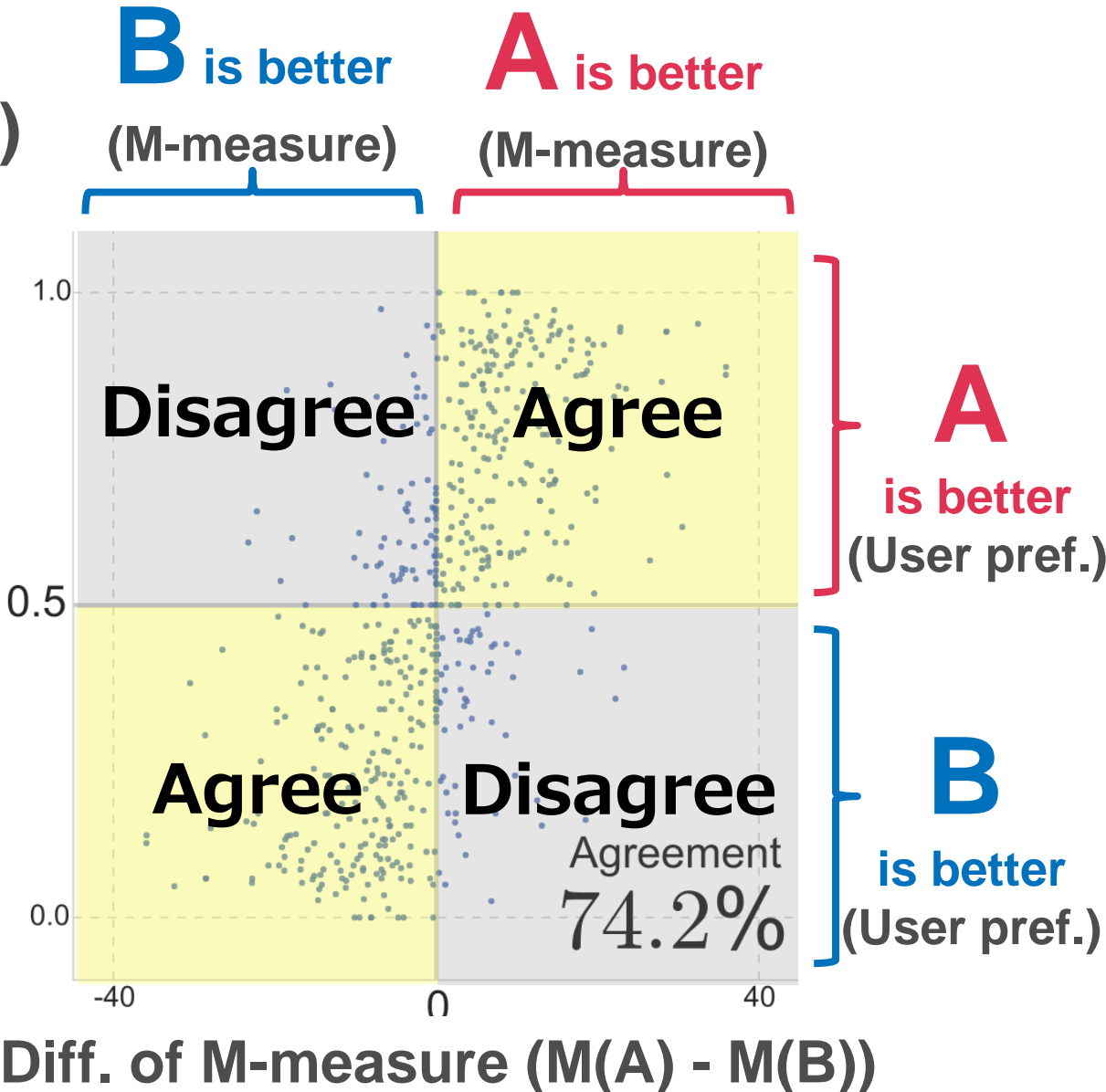


Interpretation of Results

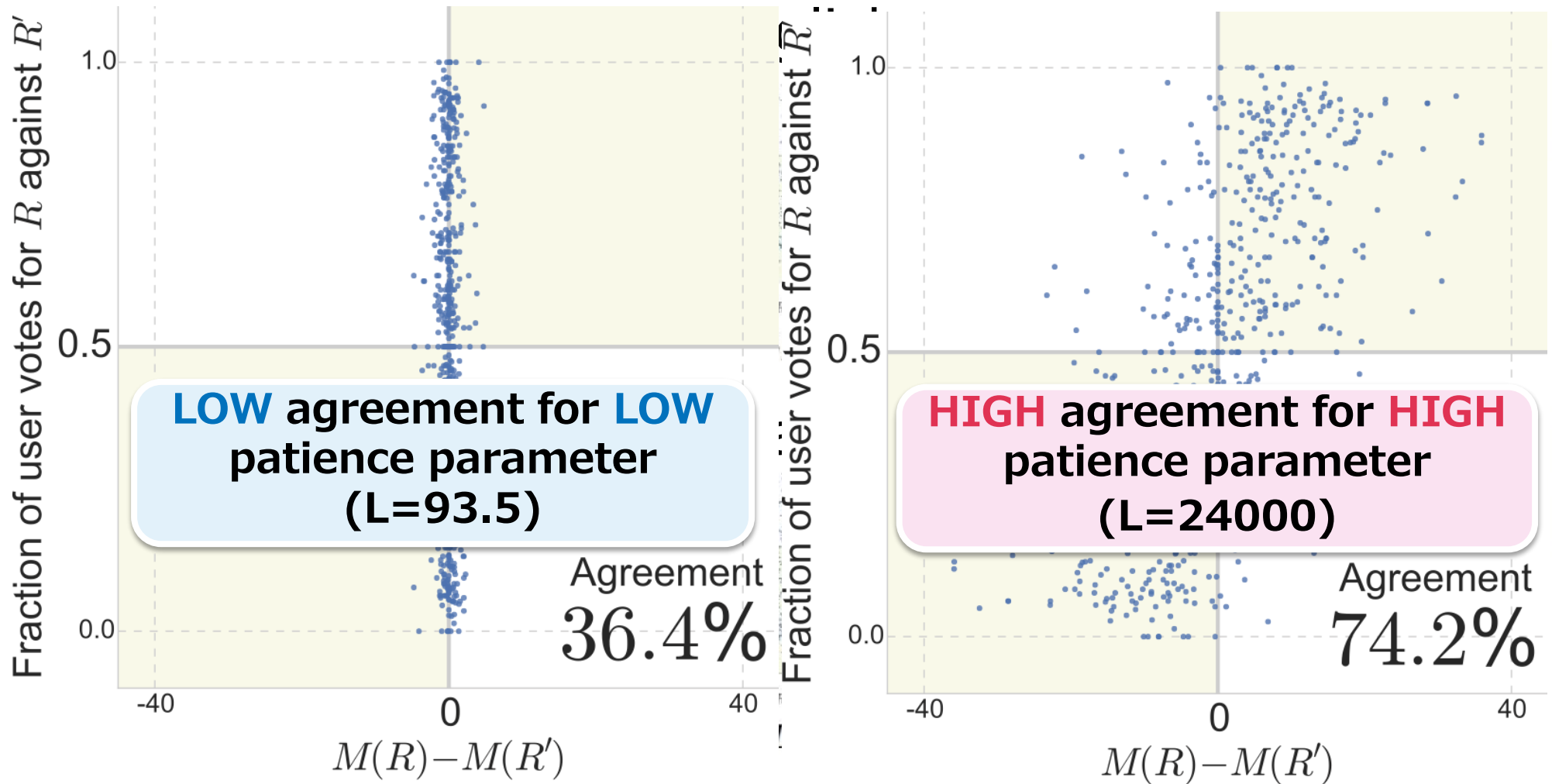
Each dot represents a pair of systems (A, B) for a particular query

$$\frac{\text{(Num. of votes for A)}}{\text{(Total num. of votes)}}$$

Agreement
= (#dots in Agree) / (#dots)



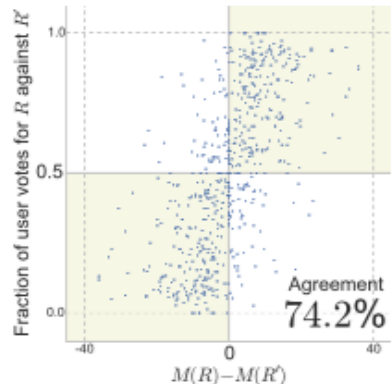
Experimental Results for Different Patient Parameters



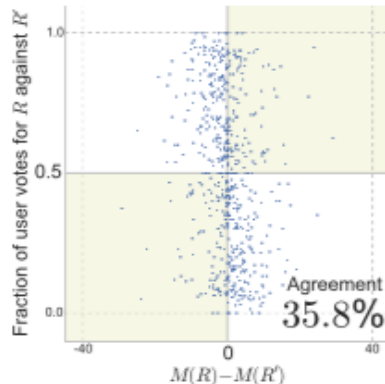
Agreement is high (70-74%) for both of the languages

Experimental Results for Simple Variants of M-measure

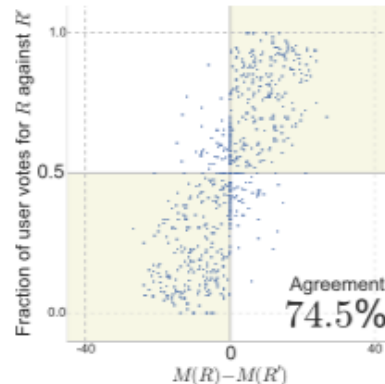
Original



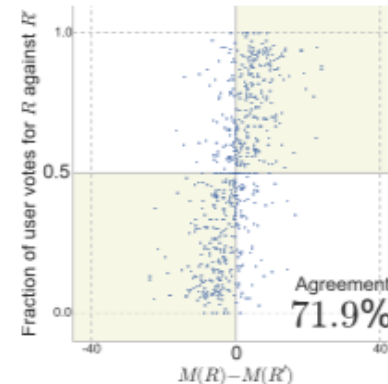
(a) $L = 24000$



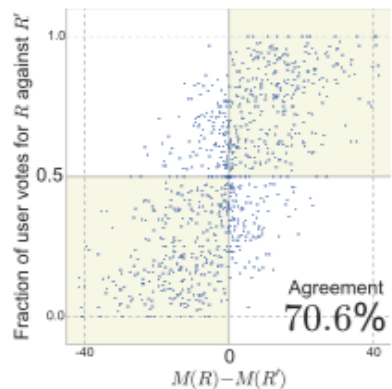
(b) Only first layer



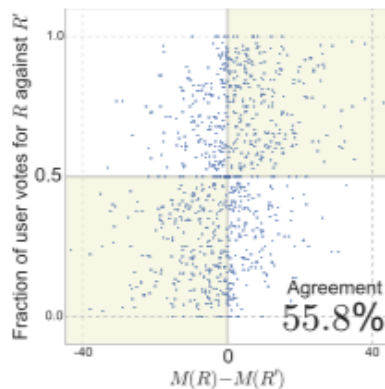
(c) Only second layer



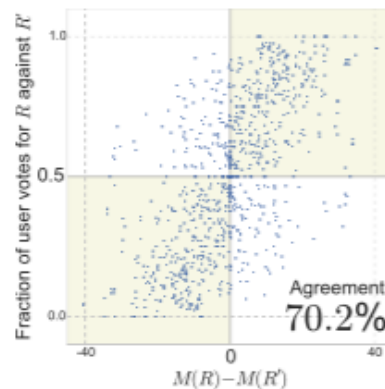
(d) Uniform $P(i|q)$



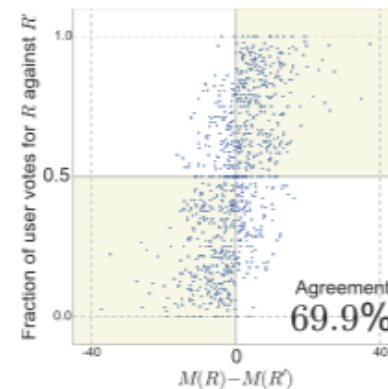
(e) $L = 2000$



(f) Only first layer



(g) Only second layer



(h) Uniform $P(i|q)$

Worse

Close

Slightly worse

Use of the second layer and intent probability improves the agreement (but the first layer doesn't)

Why did the only 2nd layer correlate to the user pref. well?

- **Possible explanations include**
 - The quality of the second layer correlates to the quality of the whole summary
 - Users decided the quality of the summary mainly based on the second layer
 - We asked the users to look at the second layer in the assessment

Conclusions and Future Work

- **Conclusions**

- **Proposed M-measure**

- A special case of intent-aware U-measure for two-layered summarization

- **Measured the agreement between M-measure and user preferences**

- Agreement was high (70-74%)

- **Future work**

- Error analysis

- Address “why did the only second layer correlate to the user preferences well?”