

Communicating Data Quality in On-Demand Curation

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Uncertain Data

Product	Rating Source			Note
	Buybeast	Amazeo	Targe	
Samesung	4.5	3.0		
Magnetbox	2.5		3.0	
Mapple		3.5	5.0	Not a TV?

The diagram highlights four types of data uncertainty with blue callout bubbles:

- Missing Data Points:** Points to the empty cells for Samesung in the Amazeo and Targe columns.
- Missing Trends:** Points to the empty cell for Magnetbox in the Amazeo column.
- Missing Information:** Points to the empty cell for Samesung in the Targe column.
- Missing Inference:** Points to the text "Not a TV?" in the Note column for Mapple.

What?

- ✓ Absence of Data in a column.
- ✓ Inability to get sufficient data points for analysis.

Product	Buybeast
Samesung	4.5
Magnetbox	2.5
Mapple	

Missing
Data Points



How to Solve this?

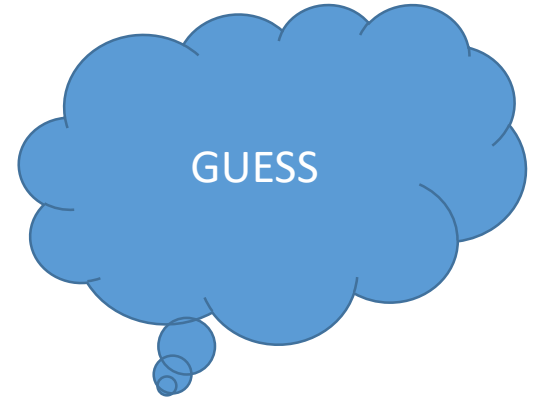
- ✓ Guess based on available data.
- ✓ Fill with a default value.
- ✓ Ignore the absent data - Delete

Magnetbox | 2.5 | | 3.0 |

Missing Trends

What?

- ✓ Absence of Data in a column.
- ✓ Inability to determine the increase or decrease in values.



How to Solve this?

- ✓ Guess based on available data.
- ✓ Fill with a statistically derived value.

Targe
3.0
5.0

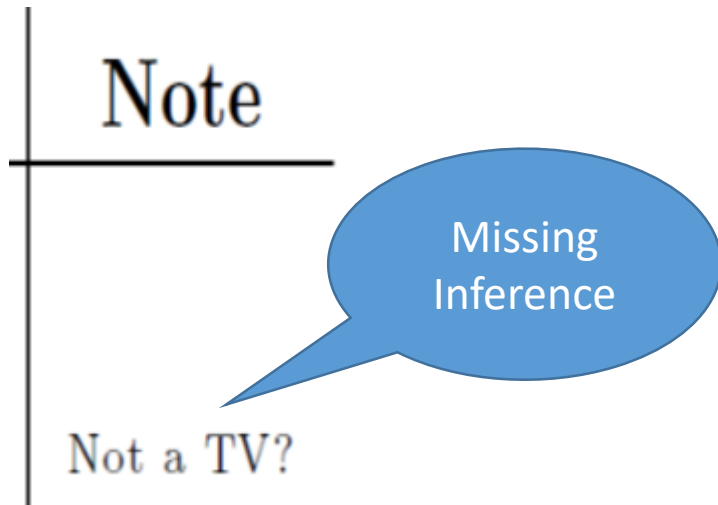
What?

- ✓ Absence of Data in a column.
- ✓ Inability to derive the range of values for a column.
- ✓ Inability to determine the if the data is continuous or discrete.



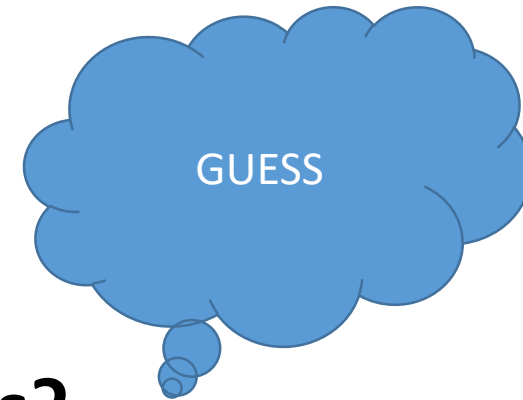
How to Solve this?

- ✓ Guess based on available data.
- ✓ Fill with a value based on the row and column data type.



What?

- ✓ Absence of Data in a column.
- ✓ Inability to make any conclusion about the data.



How to Solve this?

- ✓ Manually enter the conclusion.
- ✓ Use Machine Learning Models.

Types of uncertainty

Product	Rating Source			Note
	Buybeast	Amazeo	Targe	
Samesung	4.5	3.0		
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Attribute-level uncertainty

Attribute-level uncertainty

Row-level uncertainty

Classical Databases vs On Demand Data Curation

Classical Databases

- Erroneous data cannot be queried.

On Demand Data Curation

- Paygo, KATARA and Mimir
- Defer curation effort until necessary
- **Guesses or approximations as answers**
- **Quality and scope of guesses vary**
- **Communicate information to end user**

On Demand Data Curation

- Initial data is of low quality
- Queries liable to produce incomplete/incorrect result.
- Mitigate unreliability by providing a form of lineage
- Query tagging with quality metrics.
- ODC efforts are specialized form of probabilistic DB.
- Answers in form of certain data or probability distribution.

ODC represent low quality data in the form of probability distribution which is not understood by average database user.

What needs to be changed?

- Although we can fix the data using ODC tools.
- No matter how we fix it we are not guaranteed automated system would do a good job.
- Users uncomfortable with probability distribution.

Can we represent uncertain data in simpler form (user friendly)?

Long Term Goals

- The quality and scope of guesses may vary.
- How to communicate this information to an end-user.
- Set of UI design guidelines.
- Best practices for conveying uncertainty.

Goals of user study

- Preliminary user study
- Evaluate cognitive burden and expressiveness
- Focus on attribute level uncertainty
- Four ways of representing uncertainty

Primary Questions for the User Study

- Is the representation effective at communicating uncertainty?
- What is the cognitive burden of interpreting the representation?
- 14 participants, predominantly from CSE department at UB.

Experimental Setup

- Ranking task
- Web form with 3*3 matrix
- 3 products with ratings from 3 different website
- Participants presented with same set of information
- Multiple rounds with each round of five trials

User Interface

The table below gives ratings from different website for three products.

Name of Product	Rating 1	Rating 2	Rating 3
Product A	3	4.5	3.5
Product B	4	1.5	3.5
Product C	3	3.5	2.5

Task:

Please go through the details about the products and arrange the products in the order of your preference to buy them.

<p>Product A</p> <p>Product B</p> <p>Product C</p>		
--	--	--

Submit

Asterisk

Introduction:

The table below gives ratings from different website for three products.

Name of Product	Rating 1	Rating 2	Rating 3
Product A	4	1.5	5 *
Product B	4	2.5	3.5
Product C	1 *	2.5	3.5

Task:

Please go through the details about the products and arrange the products in the order of your preference to buy them.

Asterisk represents some error in data.

<p>Product A</p> <p>Product B</p> <p>Product C</p>		
--	--	--

Submit

Colored Text

Introduction:

The table below gives ratings from different website for three products.

Name of Product	Rating 1	Rating 2	Rating 3
Product A	3.5	4	2
Product B	4.5	2.5	3
Product C	4.5	4	5

Task:

Please go through the details about the products and arrange the products in the order of your preference to buy them.

red color represents error in data.

Product A		
Product B		
Product C		

Submit

Confidence Interval

Introduction:

The table below gives ratings from different website for three products.

Name of Product	Rating 1	Rating 2	Rating 3
Product A	1.5+/-0.5	4	1.5
Product B	2.5+/-0.5	2	2
Product C	1+/-0.5	5	5

Task:

Please go through the details about the products and arrange the products in the order of your preference to buy them.

Confidence intervals represent error in data.

<p>Product A</p> <p>Product B</p> <p>Product C</p>		
--	--	--

Submit

Color Box

Introduction:

The table below gives ratings from different website for three products.

Name of Product	Rating 1	Rating 2	Rating 3
Product A	4	1.5	5
Product B	3	2	4.5
Product C	3.5	2	2.5

Task:

Please go through the details about the products and arrange the products in the order of your preference to buy them.

red color represents error in data.

<p>Product A</p> <p>Product B</p> <p>Product C</p>		
--	--	--

Submit

Best Of 3

- Ratings were random with a bias towards a predictable ordering.
- Ratings generated using rejection sampling
- A had to have one extremely favorable rating compared to B (1 point higher)
- One slightly more favorable rating (0, 0.5, or 1 point higher)
- one slightly less favorable rating (0, 0.5, or 1 point lower).

With the preselected pattern we can detect change in user behavior.

Introducing Uncertainty

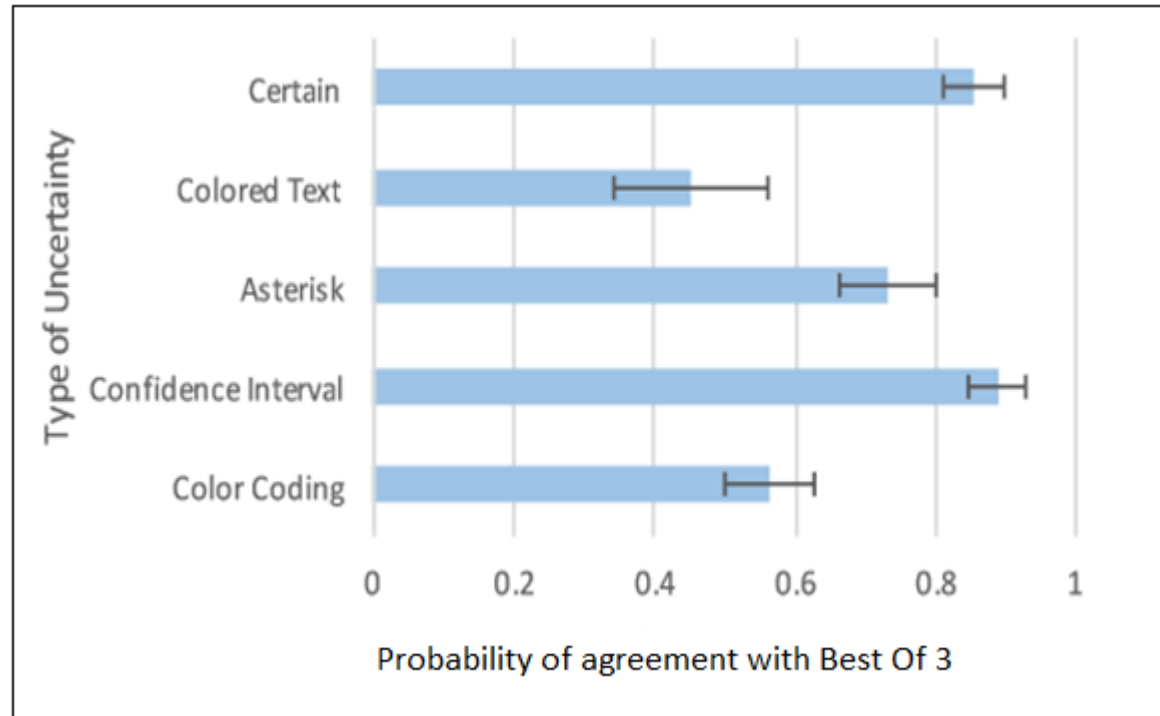
- In uncertain trials, base data generation followed Best Of 3
- Between 2 and 4 randomly chosen values were labeled as uncertain.
- Difference in user behavior due to certain values being uncertain

Rating generation process remained the same but the user were told that certain values were uncertain.

Compliance with best of 3 changes with uncertainty type

Effectiveness

Colored text and color coding altered participant behavior.



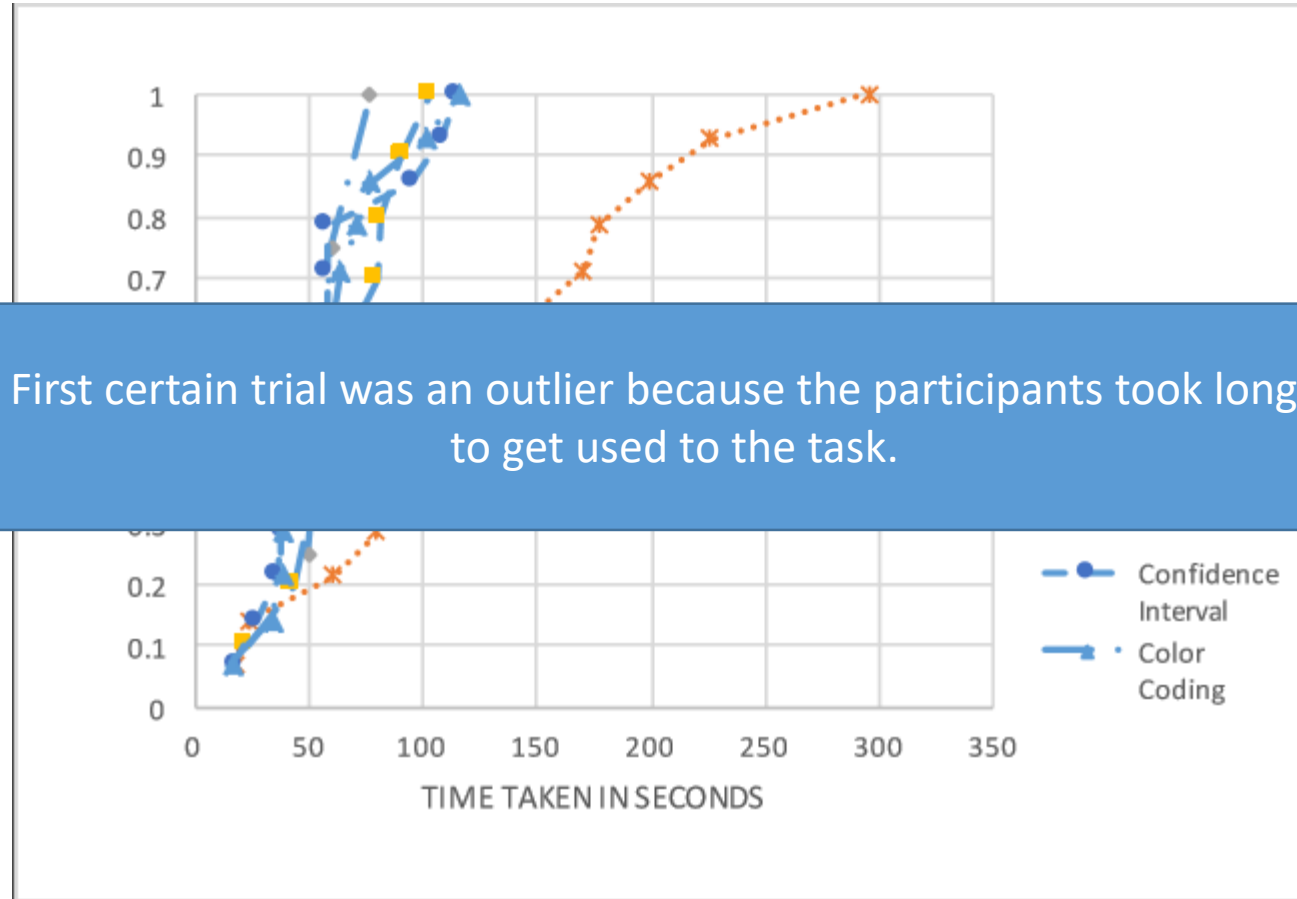
Experiment tests for changes in user behavior.

Certain and confidence interval show a consistent agreement with Best of 3.

Efficacy

First round, participants initially encounter the task and representation

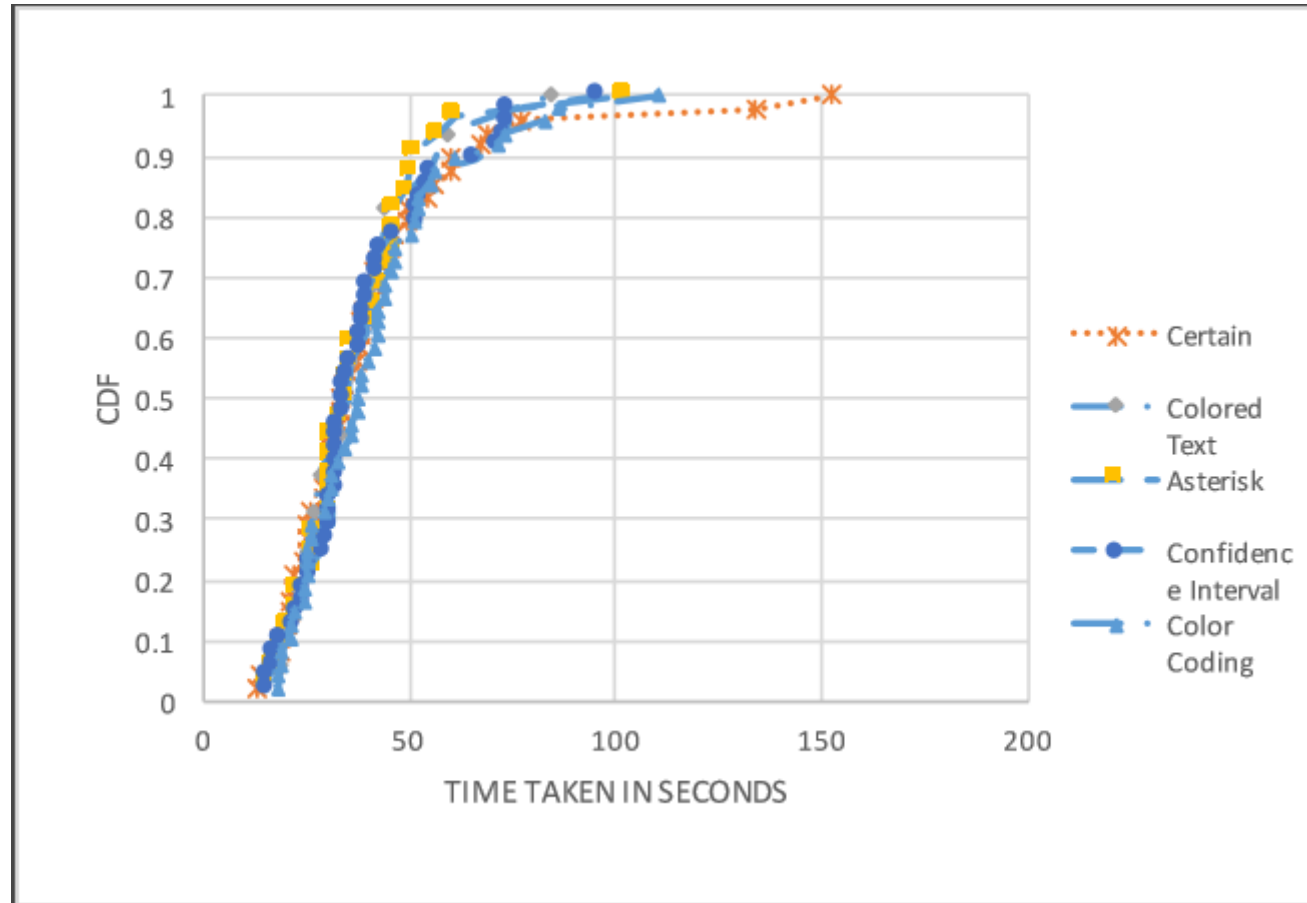
Time taken per representation was relatively consistent across all forms of uncertainty



Participants spent significantly more time familiarizing themselves with the overall ranking

Slower trial was deterministic

Efficacy



Results

- Is the representation effective at communicating uncertainty?
 - At least three distinct behavioral responses to uncertainty in the data were identified
 - suggesting differences in the efficacy of each representation.
- What is the cognitive burden of interpreting the representation?
 - All uncertainty representations required a similar amount of decision time
 - Impose similar cognitive burdens in the population under study
- Participants were predominantly from CS department.
- Participants conveyed a strong negative emotional reaction to the color coding representation.
- Several participants suggested feelings of comfort associated with the additional information that the confidence interval supplied.

Future Work

- Focus of this study was attribute level uncertainty
- Explore other types of uncertainty in relational data (row-level and open-world)
- qualitative feedback such as explanations
- giving the user mechanisms to dynamically control the level and complexity of uncertainty representation being shown
- incorporating our findings into the Mimir on-demand curation system

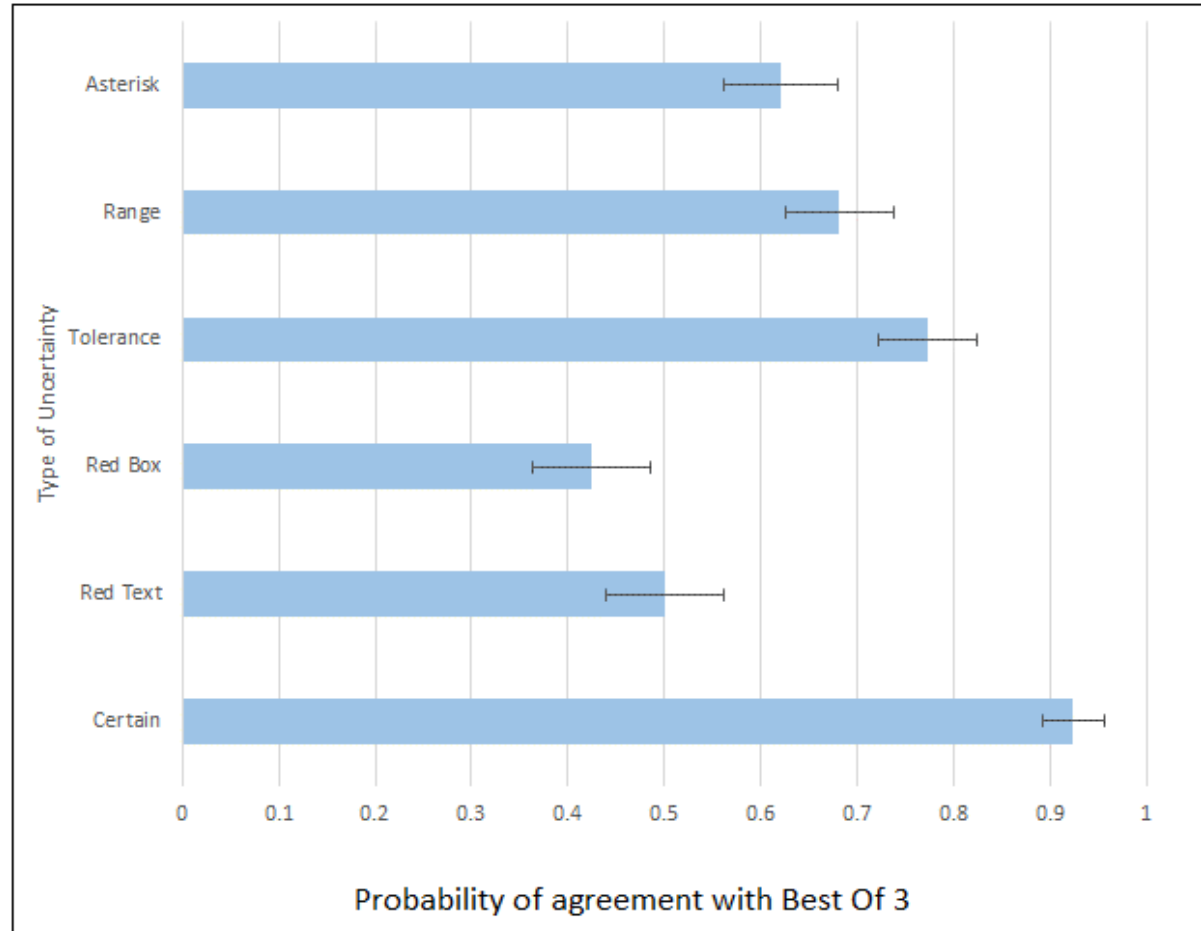
Changes made to Study 2

- Added two more representations of uncertainty.
- 3 rounds per user with 6 set of uncertainties.
- A mix of CS and non CS students were recruited as participants.
- Correlation between user characteristics and results analyzed.

Effectiveness

Certain trial still shows consistent agreement with Best Of 3

Red Box and Red Text were the most effective in altering participant behavior



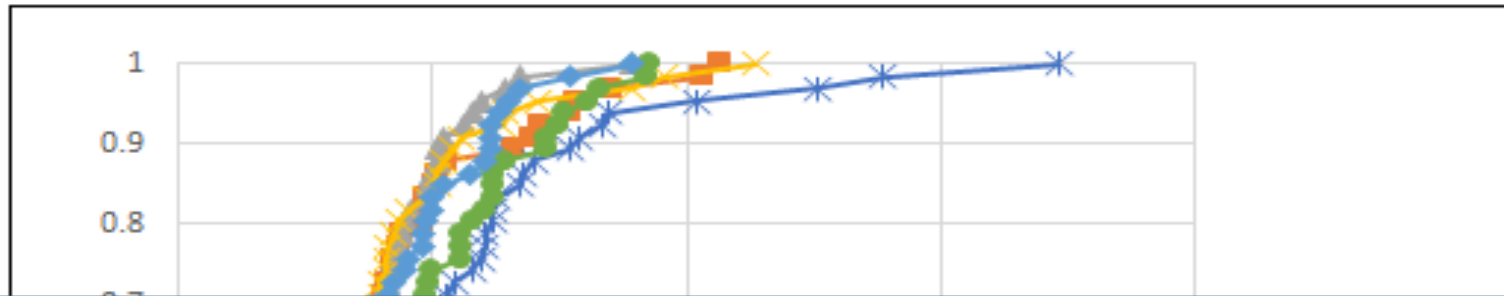
A dip was seen in agreement with Best Of 3 for Tolerance

Participants requested more information in asterisk trial

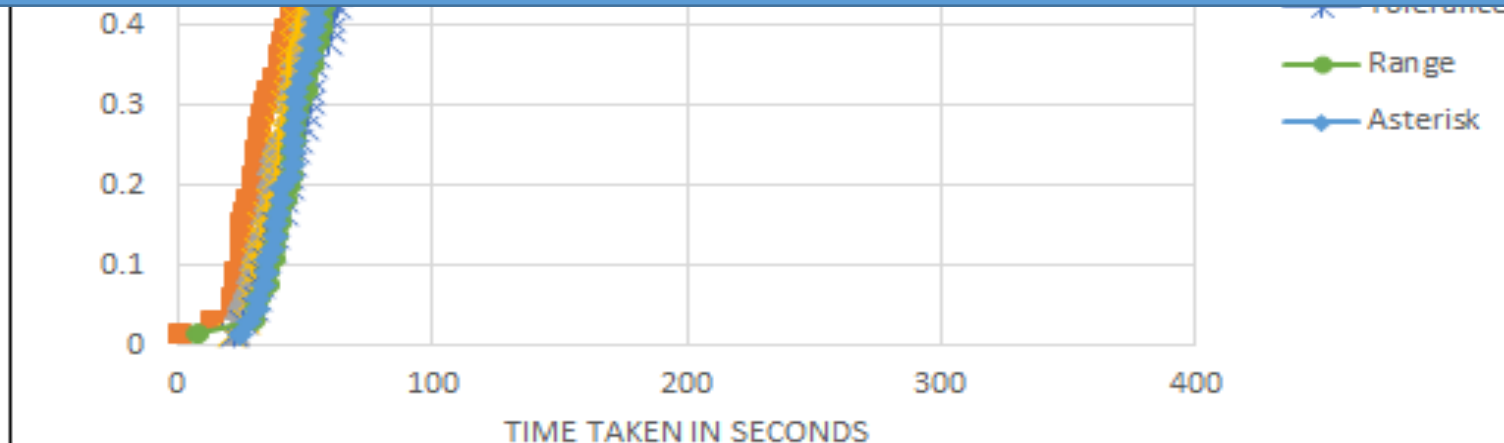
Efficacy

Participants were introduced to the task before the trial

Time taken per representation was relatively consistent across all forms of uncertainty



Non Computer Science students were not as comfortable with Tolerance when compared to Computer Science students.



Participants still spent more time on deterministic trial

Slower trial was Tolerance

Low quality data can be represented as a certain value by making a guess

We can represent low quality data in simpler form (user friendly)

Questions?

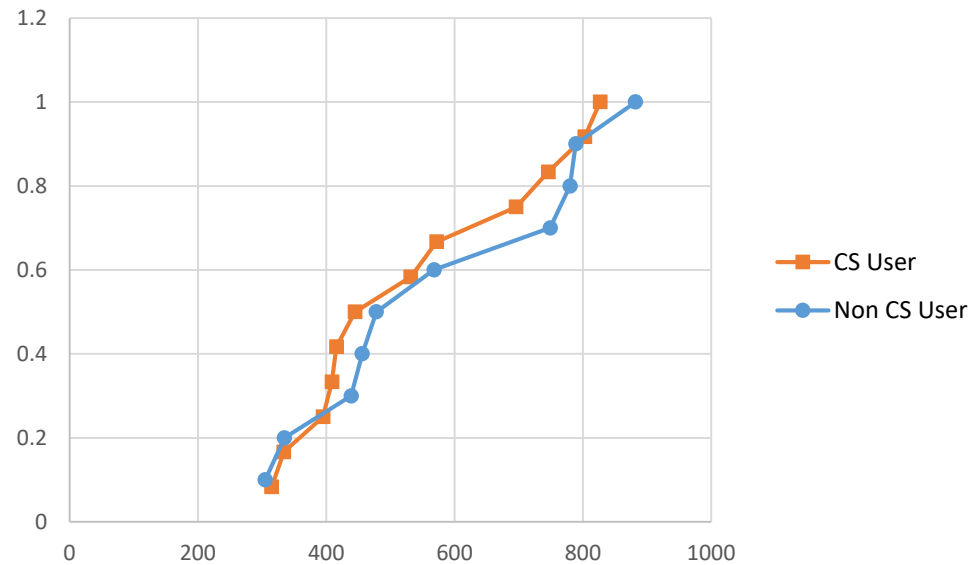
ODC tools make these guesses and represent low quality data as probability distribution

Time taken per representation was relatively consistent across all forms of uncertainty

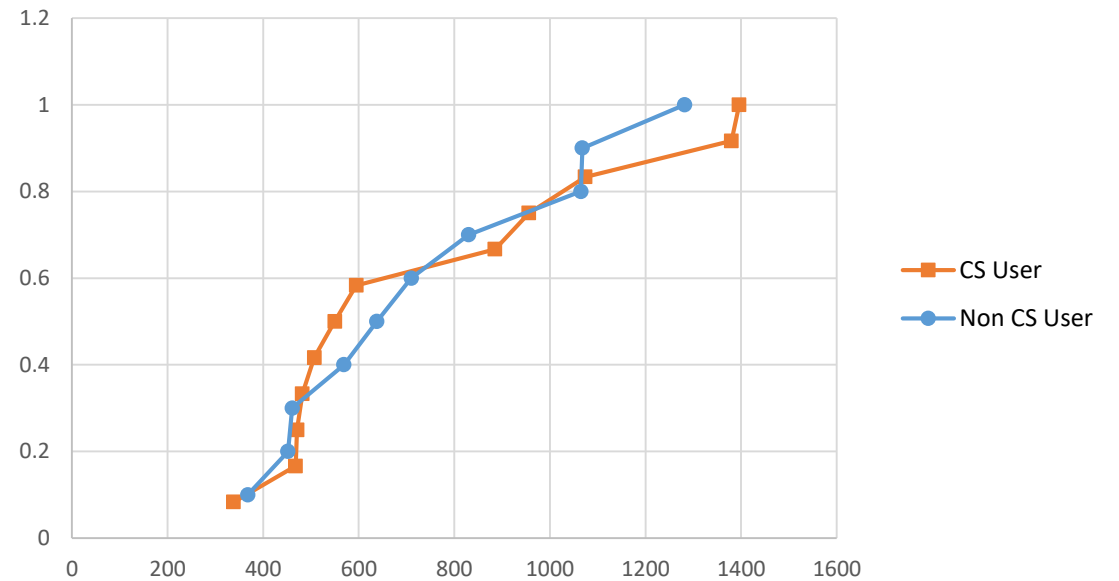
Thank You

Biases between CS and Non-CS students

Time taken per User 1st pass

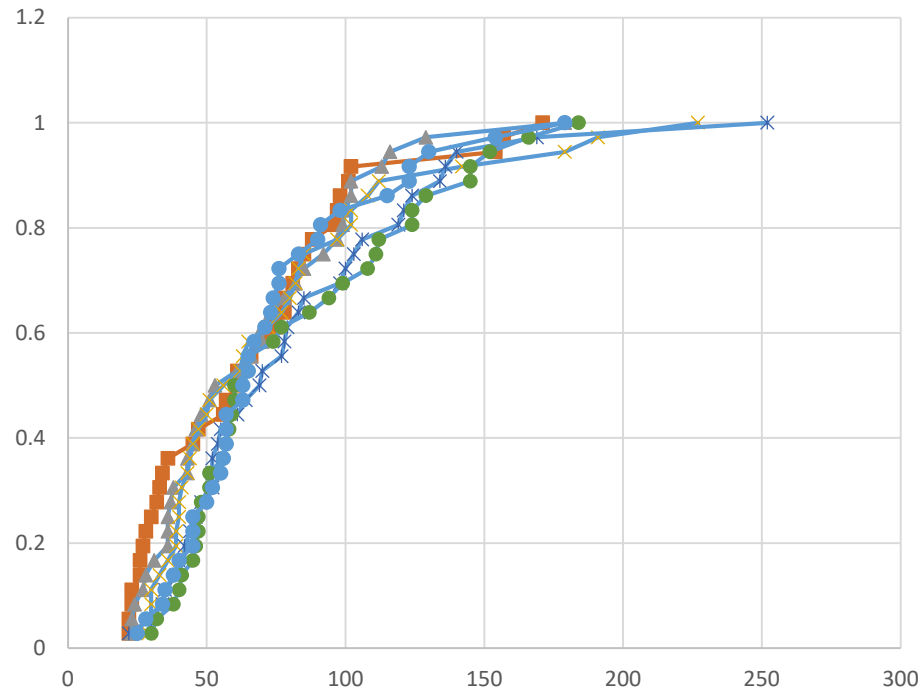


Time taken per User (2-3) pass

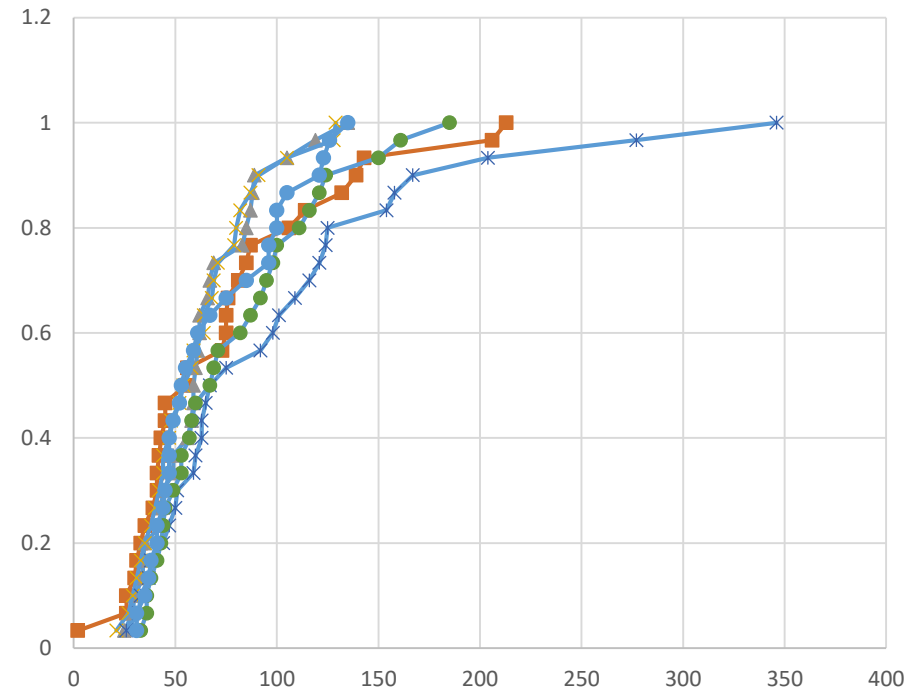


Time taken per uncertainty

Time per uncertainty (CS)



Time per uncertainty (NON-CS)



Different Regularities

