

Gathering Insights: Evaluation at trivago

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Agenda

- 1. About Me
- 2. About trivago
- 3. Content & Evaluations
- 4. C-Testing
- 5. User Research &
 - **Evaluations**
- 6. Summary

1. About Me

- Been active the field of NLG for ~14 years.
- 6 years at Aberdeen University.
 - 4 years PhD
 - 2 years PostDoc
- 5 years at Arria NLG.
- 4 years now at trivago.
 - Released 'Hotel Scribe' to generate automated descriptions of accommodations [Mahamood and Zembrzuski, 2019].
 - Currently lead a team of four data scientists.
 - Focus on a range of content based data science topics such as Image Tagging, Geospatial, Accommodation Metadata, and Data Quality problems.
 - Actively, participating in NLG research at trivago, including research projects. In particular, recently focused on the topic of evaluations.

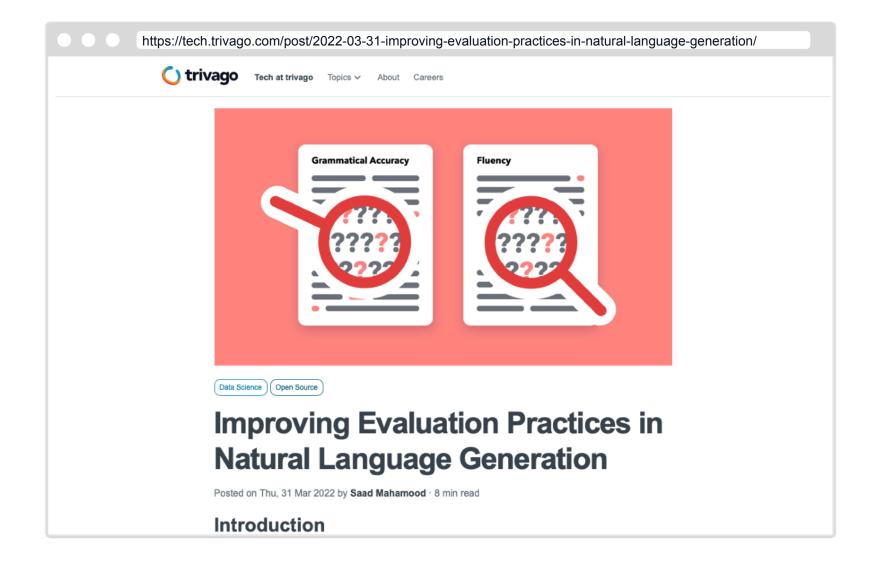
About Me

NLG Evaluation Work

- ~2010: First aware of issues in NLG evaluation, due to Reiter's & Belz's paper on investigating the validity of evaluations [Reiter & Belz, 2009]
- 2015: Worked with Dimitra Gkatzia on generating a snapshot of NLG evaluation practices for the last 10 years [<u>Gkatzia &</u> <u>Mahamood, 2015</u>].
- 2020: Collaborated with multiple researchers on an extensive 20-year overview of how NLG human evaluations are conducted [Howcroft et al., 2020].

NLG Evaluation Work

- 2021: Worked on several evaluation related research initiatives:
 - Explored Commonsense human NLG evaluations [Clinciu et al., 2021]
 - Underreporting of errors in NLG output
 [Miltenberg et al, 2021]
 - Reproducing an earlier NLG experiment
 [Mahamood, 2021]
 - Automatic construction of evaluation test sets [Mille et al, 2021]
- 2022: Active participation in the ReproHum project.



2. About trivago

Since being founded in 2005, trivago has become a leading global accommodation search website:

- Focus on helping millions of travellers to search for and compare hotels and other accommodations.
- Conducts business in 190 countries, with 5 million hotels and alternative accommodation, across 53 localised platforms in 31 languages.
- Based in Düsseldorf, Germany with around 800 employees.

About trivago

Enter Hotel or Destination



About Connectivity Insights

Connectivity Insights is one several Data Science teams within trivago...

- Team composed of four data scientists lead by myself.
- Focused on content related Data Science problems. In particular, we focus on:
 - **Images** e.g. Image tagging, image quality, image selection, etc.
 - Matching Accommodations from Partners and Destination Assignment
 - Accommodation Metadata, Descriptions, Reviews, and Ratings
 - **Pricing Data** e.g. Fetching prices, price components, etc.

3. Content & Evaluations

Content & Evaluations

Content based evaluations are mostly intrinsic evaluations...

- We seek to evaluate the performance of a given solution/approach by evaluating either against:
 - Internally collected Human annotated ground-truth datasets.
 - Requires extensive effort to align annotators to annotate consistently and correctly.
 - External ground-truth datasets provided by third-party providers.
 - Where possible try and use multiple providers and look for consensus.
 - Comparative evaluations against an existing solution.
 - "How does X compare to current Y."

Content & Evaluations

In addition to evaluations, the business impact of content related changes is also equally important...

- Impact can be measured in terms of:
 - The number of accommodations impacted with a given change.
 - The types of accommodations impacted.
 - Some accommodations are more important than others.
 - The countries or platforms impacted by a given change.
- Some examples of evaluations done for some recent projects:
 - Image Tagging
 - Accommodation Type Assignment
 - Destination Assignment

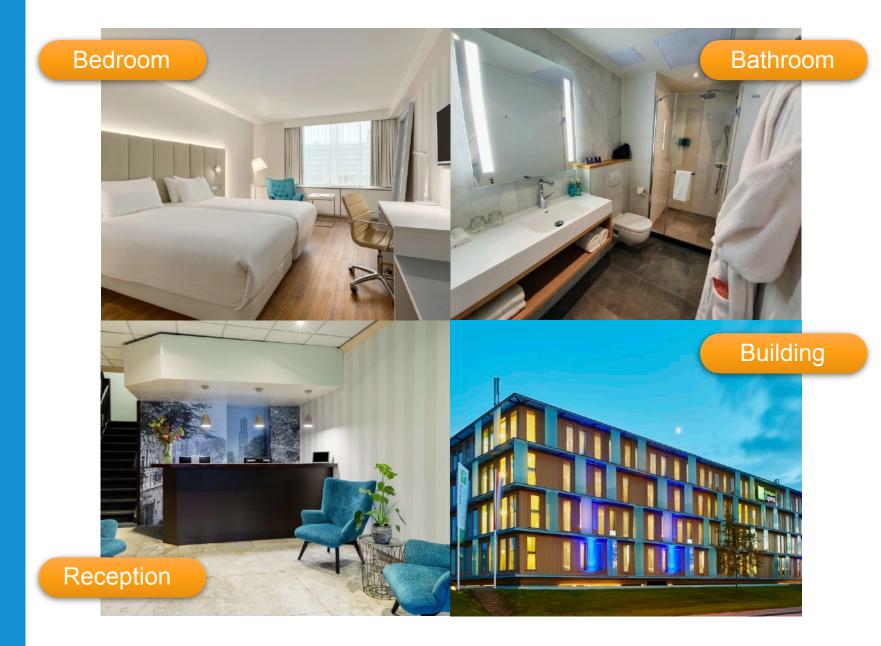


Image Tagging



Omayma Said

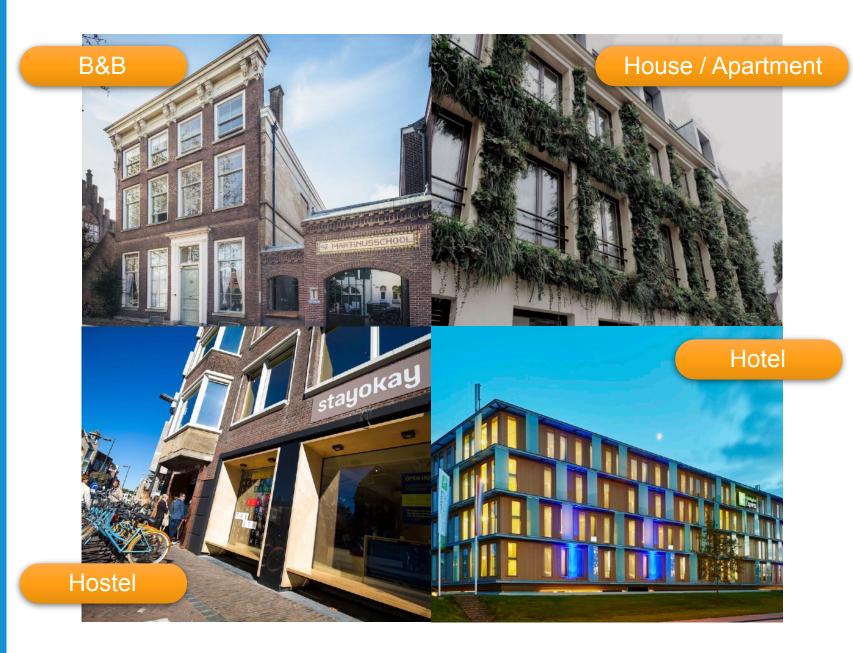
Danti Ramadanti

Image tagging allows us to give semantic "meaning" to images and enables downstream opportunities...

- Developed an in-house model after evaluating third-party image tagging models from an external provider.
- Evaluation is done by using a withheld human annotated dataset and computing precision and recall on a per tag basis.
 - In addition, we manually inspect the tagged images for correctness at random.
- Greatest challenge is not the evaluation, but the human dataset collection for training, testing, and validation:
 - Need robust and concrete definitions per tag.
 - Must extensively align annotators to tag images consistently and correctly.

Image Tagging







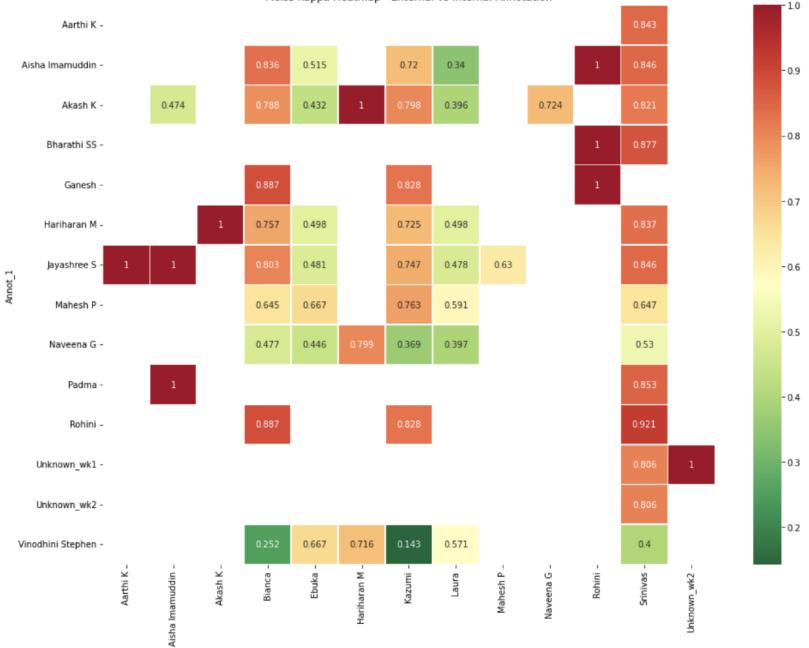
Srinivas Ramesh Kamath

Accommodation type assignment is the process of assigning a type for a given item...

- Normally, this data is provided by the partner who is sending the accommodation to trivago.
- However, in some cases this information is missing or differs from our definitions and we must try and infer the accommodation type.
- Developed a robust rule-based model to replace a non-functioning neural model for type assignment.
- Like image tagging, evaluation was done against high quality human annotations over 12,000 items.

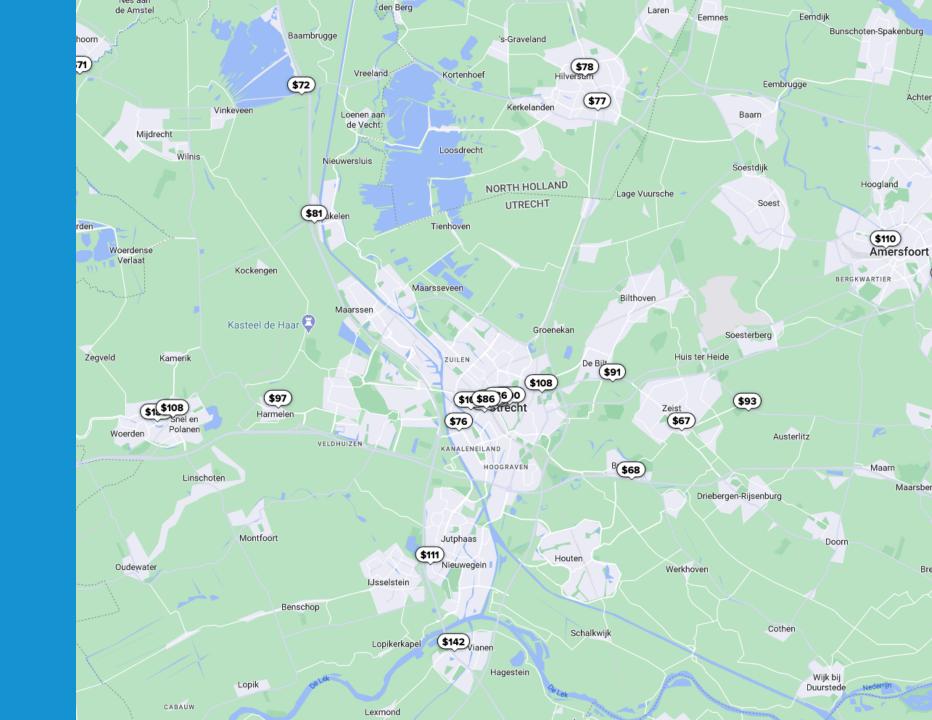
Accom. Type Assignment

Fleiss Kappa Heatmap - External vs Internal Annotation



Annot 2

Destination Assignment





Veronica Gonzalez Solano

Destination assignment is the process of assigning accommodation to a location such as a city, town, or attraction.

- Biggest challenges when performing destination assignment are the following:
 - Missing, incorrect, or misleading Geocodes.
 - Missing, incorrect, or contradictory addressing data.
- Therefore we need to not only locate where an accommodation should be, but then assign it a destination.
- However, this all depends on the correctness of our data and processes to perform the above.

Destination Assignment



Destination Assignment

We focused on evaluating the correctness of our destination ground truth data...

- In particular, check that we have correct data for where each destination is located w.r.t their geocodes.
- Ensure that geographical polygons have the right destination.
- Evaluation consists of:
 - Checking multiple data providers e.g. Google, OSM, TomTom, ArcGIS, etc.
 - Looking of either complete or majority consensus.
 - Using human annotators when there is **no consensus** between data providers.
 - Highly challenging as there is high degree of subjectivity involved.



At trivago we make extensive use of C-testing to perform multi-group testing...

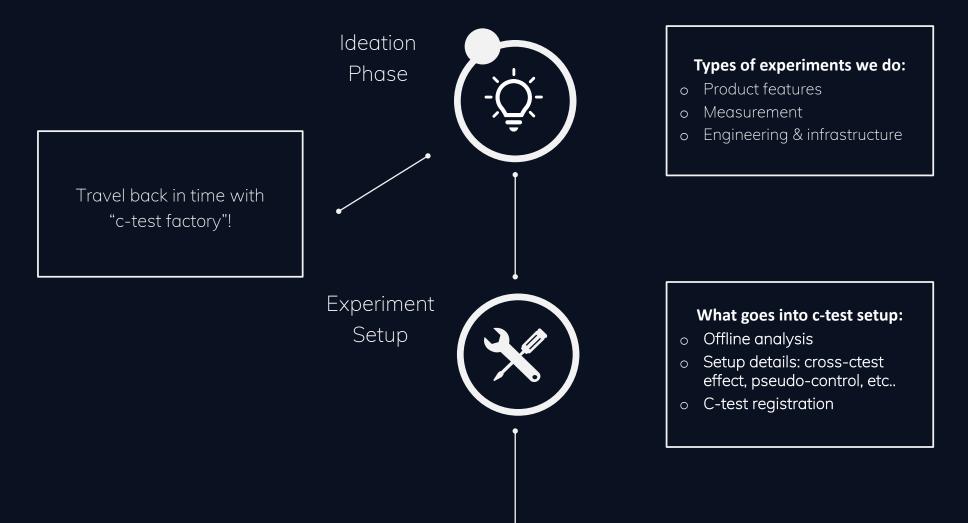
- We use C-testing to for evaluating product changes with our users.
 - Done both for front end UI/UX changes as well as backend changes as well (e.g. changes to the ranking model).
- C-testing is transparent to users and we can test specific groups such as by platform (web browser vs. Mobile Apps), locale, and by specific accommodations.
- Changes are significant if top line KPIs change more than 1% either way.
- C-tests are repeated to check if the result is reproducible.

C-Testing

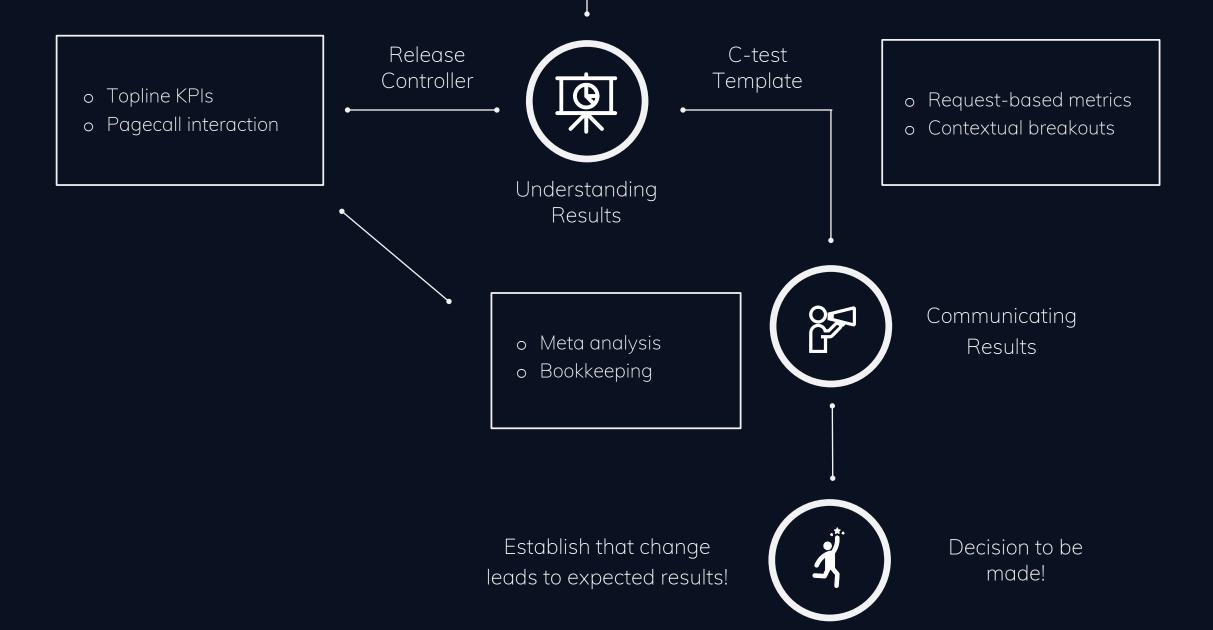
() trivago

The Art of Decision Making

C-testing lies at the heart of how we make decisions at trivago.



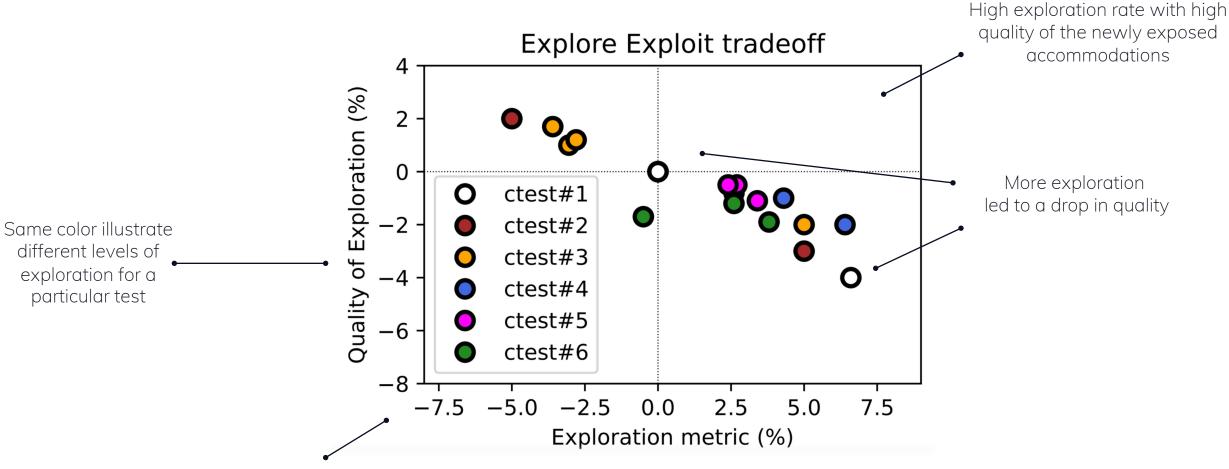
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Meta Analysis

In the context of ranking accommodation at trivago-

show the ones you know perform well or ones that have never been shown before?



Trade-off between extent and quality of exploration



Aida Orujov

https://tech.trivago.com/post/2022-11-04-explore-exploit-dilemma-in-ranking-model/ C trivago Tech at trivago Topics - About Careers **Explore-exploit** dilemma in **Ranking model** Data Science (Machine Learning) **Explore-exploit dilemma in Ranking model** Posted on Fri, 04 Nov 2022 by Aida Orujova · 9 min read **Problem Statement** Imagine, out of thousands of accommodations that match a user search, you have to select the "best" 25 to

5. User Research & Evaluations

User Research & Evaluation

User Research is our main approach of evaluating changes qualitatively with users

- We have several techniques for performing user research:
 - 1. Usability Testing
 - 2. Continuous Interviewing
 - 3. Diary Studies
 - Ask users to document their experiences over a period of time.
 - 4. User Interaction Recordings
 - Recorded for a small number of evaluation participants (1%).
 - Main interest is quantitatively seeing what users interact with.

5. Surveys

Usability testing is our way of evaluating designs and getting feedback from users...

- We recruit participants based on set of key characteristics: age, gender, employment status, country, web expertise, etc.
- Users are screened before testing using a questionnaire for suitability.
- Participants are given a scenario and a set list of tasks that they must accomplish. Their screens are recorded during the evaluation.
- The recorded is reviewed and we seek to assess:
 - User feedback
 - Task accomplishment success
 - Time taken by the user
 - User confidence in completing the tasks
- Results are aggregated over many users for a given experiment to produce a set of usability findings.

Usability Testing

Results:

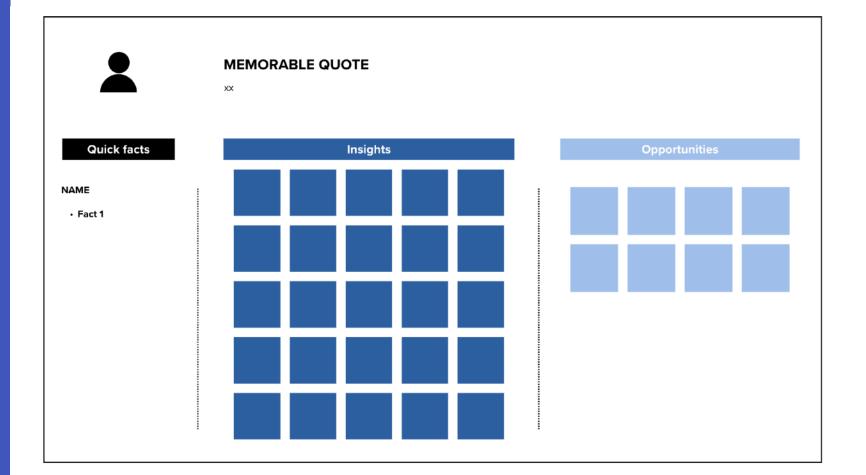
Jsability issue	P1	P2	P3	P4	P5
Description of usability finding 1	H		Μ	L	
Description of usability finding 2		Μ	Μ		
			L		
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Usability Testing

Continuous Interviewing Another source feedback is through user interviews...

- We perform interviews with different cohort of users on a continual ongoing basis.
- Interviews are half-hour with a fixed interview script.
- Attempt to understand user needs and their pain points.
- Collate from the interviews recurrent themes and translate these themes into product action points.

Continuous Interviewing



6. Summary

Evaluation within trivago is multi-faceted...

- Intrinsic quality evaluations using both human or third-party sources of ground-truth for our content based work.
- Extrinsic user-based evaluations using our Ctesting framework to perform multi-group testing.
- Qualitative user-centric evaluations using methods such as interviews, diary studies, usability testing, etc.
- Key challenge with intrinsic evaluations is to create or find sources of truth to evaluate against.
- Evaluation result significance depends on business metrics or objectives.



