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Practical A/B Testing

Creating Experimentation-Driven Products

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ISBN-13: 979-8-88865-008-0 Encoded using the finest acid-free high-entropy binary digits. Book version: P1.0—May 2023 Idea generation is easy. We could brainstorm new product ideas all day. The hard work comes when validating a new idea improves the product. A/B testing can help validate new ideas so that we're improving the product for both the users and the business. To do this, we need data to demonstrate the impact of an A/B test.

You may have noticed an important detail yet to be discussed, which is the complexities of gathering, querying, and visualizing data for your experiments while also ensuring your test configuration is valid.

The previous chapter focused on the different types of experiments. We dabbled in non-inferiority tests that seek to measure if a change is as good as the control, given a predefined margin. We learned the logistics for evaluating changes on a longer time scale by leveraging long-term holdbacks for a degradation or cumulative holdback test.

Regardless of which type of A/B test is selected, you'll need access to quality data to analyze and adequately visualize your test results. How you interface with data to create your A/B test results directly influences the A/B testing process. In this chapter, we'll explore the following:

- Common practices for storing and accessing data.
- Different types of data to create product insights and monitor the performance of an A/B test.
- Various visualizations to present the results of an A/B test.

Let's tackle the more complex and data-intensive aspects of A/B testing. Here we go!

Working with Different Types of Data

The goal of A/B testing is to observe what happens as you introduce changes to the product. To do this, you need data, a lot of data. You also need different types of data. Having different categories of data will help you with the following:

- Enable you to cross-reference and validate outcomes to increase confidence in the analysis.
- Give you the capability to monitor your engineering system health that creates the product experience.
- Provide you with a deeper, richer context for data analysis.

Let's look at the different categories of data that will create a deeper understanding of how your changes impact the product.

Defining User Engagement Data

To understand what users are up to on the product, you'll need data representing their actions and engagement. The typical path to get user engagement data is for the front end, or client side, to log events. The events will initially be raw and maybe a bit messy. When aggregated and enriched, these events can be used to understand the user's journey as they engage with the product. Examples of user engagement events include the following:

- Impressions
- Clicks
- Pageviews

When you have user engagement events such as impressions, the work to create the events is referred to as instrumentation. Let's emphasize the importance of instrumentation. To run an experiment, you'll need instrumentation within the product to create the raw events, or logs, representing how users engage with product features. Instrumentation is a crucial mechanism that will enable you to understand what the users see on the product and then do as a result of seeing a change or new feature. If you're unsure what your product should instrument, think about the actions a user can take on the product. Can a user click or scroll? How long does it take them to scroll or click something they recently saw? Implementing instrumentation will give you a richer understanding of the effect of changes introduced in an experiment.

Remember that these data points alone are usually meaningless as a unit by themselves. Any data, regardless of the type, is only valuable if you can understand it. Therefore, enriching and aggregating it to improve interpretation is key. Let's look at the data architecture at CableMax to illustrate how user engagement data can be combined with additional contexts to create a richer understanding.