# Estimating incrementality over time

## Notation

- c: Cohort of shoppers; c = 0 for "control" cohort that do NOT experience Fenix, and c = 1 for the "treatment" cohort that experience Fenix products.
- G: Set of mutually exclusive and collectively exhaustive subgroups of shoppers (e.g.,  $G = \{$  Male in CA, Female in CA, Everyone else  $\}$  would be a valid set of three subgroups)
- $\theta_{c,g}$ : Conversion rate, defined as the proportion of unique sessions in cohort c and group g that result in an order being placed. The subscript for g may be omitted to mean the entire cohort c (e.g.,  $\theta_1$  represents the conversion rate for all sessions in the treatment group and  $\theta_{1,g}$  represents the conversion rate for sessions in treatment group that meet the criteria for group g)
- $\delta_g = \theta_{1,g} \theta_{0,g}$ : The causal effect of Fenix product on the conversion rate of group g

The subscript g may be omitted to mean the entire cohort (e.g.,  $\theta_1$  represents the conversion rate for all sessions in the treatment group, and  $\delta$  represents the average treatment effect over the entire population.)

## Problem statement

Given at least one prior experimental measure of  $\delta$ , and a new set of non-experimental observations under treatment (c = 1), we wish to extrapolate estimates of  $\delta$  to the new observations.

#### Setup

- We have some prior experimental measure of  $\widehat{\delta_g}$  for every  $g \in G$  .<sup>1</sup>
- Once Fenix is "live", without an explicit holdout, for some specific period, we observe
  - $\circ n_g$ : Total number of unique sessions in some group g
  - $\circ o_{g}$ : Total number of orders from sessions in group g
  - $\circ \quad \theta_{1,g} = \frac{o_g}{n_g} \text{(i.e., every session is considered "treatment")}$
  - $N = \sum_{g \in G} n_g$ : Total number of unique sessions

<sup>&</sup>lt;sup>1</sup> Note, this requires that, for each unique session, we have group identifiers (e.g., "male/female", location info regarding the session) *prior* to the order.

#### Proposal (with concrete grouping data at session-level)

• We wish to know the counterfactual conversion rate:

$$\theta_0 = \sum_{g \in G} (\theta_{1,g} - \widehat{\delta_g}) \frac{n_g}{N}$$

- Assume the treatment effect on any given <u>group</u> does NOT change over time (e.g., the conversion rates for "Men in California" might change, but the *difference* of what conversion rate might have been with/without Fenix is always the same for that group)
- What <u>does</u> change is the conversion rate itself and composition of shoppers (e.g., more "Men in California" are visiting, and the conversion rate for that group is dropping)
- Then, given all the information we already have and observe, we can readily compute  $\theta_0$
- The main challenge is in collecting the data necessary for assigning sessions to a group *prior* or orders, both for the experiment and post-experiment.

### [WIP] Proposal (WITHOUT concrete grouping data at session-level)

- Now assume that we *cannot* identify the group of a session prior to an order being placed
- This will have implications on how we measure  $\widehat{\delta_g}$  to begin with; but let's start by assuming we have those available from a previous experiment, as before.
- Then, we no longer observe  $n_a$ , and only observe  $o_a$
- As a result, we also do not observe the subgroup conversion rate  $\theta_{1,g} = \frac{o_g}{n}$

Rewriting the expression for  $\theta_0$ , with substitution, we can arrive at

$$\begin{split} \theta_{-}0 &= \sum_{g \in G}^{\{\}} \vdash ( [\{o_{-}g - n_{-}g^{\{}\delta\}_{-}g\} \{n_{-}g\} \vdash ) [\{n_{-}g\} \{N\}] \\ \theta_{-}0 &= N^{\{-1\}} \sum_{g \in G}^{\{\}} \vdash (o_{-}g - n_{-}g^{\{}\delta\}_{-}g \vdash ) \\ \theta_{-}0 &= N^{\{-1\}} \vdash (o - \sum_{g \in G}^{\{\}} \{n_{-}g^{\{}\delta\}_{-}g \vdash ) \end{split}$$

In other words, the problem is reduced to estimating  $n_g$ : the number of sessions that should be assigned to each group.

#### Notes regarding implications on initial measurement

If we don't have session-level grouping data, we wouldn't be able to identify group-level conversion rate because we can't know the total number of sessions for each group. What we *can* know is the total number of *orders* for each group, from which we will have to

extrapolate the proportion of each group in sessions in such a way that control and treatment agree. [TBD].