Decision Platform for Pattern Discovery and Causal Effect Estimation in Contraceptive Discontinuation

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Abstract

Contraceptive use improves the health of women and children in several ways, yet data shows high rates of discontinuation which is not well understood. We introduce an AI-based decision platform capable of analyzing event data to identify patterns of contraceptive uptake that are unique to a subpopulation of interest. These discriminatory patterns provide valuable, interpretable insights to policymakers. The sequences then serve as a hypothesis for downstream causal analysis to estimate the effect of specific variables on discontinuation outcomes. Our platform presents a way to visualize, stratify, compare, and perform a causal analysis on covariates that determine contraceptive uptake behavior, and yet is general enough to be extended to a variety of applications.

1 Study of Contraceptive Discontinuation

Family Planning (FP) has emerged as a crucial component of sustainable global development [Osotimehin, 2015]. Effective use of contraceptives can significantly improve the nutritional status of both mother and children, reduce maternal mortality risks, and improve child survival [Ahmed et al., 2019; Larsson and Stanfors, 2014; Azuike et al., 2017; Johnson, 2017; Cahill et al., 2018; Dey, 2019; Michael and Scent, 2017].

Surprisingly, complex pattern behaviours of contraceptive use and discontinuation\textsuperscript{1} remain less explored. This work was done in partnership with the FP group at Bill & Melinda Gates Foundation (BMGF)\textsuperscript{2}. We explore three questions specifically: (Q1) What do women transition to when they discontinue or switch between contraceptive methods? (Q2) Are there any recurrent sequences\textsuperscript{3} of contraceptive use and discontinuation across countries? (Q3) Can we go beyond covariate analysis to establish a causal effect of particular contraceptives for a specific discontinuation reason?

Subject matter experts need the ability to look at large amounts of survey data (spanning multiple years, tens of thousands of subjects, across several countries), and be able to answer the above questions. Further, they need the ability to use this data to guide the design of policies and interventions e.g.: why do women discontinue injectables at a higher rate compared to other methods, and what if we designed better educational policies; a problem well suited for Machine Learning (ML) methods.

In contrast to previous works that look at traditional statistical methods [Michael and Scent, 2017; Johnson, 2017; Cahill et al., 2018; Dey, 2019], we use ML methods to provide novel, complex insights into contraceptive discontinuation across several countries. This technique enables us to systematically verify the hypothesis generated in the earlier step.

2 The Decision Platform for FP

The Decision Platform for FP was designed to explore contraceptive use in countries with available DHS Surveys and Calendar data information. The platform provides the following facilities:

- **Automatically characterize subgroups of interest using patterns** that are unique to a subgroup (i.e. discriminatory), providing the ability to automatically generate hypotheses for various subgroups.
- **Estimate the causal effect** of using a particular contraceptive method on discontinuation across several countries. This technique enables us to systematically verify the hypothesis generated in the earlier step.
- **Integrated dashboard with visualization techniques** to make these ML insights accessible to a domain expert to explore discontinuation and contraceptive use.

Figure 1 depicts our prototype system and a brief description of each module is detailed below.

2.1 Pre-loaded Datasets in the Decision Platform for FP

In our particular case, we used the DHS data in consultation with our partner, team of Family Planning experts within the BMGF. The DHS calendar dataset [Croft et al., 2018] is a month by month history of certain key events in the life of the respondent for the calendar period preceding the date of

\textsuperscript{1}A discontinuation episode refers to an event that occurs in a specific month when a contraceptive method is not used for any reason.

\textsuperscript{2}https://www.gatesfoundation.org/

\textsuperscript{3}A sequence is an ordered list of items, in our case, a series of monthly ordered episodes from calendar data.
2.2 Hypothesis Generator

This module performs pattern mining on calendar data to check for common patterns in subjects that discontinue their contraceptive uptake. These patterns serve as hypothesis for downstream causal analysis to determine if and to what extent there exists causal relationship to discontinuation.

To analyze the sequential calendar data, we leverage sequence mining techniques, namely PrefixSpan [Jian Pei et al., 2004]. However, given the size of the data, the unique journey of women in their fertility and contraceptive choices, out-of-the-box PrefixSpan mines very large numbers of frequent sequences that are often not unique to the subgroup of interest and hence largely unusable for manual inspection by an analyst. Thus, we extend the PrefixSpan algorithm to analyze large-scale and noisy data. We then compare two subgroups and mine patterns that would only appear predominantly in one of the subgroups. We collectively call these extensions the Discriminatory Subsequence Mining (DSM) technique. After the DSM model is trained, we extract the \( n \) discriminatory subsequences for each country \( C_j \), we keep the top \( k \) more discriminatory sequences, based on lift metrics. An example of a discriminatory sequence found in DHS Nigeria 2013 was \([\text{Injectables} \rightarrow \text{Injectables} \rightarrow \text{Non-use} \rightarrow \text{Non-use}]\) under the sub-population that discontinue due to health concerns, this pattern appears in 34% of that subgroup and is 663 times more likely to appear in it than the rest of the population.

2.3 Causal Effect Estimator

This module uses both information from the subsequences found by the Hypothesis Generator and DHS Survey Data. The \( H_0 \), treatment and outcome \( (R_i) \) are extracted from the discriminatory subsequence. We conduct a series of steps from the observational data obtained from each individual woman’s records of the DHS survey data. We want to evaluate if use of a particular contraceptive method had a causal effect on discontinuation for health concern reasons. The outcome is defined as whether or not a woman surveyed in the DHS reported “discontinuation for reason \( R_i \)”. Third, we create the treatment assignment based on the use of that contraceptive in the past year. With that problem setup, we use Inverse Probability Weighting (IPW) [Rosenbaum and Rubin, 1983] with stabilized weights to calculate weighted population outcome for each subgroup stratified by treatment assignment. Finally, an example of a finding generated by the FP platform is that the use of injectables was 5 times more likely to cause discontinuation for health concerns for women surveyed in Kenya in 2014 compared to women surveyed in Ethiopia in 2016.
2.4 Decision Dashboard

The decision dashboard is updated with information from both previous modules. For example, we use interactive violin plots to display consecutive months per contraceptive type and country (multiple selections for comparisons are available). Furthermore, to easily interact with contraceptive switching patterns we implemented Sankey plots where the source and target can be set by the domain expert. For sequence mining insights, we use a combination of directed graphs and bar plots (for support metrics). For causal analysis, we display average treatment effect and several evaluation metrics of the Estimator, such as violation of positivity, consistency, balanced covariates and accuracy of our propensity model.

3 Concluding Remarks

The Decision Platform for FP provides automatic sequence mining and causal analysis for DHS Survey and Contraceptive Calendar Data. We demonstrate in this work the potential to provide decision support systems for policymakers regarding contraceptive use and discontinuation. At its core, our work is about analyzing sequential event data, extracting patterns that might help identify associations between behavioral patterns and discontinuation, and identifying if there is causal link to such associations.

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References


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