

Optimizing Targeting Strategy for Services

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IMPACT

Capability to **personalize service offerings**, improving customer experience & leading to **multi-million dollar uplift**



Scalable, holistic framework that can be expanded to **additional services and tools**, as well as to **new customer touchpoints**



Methodology to **inform business decisions**, enabling **better customer targeting and identification of areas of opportunity for services**

CONTEXT

In addition to offering "A Zillion Things Home," Wayfair offers a range of **services** aimed at alleviating stresses of the home-shopping experience.

There is an underlying belief that these services provide **positive experiences and increased customer loyalty**. This increase in loyalty generates additional value for Wayfair through **incremental long-term revenue**.

Currently, service impressions are not personalized. Presenting irrelevant services leads to increased **cognitive burden** on the customer and increased **page load times**, harming the customer experience and possibly leading to lower **conversion rates**.

Curating the services Wayfair exposes to its customers is important, and this curation should be informed by accurately modeling customers' **estimated incrementality and likelihood of signing up** for services.

CENTRAL QUESTION

For each Wayfair customer session, which **service(s)** should be presented at **which point** of the online shopping experience to **maximize the net present value (NPV)** of engagement?



GOALS

- To develop an analytical approach for estimating service NPV and sign up propensity, at an individual level, to determine which services to display at different points in the session experience
- To inform business decision making based on insights obtained from interpretable models

KEY DEFINITIONS

- Net Present Value (NPV):** Immediate Revenue Generated + Incrementality – Service Fulfillment Cost
- Incrementality:** Additional revenue generated due to a positive experience with a Wayfair service
- Service Impression Action:** An action on the part of Wayfair to show a service, or combination of services, to a customer on a particular website page

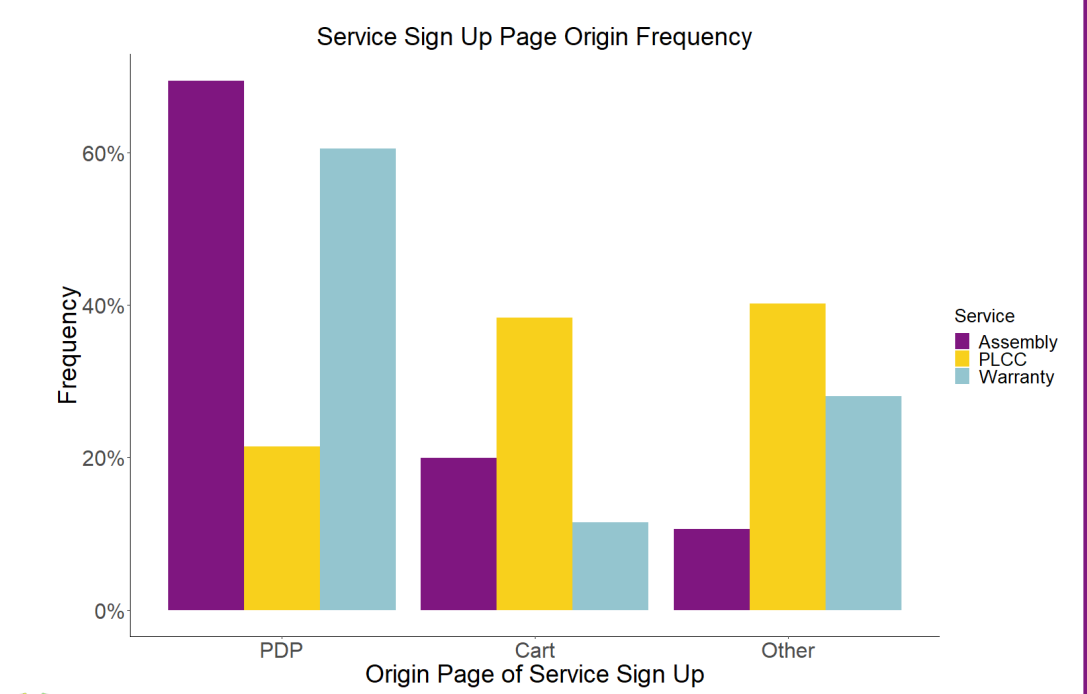
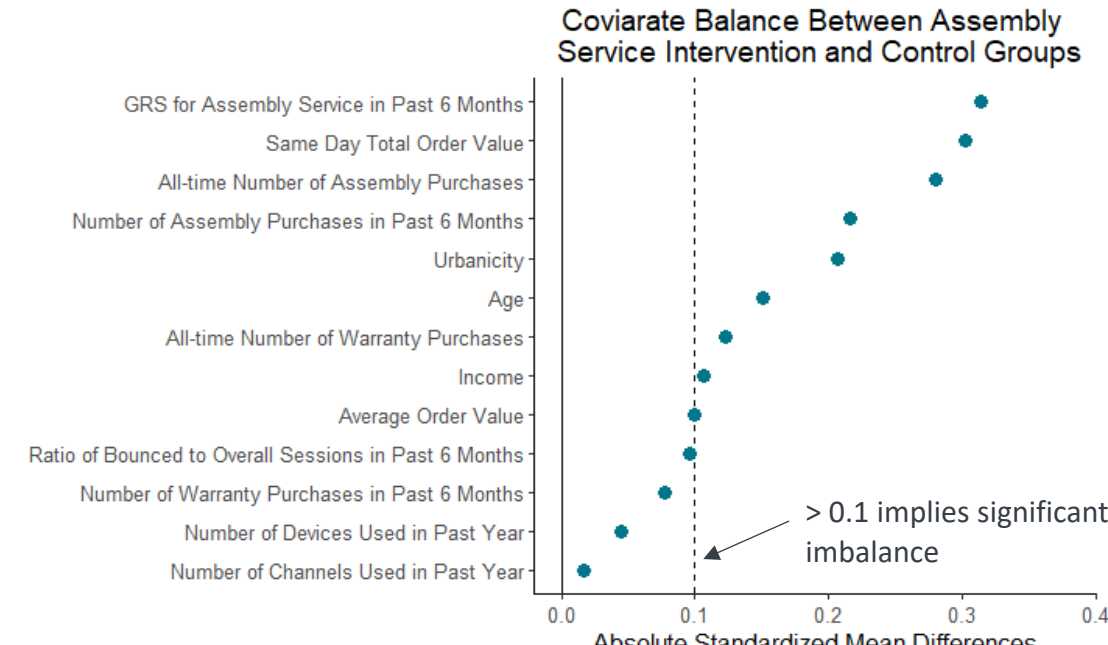
DATASETS

Scope: B2C Customers	3 Services	3 Modeling Components	> 1M Customer Sessions	> 100 Features
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- Past Website Interactions:** Devices & Platforms Used, Marketing Acquisition Channels
- Past Purchase Behavior:** Price Sensitivity, Average Order Value, Number of Past Orders, Gross Revenue Stable at Specified Time Periods
- Engagement Metrics:** Frequency of Website Visits, Ratio of Bounced Sessions to All Sessions, Idea Board Usage, Internal Search Usage, Average Session Duration

- Static Session-Level Features:** Marketing Channel Leading to Session, Marketing Visitor Type, Days Since Last Visit, Days Since Last Order
- Dynamic Session-Level Features:** Current SKU Display Price, Current SKU Service Prices, Number of PDPs Visited, Running Add-to-Cart & Remove-from-Cart Counts

Pre-intervention attributes differ between customers in the intervention group vs. the control group. → Our NPV models must account for potential bias induced by these differences.



Across all services, a large percentage of service sign ups originate at the product display page (PDP). → We focus our propensity modeling on the PDP and, consequently, develop service impression recommendations at each PDP viewed during a session.

NPV ESTIMATION

Key Question: How much more will a customer spend in the future as a result of engaging with a service (the intervention), relative to what they would have spent otherwise?

Key Modeling Considerations

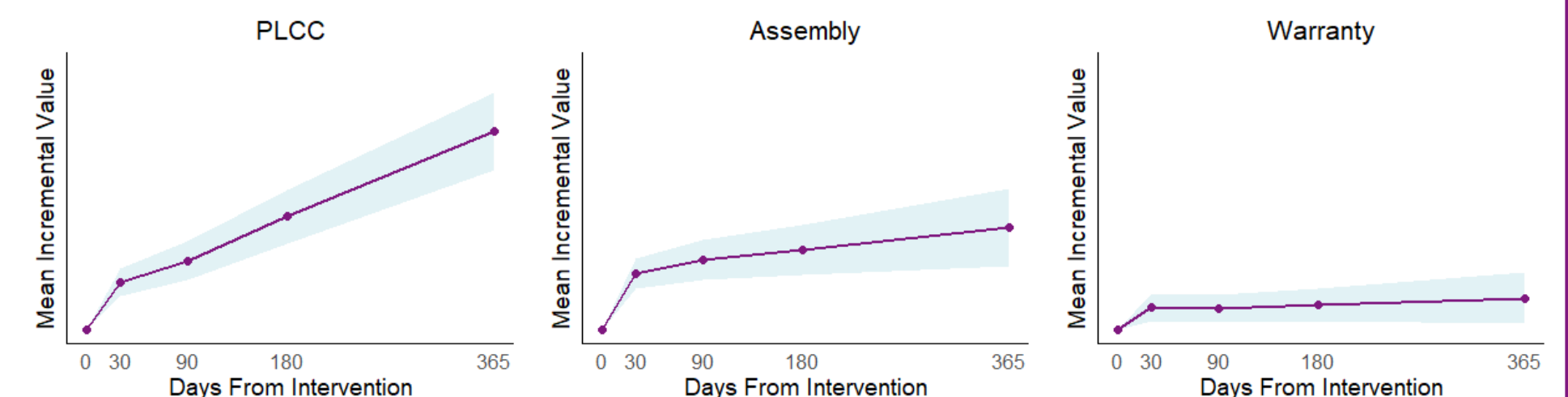
- This is a causal question. For each customer, we observe (1) what happened if they signed up or (2) what happened if they did not, but **never both**. We have no ground truth.
- Features (X) might be related to the intervention (I) and the outcome (Y). We will not be able to learn the effect of the intervention unless we **control for confounding variables**.



Approach: Utilizing **Double Machine Learning** and **Causal Forests** allows us to:

- Estimate incremental value at the **individual** customer level
- Control for a **high-dimensional** set of potential **confounders**
- Identify **differentiated incremental values** across different customer segments
- Quantify our uncertainty via **confidence intervals** and perform statistical significance tests

For each service, we create models for different time points to estimate the trajectory of incremental value.*



- Highest average incremental value**
- Linear trend for over time**
- Asymptotic trend**
- Comparison to PLCC suggests differences in **long vs. short-term engagement** services
- Hypothesis:** Case-specific nature of engagement with Warranty leads to fewer touchpoint opportunities

* Note: Y-axis values removed for confidentiality purposes.

MODELING

SIGN UP PROPENSITY

Key Question: At each product display page (PDP), which service is a given customer **most likely** to sign up for?

Key Modeling Considerations

- Sign ups depend on customer needs and preferences, which are revealed through their **historical and current interactions** with the Wayfair website.
- Wayfair wants to understand which customer features contribute to differences in response behavior, so **interpretability matters**.

Approach: We develop a novel algorithm, **Cluster-While-Classify (CWC)**, to find customer segments such that customers are **similar within** each segment but **differ across** segments in **how they respond** to Wayfair service impression actions.

CWC Procedure:

- Initialize clusters
- Model response to Wayfair service impression actions using **regularized multinomial logistic regression** for each cluster
- Iteratively update cluster assignment based on **logistic loss**
- After obtaining final assignments, fit a **decision tree** to classify observations into clusters and use the fitted tree to **classify new observations**

Results: 0.71 Macro Avg F1-Score (Using Only 7 Features)

MODELING

- Cluster 1**
 - Has purchased assembly in the past
 - Has visited at least 1 assembly-eligible PDP in current session
 - Current PDP assembly price > \$40

- Cluster 2**
 - Current PDP product price > \$446
 - Email acquired more than 1 year ago, but has never purchased
 - No services in Cart currently

- Cluster 6**
 - Has only viewed current PDP once
 - Has visited at least 1 assembly-eligible and 2 warranty-eligible PDPs in current session
 - Current PDP assembly price < \$40

At the current PDP, if Wayfair shows service impressions for PLCC, Assembly, & Warranty, we obtain an **ordering of responses from most likely to least likely** for each cluster:

	Most Likely Response				Least Likely Response
1	Assembly	Warranty	No Service	PLCC	
2	PLCC	Assembly	Warranty	No Service	
6	No Service	Assembly	PLCC	Warranty	

OPTIMAL SERVICE PRESENTATION

Key Question: How can we make use of our NPV estimates and service sign up propensities to **personalize service impressions** to customers?

Key Modeling Considerations

- Formulate as an optimization problem that can be **solved efficiently**
- Strike the right balance between **reducing customer cognitive load** and **maximizing revenue**
- Quantify expected business impact** on key metrics through simulation and **sensitivity analysis** under different assumptions

Objective: Maximize Expected Value of Sign Up – Cost of Display
Decision: Which Service Impression Action to Display



Opportunities to personalize services for over **60%** of customer sessions



Approximate expected uplift of **\$80-100M** in annual revenue⁽¹⁾

CONCLUSIONS & NEXT STEPS

Conclusions:

- We have created an **end-to-end analytical framework** for personalizing service messaging for Wayfair, which can also be **applied generally in a wide range of retail settings**.
- Our **methodologies can be adapted** to answer questions of a similar nature across Wayfair's other business functions.

Next Steps:

- Stress-test** and improve our models based on stakeholder and service owner feedback
- Expand scope** of modeling to more services and additional session touchpoints
- Perform **A/B testing** to validate our estimates about the impact of our framework on revenue

(1) Hypothetical estimates were obtained by MIT Capstone students using financial and web traffic data for FY2019