



# **Predicting a User's Next Instacart Order**

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## Objective and Problem Setup





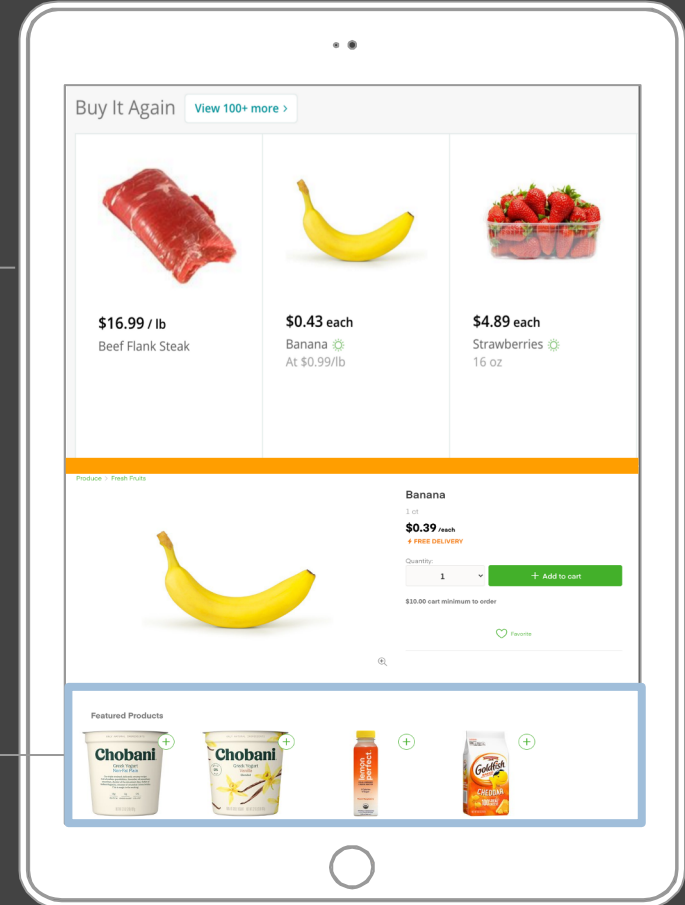
*Objective: To predict all of the reorders in a user's next cart*

*But first... why?*

## Use Cases:

1) Buy-it-again recommendations

2) Frequently bought with...



## The Dataset





**33,819,106**

**Total rows (1 per product per order)**

**3,346,08**

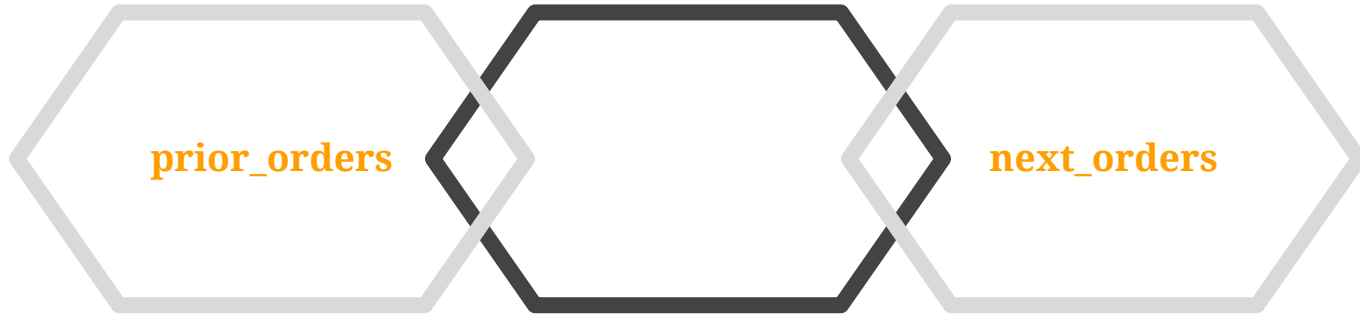
**Total orders**

**3**

**337,418**

**Total users**

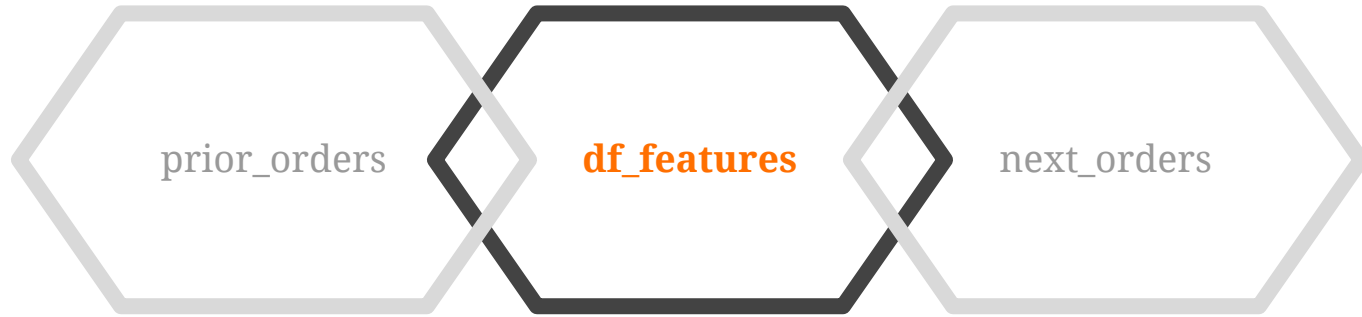
## THE DATASET



Contains prior order  
details for all users

Contains order  
details for each user's  
'next' order

## THE DATASET



Contains prior order  
details for all users in  
next\_orders

Contains order  
details for each user's  
'next' order

### **df\_features**

- Contains user & product statistics from prior\_orders
- Contains next\_order details
- Modeling done on these inputs

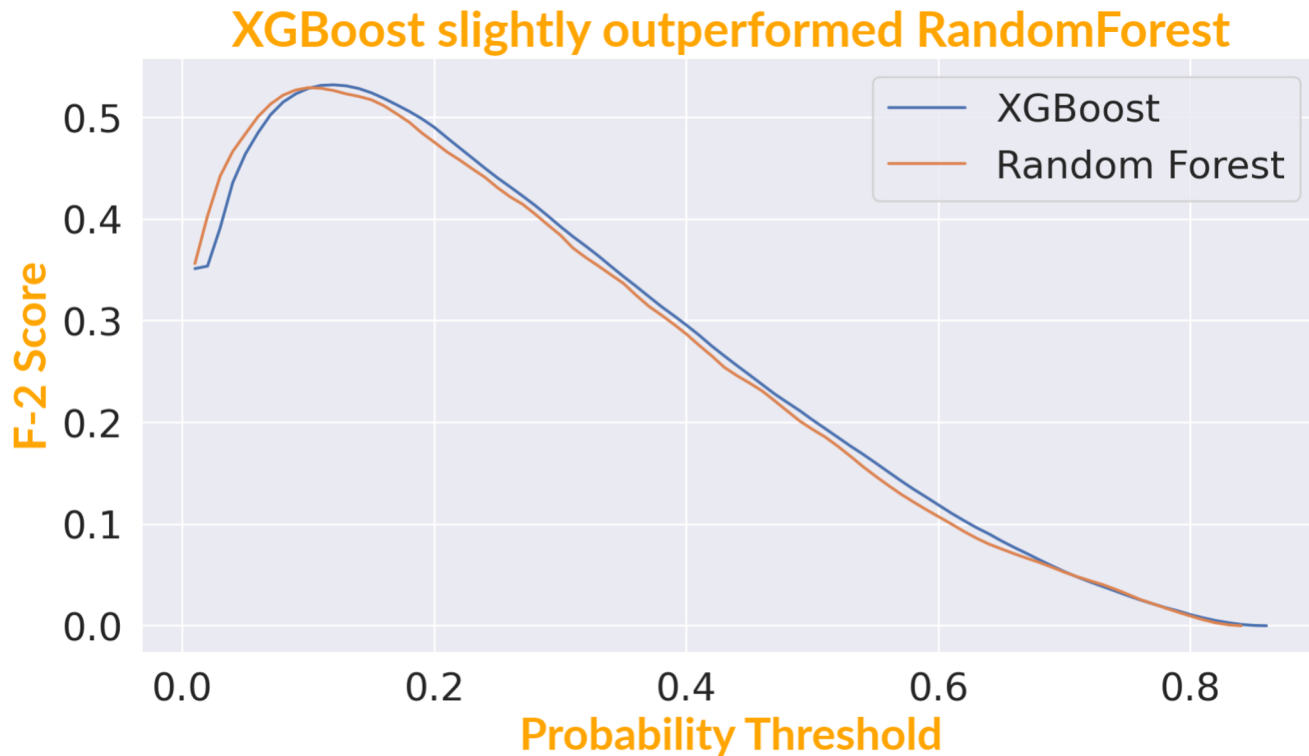


## Model Selection & Results



Class imbalance  
apparent

- 10% of target are reorders



## Used grid search to define the optimal XGBoost parameters:

*Took over 28 hours using  
16 vCPUs and just 25% of  
the dataset!*

learning_rate	<b>0.009</b>
n_estimators	<b>400</b>
max_depth	<b>7</b>
colsample_bytree	<b>0.8</b>
min_child_weight	<b>9</b>



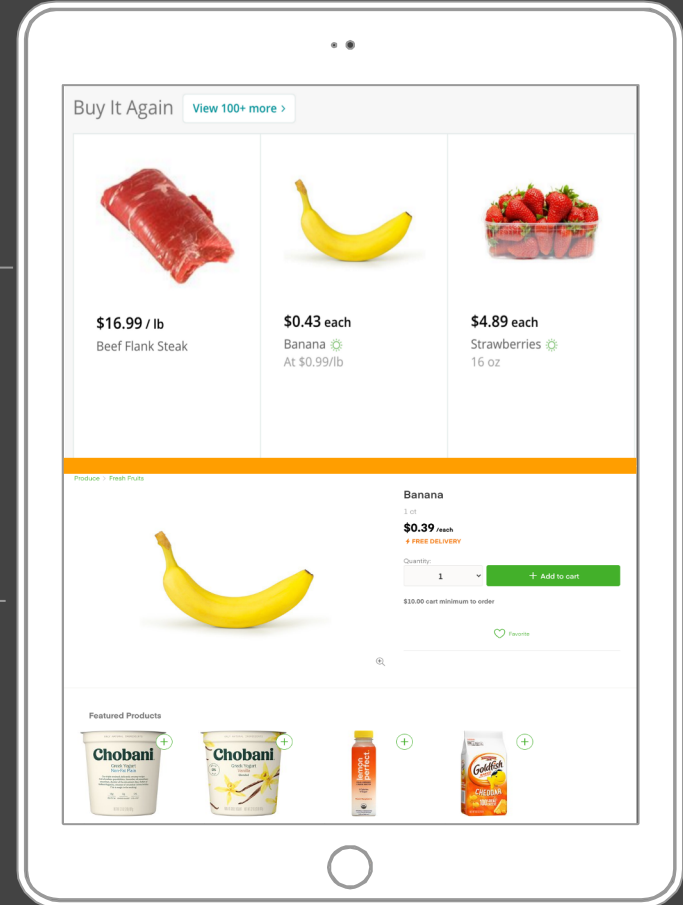
# Scoring

2018 Kaggle competition crowned winner based on  
resulting **F-1** scores

*Do you remember our use case?*

## Use Cases:

- 1) Buy-it-again recommendations
- 2) Frequently bought with...



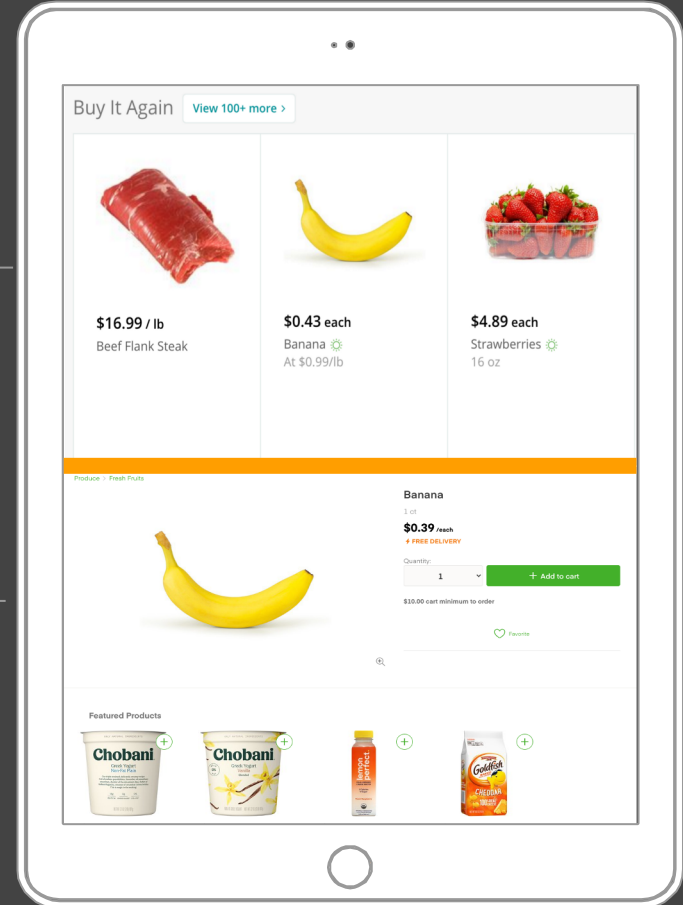
## Use Cases:

- 1) Buy-it-again recommendations
- 2) Frequently bought with...

*But how do these features help Instacart?*

1. User ease of use

1. **Increase product conversion rates**



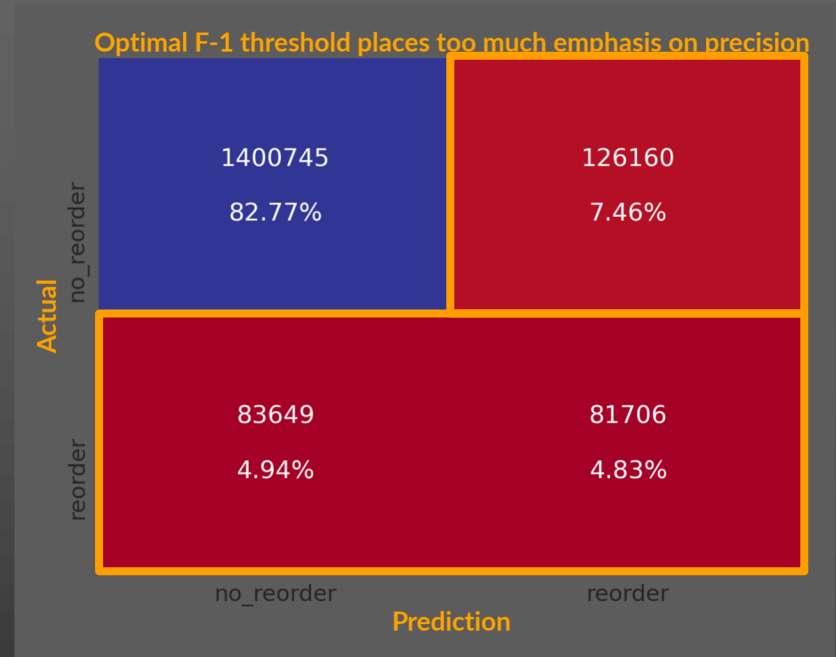
What's Instacart's risk of incorrectly classifying an input as positive (reorder)?

Not Much.

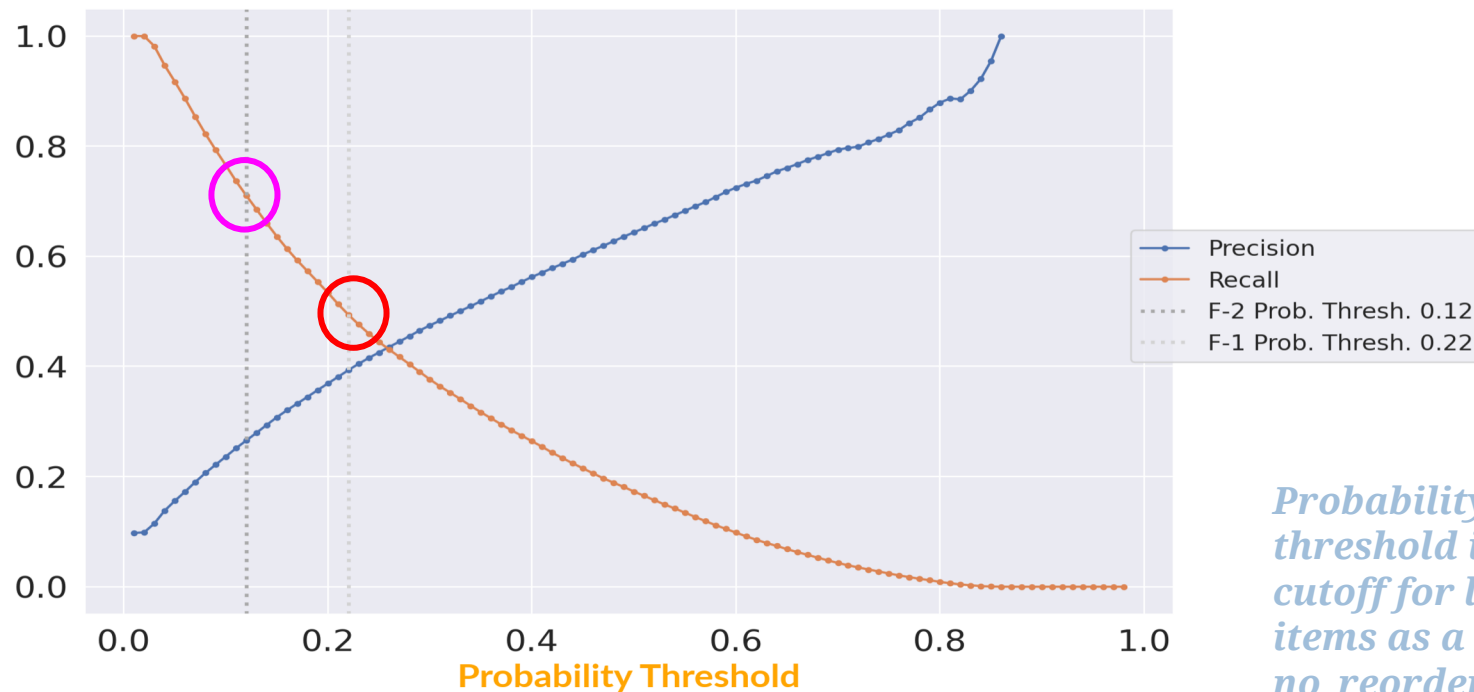
In fact, we may be better off including items that the user is *less likely to buy* based on their prior orders. This will help Instacart to increase conversion!

Therefore, we ought to prioritize **recall**!

Recall = True Positives / Actual Positives



## Choosing a threshold with an ideal recall/precision balance using F-2 Scores



*Probability threshold is the cutoff for labeling items as a reorder/no\_reorder*



## MODEL RESULTS

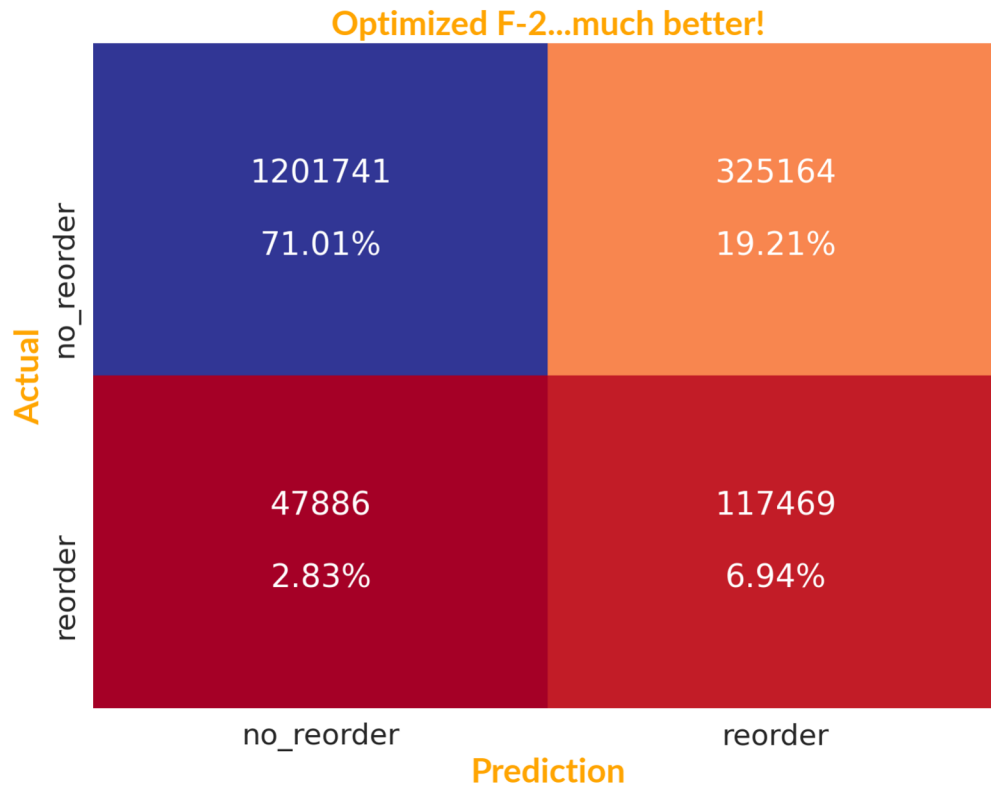
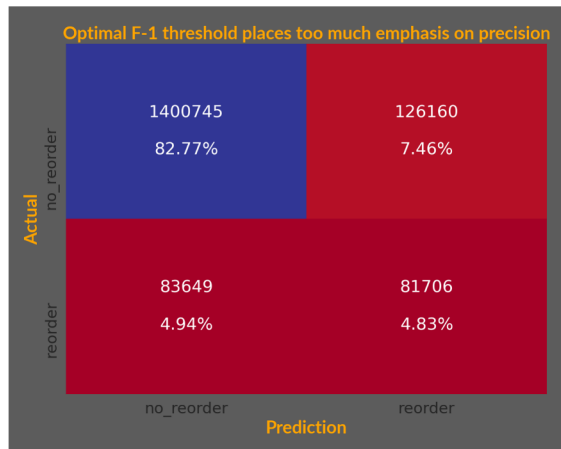
Adjusted F-2 Score

0.53

Probability Threshold

0.12

*Previous threshold (0.22)*



**THANK  
YOU!**

**Any questions?**

**You can find me...**

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On GitHub: [www.github.com/edubu2](http://www.github.com/edubu2)

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**THANK  
YOU!**

## APPENDIX



## Algorithms

- XGBoost
- RandomForest

## Techniques

- Grid Search & K-Fold Cross Validation
- Precision/Recall Curves
- F-1 & F-beta (F-2) scores

## Python Libraries

- Scikit-learn, StatsModels
- multiprocessing
- pickle
- pandas/numpy

## Tools

- PostgreSQL
- Tableau
- Google Cloud Deep Learning VMs
- Jupyter Notebook



Some key features (32 total):

User Features	Product Features	User/Product Features
avg_cart_size	percent_reorders	order_streak
days_since_prior_order	qty_sold	last_five_buys
avg_time_between_orders	qty_reordered	ln_last_cart (0/1)