

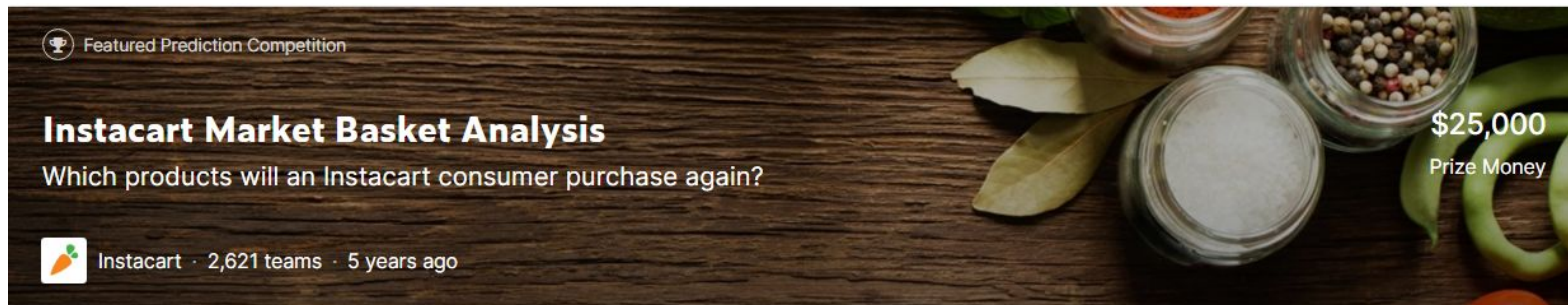
Instacart market basket analysis

Karen 2022/07/11

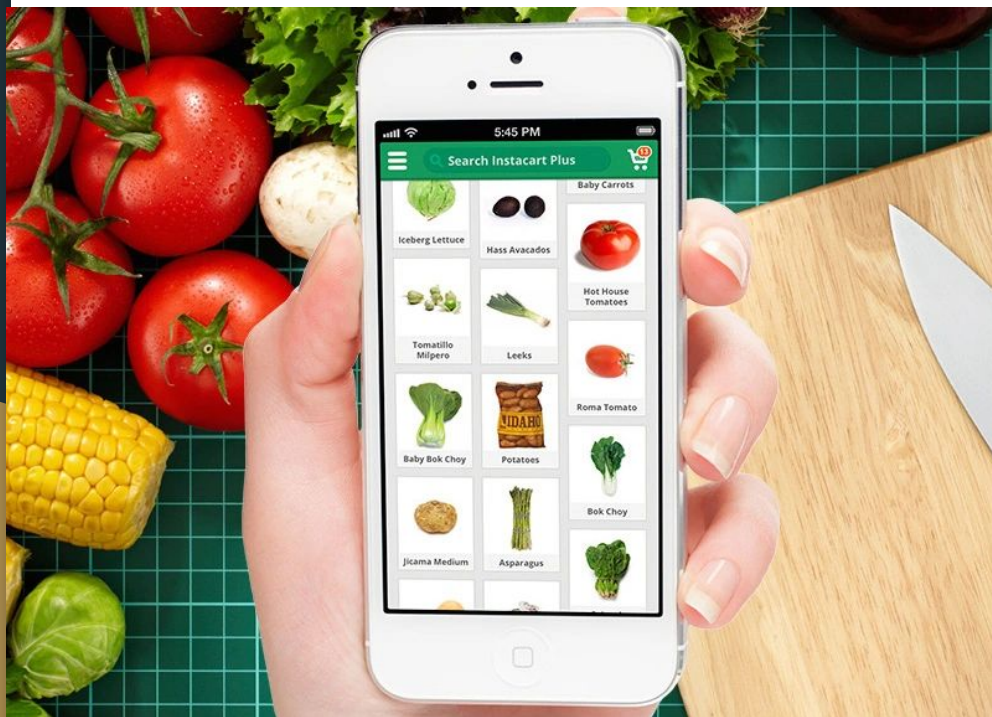
Outline

- Background
- Exploratory Data Analysis (EDA)
- Modeling (Xgboost)
- Apyori association analysis
- Summary

Background



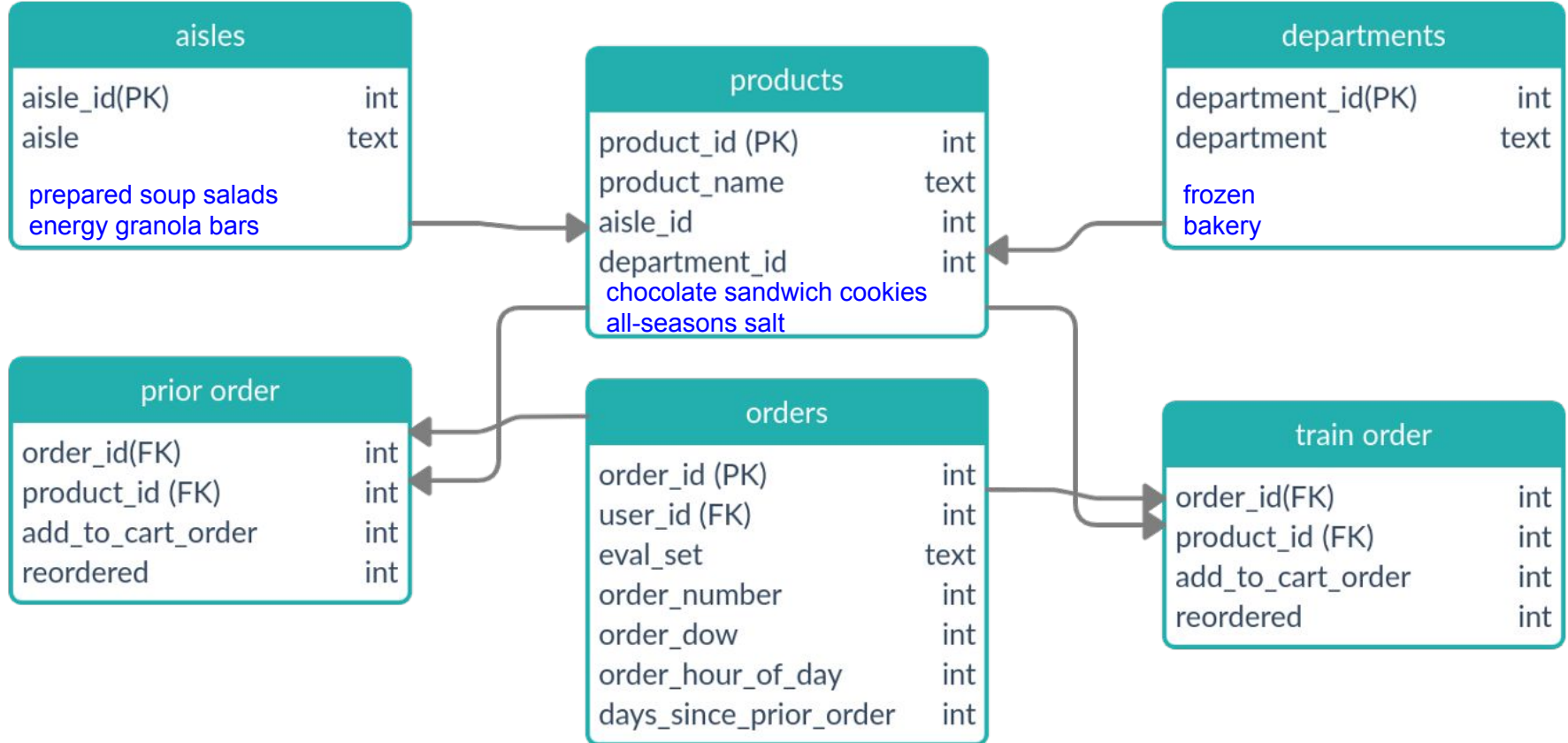
- 公開資料集 :Instacart Online Grocery Shopping Dataset 2017
- 根據之前的購買資料計算客戶再度訪問平台時可能再次購買的商品



- 成立於 2012 年
- 提供 O2O 生鮮雜貨代買代送服務
 - O2O (Online to Offline)
 - 團購服務
- 提供使用者比價資訊
- 提供購物專家購買的最佳路徑，依照天氣、交通調整運費

EDA - Briefly review

- 來自約 20 萬名 Instacart 用戶
- 約 340 萬的訂單數量
- 將近 5 萬件商品項目
- 這些商品的類別, 分佈 21 種
- 商品擺放的位置, 約有 134 個商品陳列走道位置
- 對於每個用戶提供 4 ~ 100 個訂單資料



	order_number									
	1	2	3	4	5	6	7	8	9	10
User A	p	p	p	p	p	tr				
User B	p	p	p	p	p	p	p	p	te	
User C	p	p	p	p	p	p	p	tr		
User D	p	p	p	tr						

Prior (p)

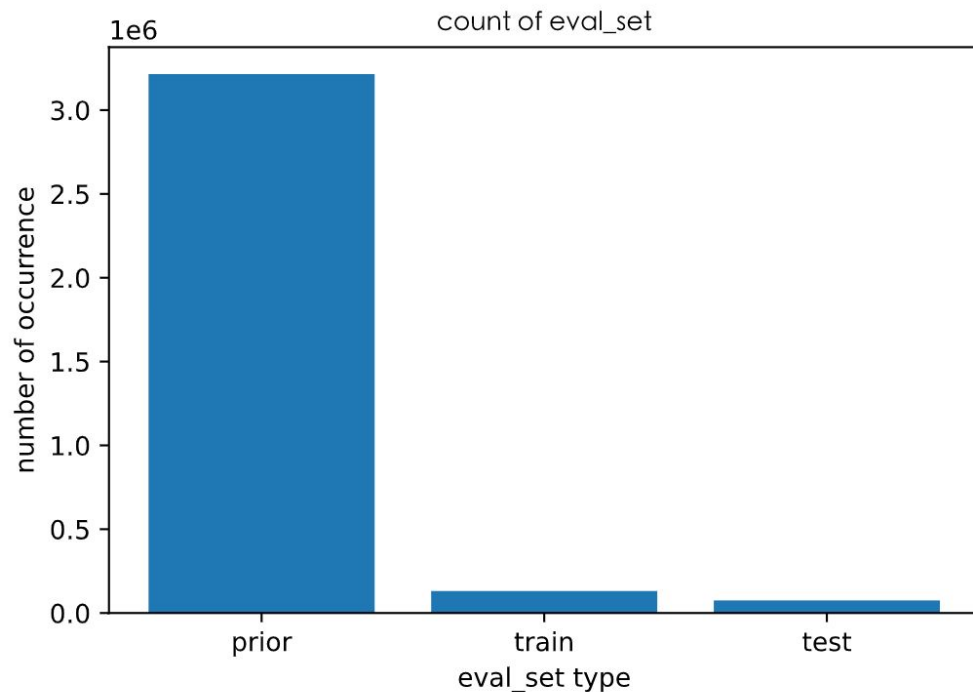
Train (tr)

Test (te)

ref: <https://medium.com/@PTLin0519/kaggle%E7%AB%B6%E8%B3%BD-instacart-market-basket-analysis-%E4%B8%80-%E7%AB%B6%E8%B3%BD%E7%Bo%A1%E4%BB%8B%E8%88%87%E6%8E%A2%E7%B4%A2%E6%80%A7%E6%95%B8%E6%93%9A%E5%88%86%E6%9E%90-972183f2a19b>

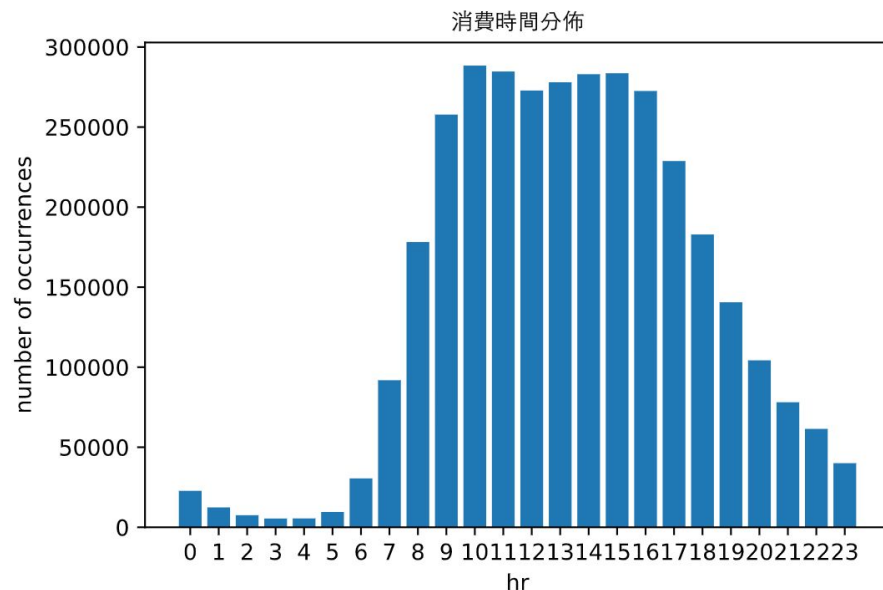
EDA- Dataset review

- 拆分三個資料集
 - Prior dataset
 - Train dataset
 - Test dataset



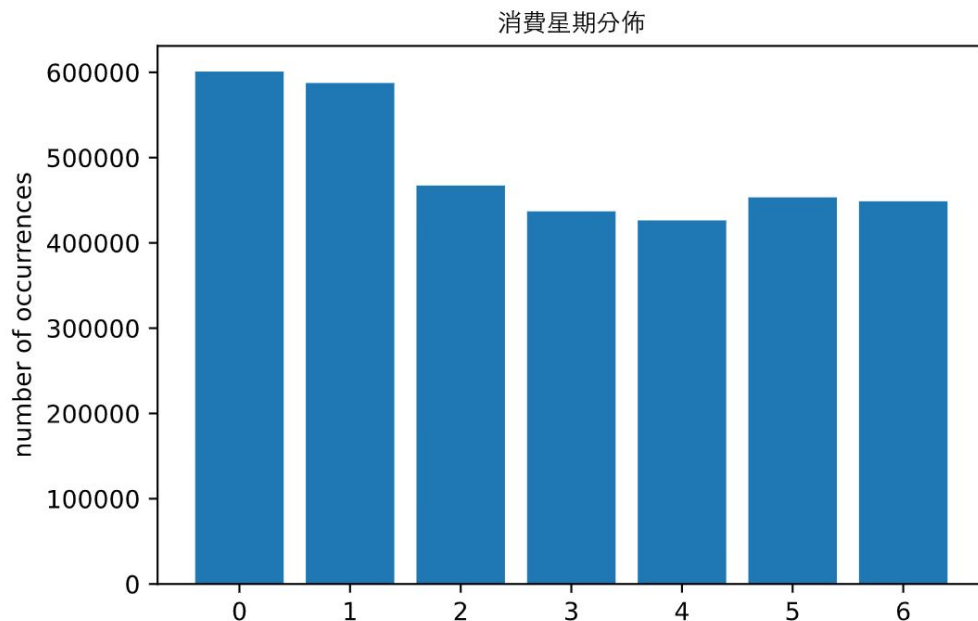
EDA- Distribution of purchase time

- 消費集中在10am~4pm



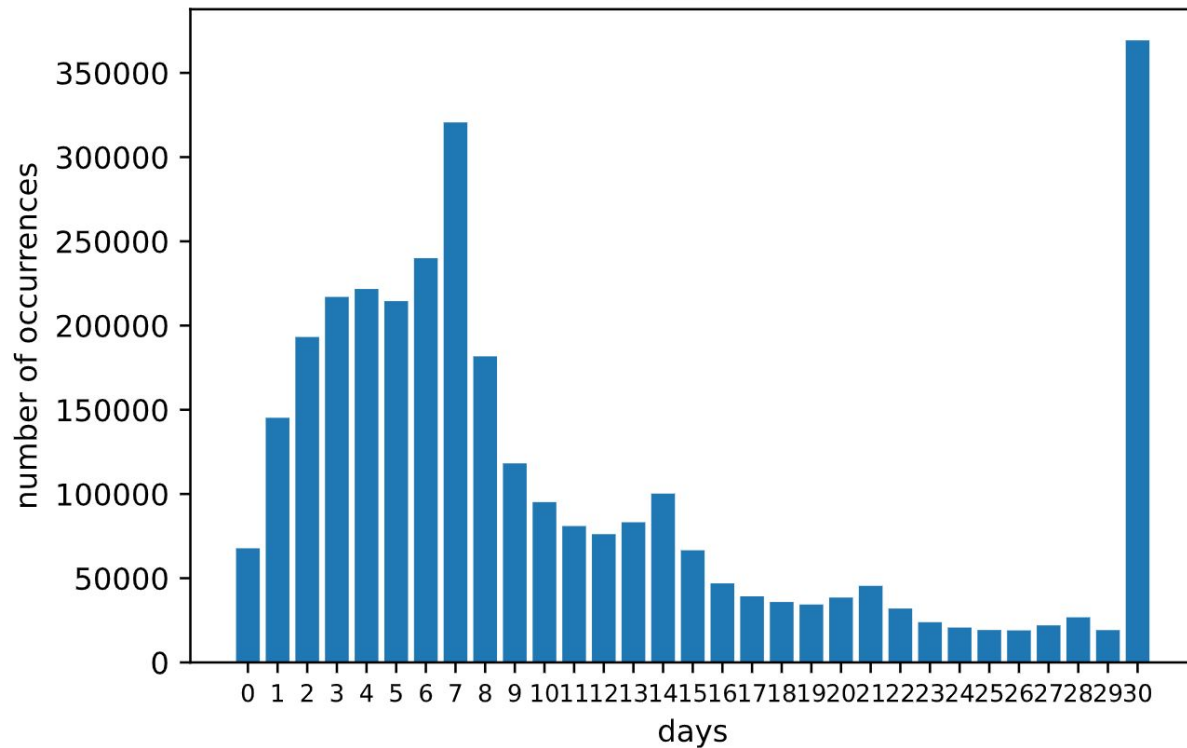
EDA- Distribution of purchase weekday

- 0為週六、1為週日以此類推
- 週六最愛買！
- 週三小週末反而消費少



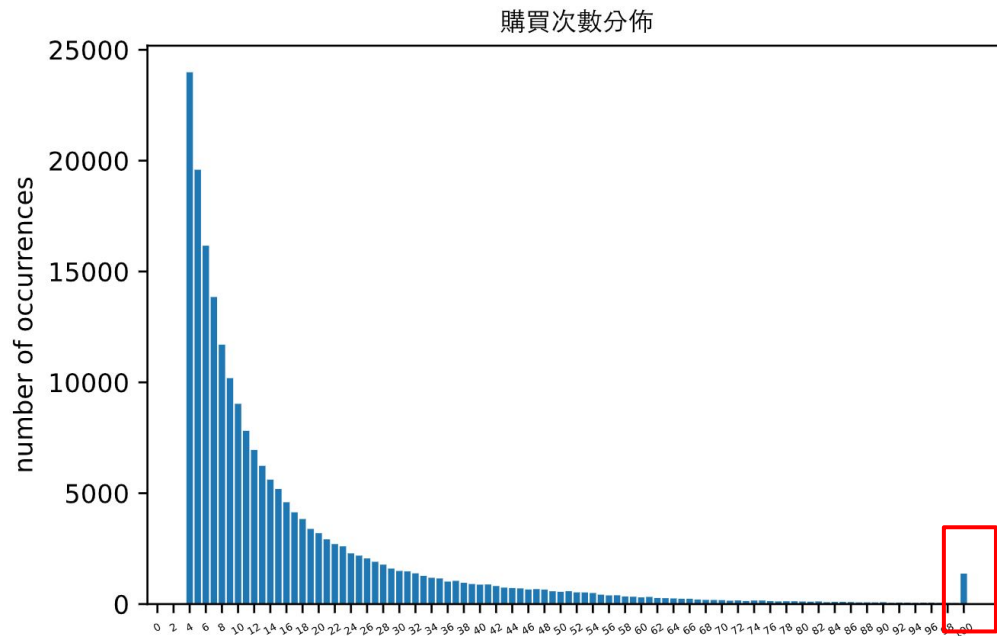
EDA- Distribution of order interval

距上次購買間隔天數分佈

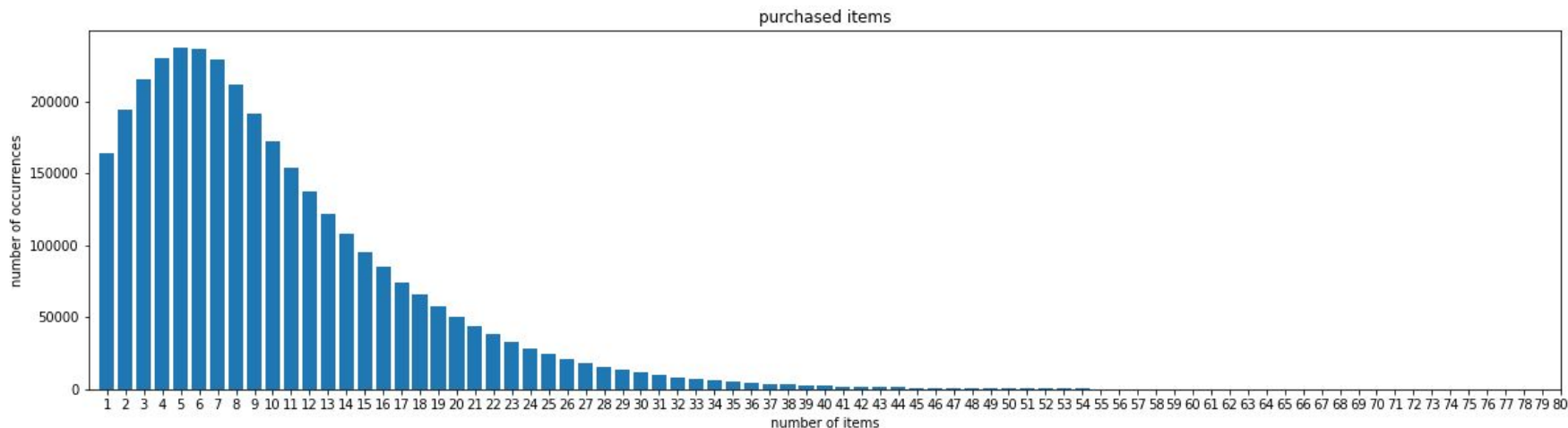


EDA- Purchase frequency

- 最低購買次數為4次
- 最高為100次



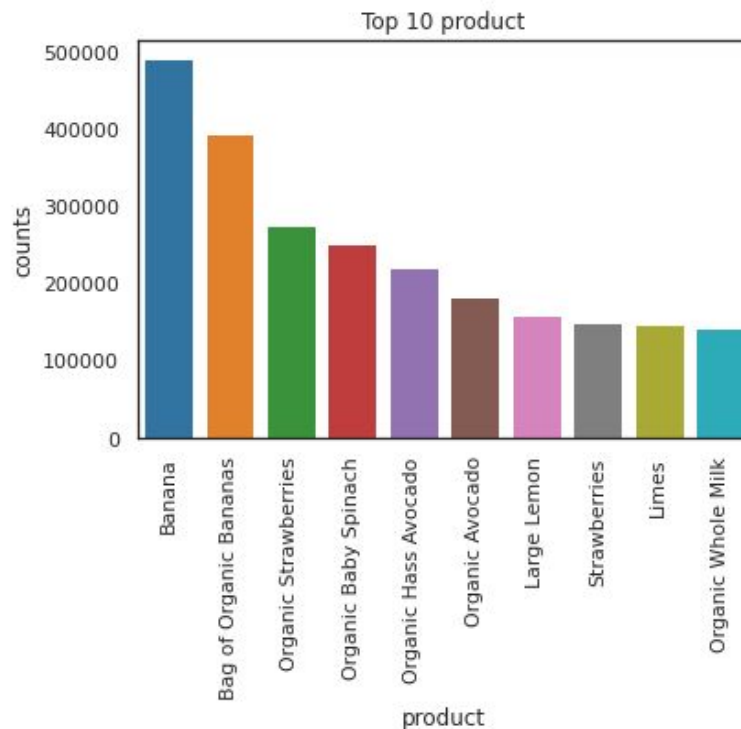
EDA- Distribution of numbers of items



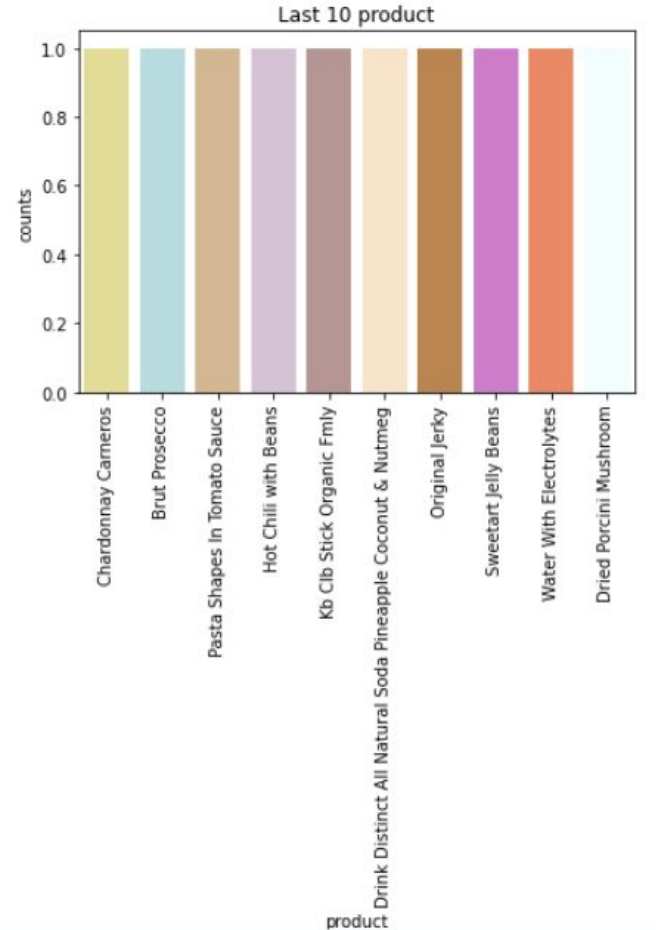
[illegible]

EDA- Purchase frequency by product

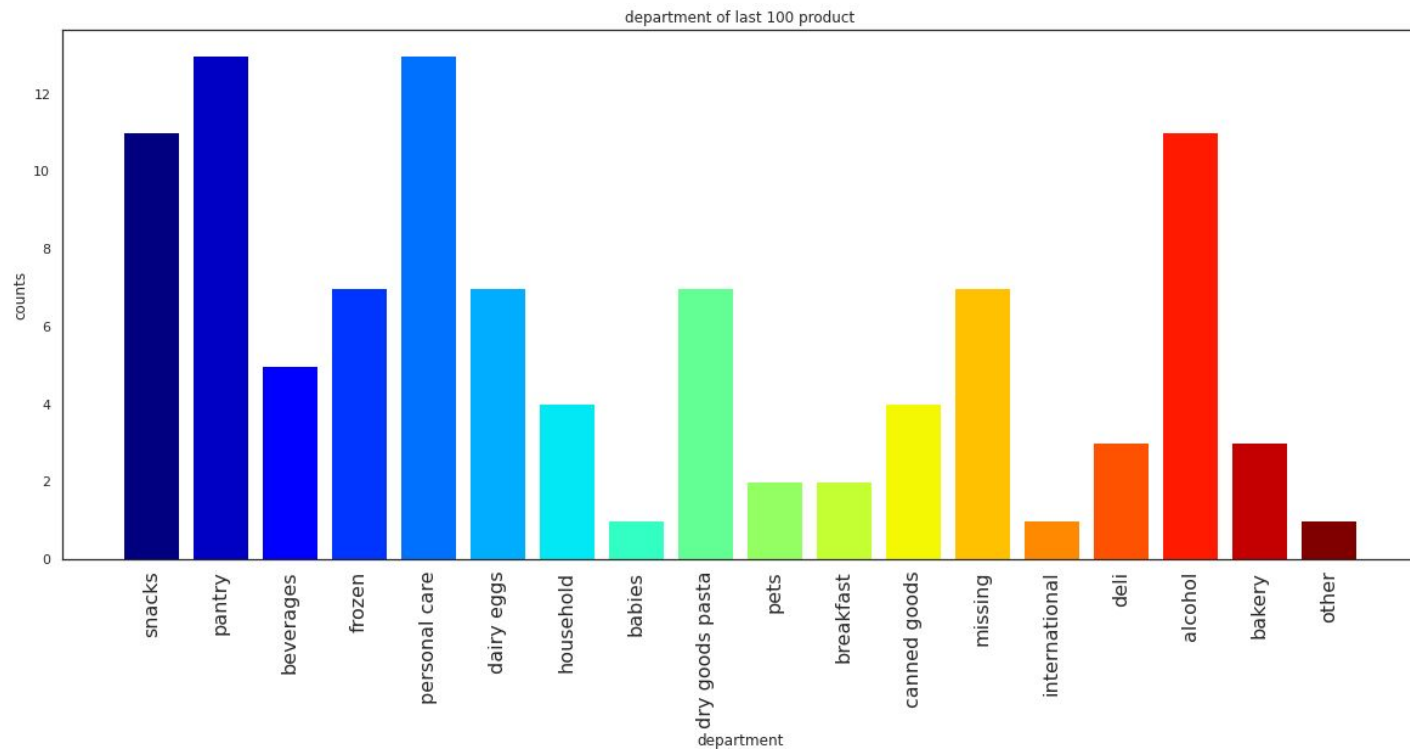
- 分析過往訂單記錄, 找出 TOP 10 明星商品
- Banana, the king of the fruit?



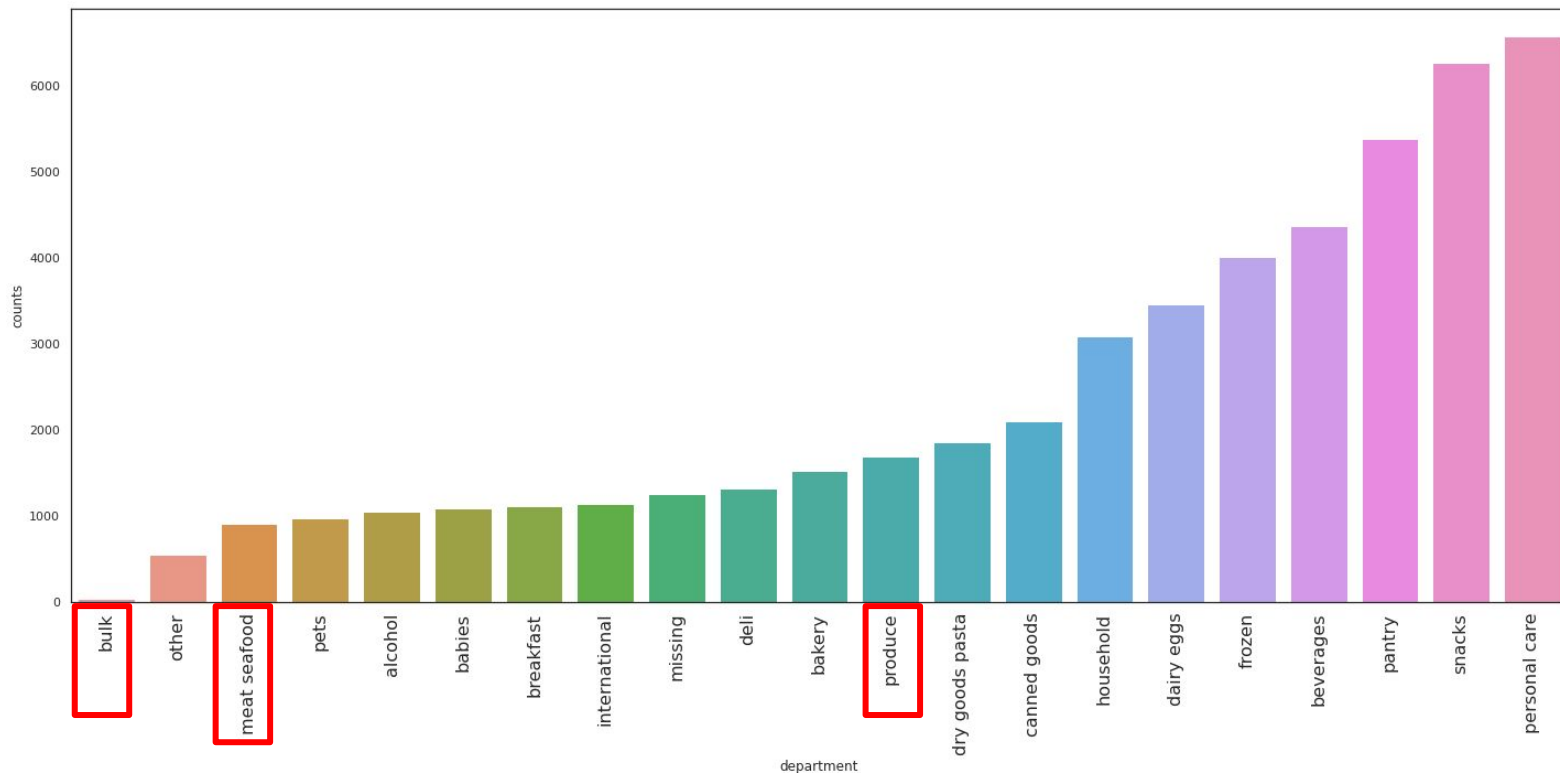
- 分析過往訂單記錄，找出 Last 10 不受歡迎商品
- 特殊食品、酒？！



EDA- Distribution of the department which product is only purchased one



EDA- Distribution of product department



Define reorder

- reordered: 1 if this product has been ordered by this user in the past, 0 otherwise

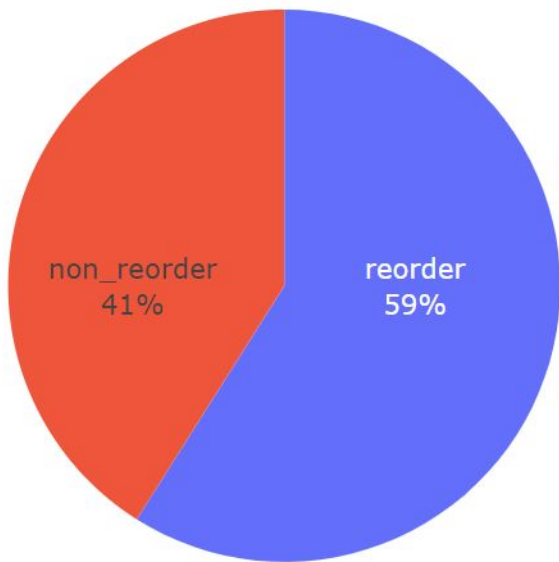
first purchase	
product_name	reordered
A	0
B	0

second purchase	
product_name	reordered
A	1
C	0
D	0

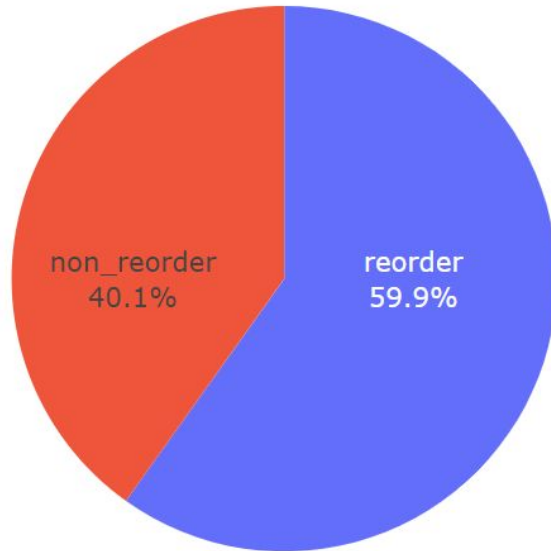
third purchase	
product_name	reordered
A	1
B	1
E	0

EDA - Reorder ratio

prior

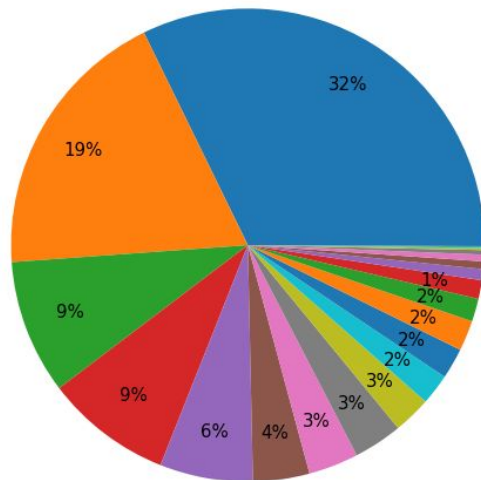
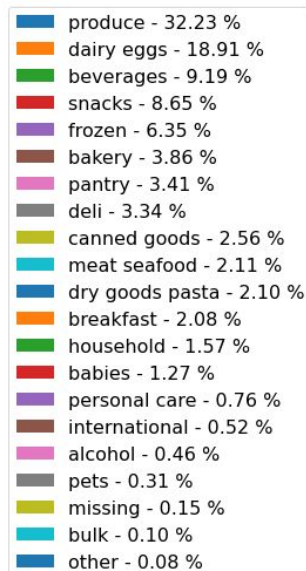


train

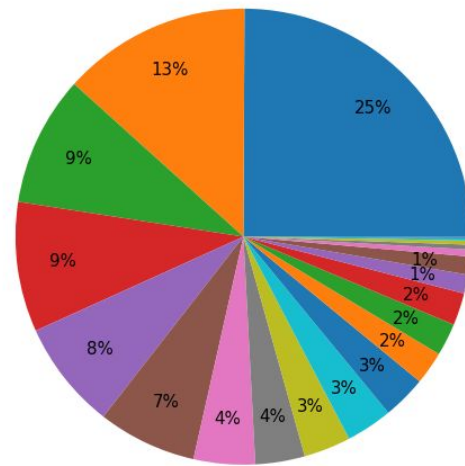
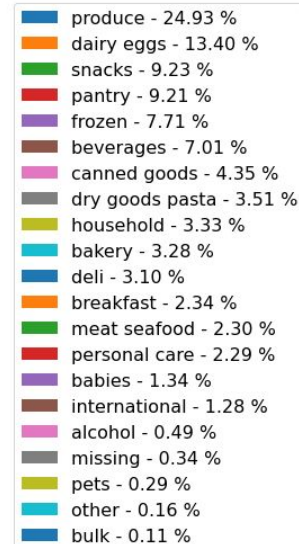


EDA - Distribution of department by reorder

reordered



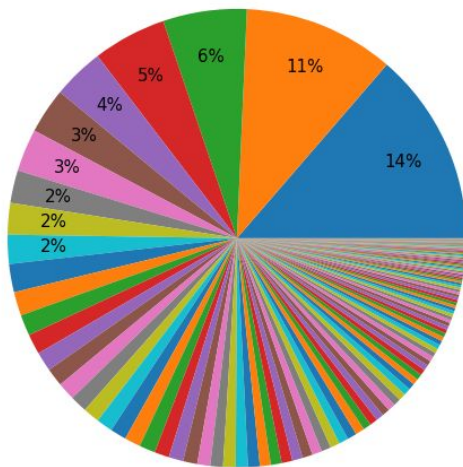
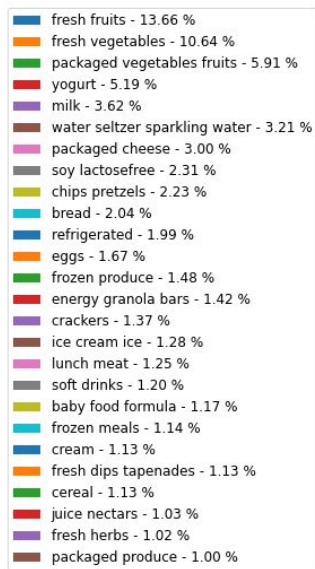
non_reordered



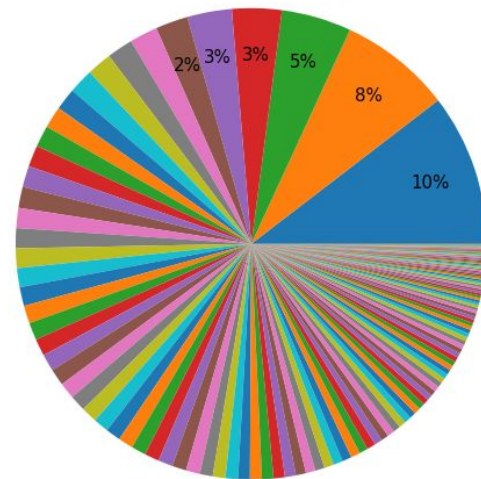
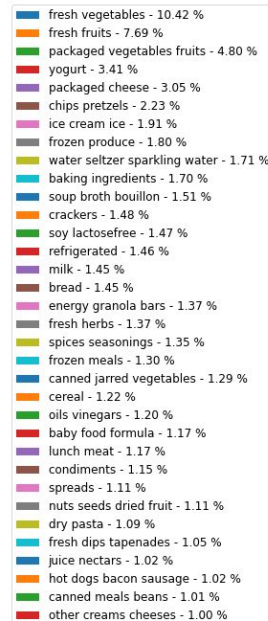
	product_id	product_name	aisle_id	department_id	department	aisle
42767	39812	Organic Thyme	16	4	produce	fresh herbs
42755	31717	Organic Cilantro	16	4	produce	fresh herbs
41883	11165	Tuscan Kale	83	4	produce	fresh vegetables
42464	19881	Bartlett Pear	24	4	produce	fresh fruits
41867	10358	Organic White Mushrooms	83	4	produce	fresh vegetables
42634	42411	Young Coconut	24	4	produce	fresh fruits
41821	6773	Onions	83	4	produce	fresh vegetables
41187	4539	Santa Fe Caesar Complete Salad Kit	123	4	produce	packaged vegetables fruits
42796	15772	Pineapple Spears	32	4	produce	packaged produce
41677	43787	Bolthouse Farms Baby Cut Carrots	123	4	produce	packaged vegetables fruits

EDA - Distribution of aisles by reorder

reordered



non_reordered




```

fresh fruits                2726251
fresh vegetables            2123540
packaged vegetables fruits  1178700
yogurt                     1034957
5 milk                      722128
water seltzer sparkling water 640988
packaged cheese             598280
soy lactosefree             460069
chips pretzels              444036
bread                       408010
refrigerated                397213
eggs                        333408
frozen produce              295616
energy granola bars         283351
crackers                    272645
ice cream ice               256194
lunch meat                  249963
soft drinks                 238981
baby food formula           233042
frozen meals                228222
Name: aisle, dtype: int64

```

```

fresh vegetables            1445090
fresh fruits                1066410
packaged vegetables fruits  665106
yogurt                     472626
packaged cheese             423182
chips pretzels              309703
ice cream ice               264907
frozen produce              249491
water seltzer sparkling water 237162
baking ingredients          235996
soup broth bouillon         208858
crackers                    205785
soy lactosefree             204424
refrigerated                201896
15 milk                     201531
bread                       200459
energy granola bars         190484
fresh herbs                  190007
spices seasonings           187516
frozen meals                180298
Name: aisle, dtype: int64

```


主鍵Primary key
(根據prior訂單購買的產品)

參數X
(根據prior訂單建立)

欲預測的變數Y

		(根据prior的串建立)			train/test	future orders	
user_id	product_id	eval_set	order_id	reordered
1	196	...			train	1187899	1
1	10258				train	1187899	0
1	10486				train	1187899	1
1	10686				train	1187899	1
1	15435				train	1187899	0
1	12376				test	1187968	
2	11698				train	1256788	1
2	12495				train	1256788	1
2	14571				test	1257530	

ref:<https://medium.com/@PTLin0519/kaggle%E7%AB%B6%E8%B3%BD-instacart-market-basket-analysis-%E4%B8%80-%E7%AB%B6%E8%B3%BD%E7%Bo%A1%E4%BB%8B%E8%88%87%E6%8E%A2%E7%B4%A2%E6%80%A7%E6%95%B8%E6%93%9A%E5%88%86%E6%9E%90-972183f2a19b>

Feature Engineering

● 商品面

- 被購買的次數
- 被重複購買比例
- 被第一次購買的次數
- 被第二次購買的次數

● 客戶面

- 距離上一次購買天數的總和和平均
- 平均單次購買的產品數量
- 平均將該商品放入購物車的順序
- 平均購買該商品的次數
- 連續沒有購買該商品的次數

[illegible]







Modeling

XGBoost

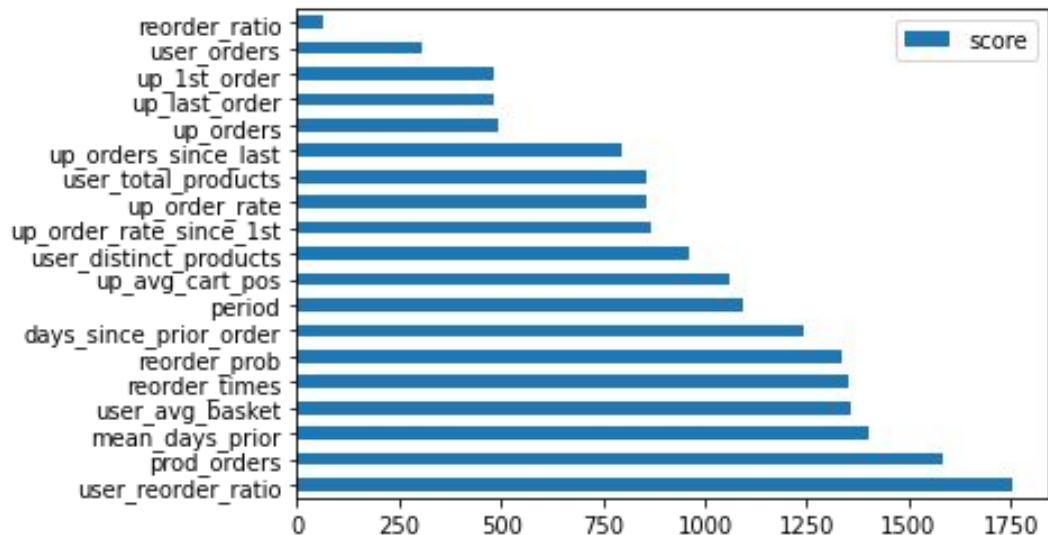
- 高準確度
- 適合作為 baseline model
- 準確度較不受無效特徵干擾

$$F1\ score = 2 * \frac{Precision * Recall}{Precision + Recall}$$

- Baseline model score : 0.37625
- Grid Search score : 0.37968 (38.3%)

#	△	Team	Members		Score	Entries	Last
1	—	胡萝卜			0.40914	62	5Y
2	—	===== KEEP OUT 🍌 =====			0.40820	138	5Y
3	—	sjv			0.40810	76	5Y

Features importance analysis



- 重複購買率
(user_reorder_ratio)
- 被購買的次數 (prod_order)
- 距離上一次購買天數的平均
(mean_days_prior)
- 平均單次購買的產品數量
(user_avg_basket)

Apyori association analysis

- 關聯分析: 找尋資料彼此之間的關聯, 透過兩種主要的方式來進行分析
 - 頻繁項集: 經常一起出現的物品集合
 - 關聯規則: 表達數據之間的可能存在很強關聯性
- 支持度(Support): 表示為 item-set 在整個 AllSamples 中出現的頻率
$$\text{Support}(X) = \text{number}(X) / \text{number}(\text{AllSamples})$$
- 信心度(Confidence): 表示當事件X發生的情況下, 同時會發生Y的可能性
$$\text{Confidence}(X \rightarrow Y) = P(Y|X), = P(X \cap Y) / P(X)$$

交易編號	商品
0	豆漿、萵苣
1	萵苣、尿布、葡萄酒、甜菜
2	豆漿、尿布、葡萄酒、橙汁
3	萵苣、豆漿、尿布、葡萄酒
4	萵苣、豆漿、尿布、橙汁

Apyori Association Analysis Case

prior



test

Support: 0.028
Confidence: 0.45

Organic Strawberries (275577)

Organic Hass Avocado (42333)

15%!



Support: 0.006
Confidence: 0.278

Boneless Skinless Chicken Breasts (52369)

Organic Baby Spinach (8823)

16%!



Summary

- 藉由過去的訂單數據, 利用 **model** 能預測下次顧客是否再度購買商品
- 觀察 **organic** 在資料集中, 可能成為重要的特徵
- 通過關聯分析發現訂單中的商品有交互關係, 例如購買A產品經常購買B

Future works

- **data augmentation**: 加入用戶最近的 3 - 5 個訂單提高訓練數據量
- **product feature engineering**: **organic** feature, alternative item
- **none prediction model**: 有可能用戶下一次訂單中不回購任何商品
- **other models**: RNN, CNN, **LGBMClassifier**
- **design new training flow**: 由於產品間有交互作用, 可以設計新的訓練流程, 不只是將每一個產品當成獨立的分類問題



Thank you