Breakfast Coupon Campaign Recommendations A data driven Solution

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Disclaimer

The proposed recommendation is exclusively data-driven, and it does not incorporate elements from any strategic marketing literature. It is not recommended for adoption, as the data may contain selection bias.

Basic Information's: Analyzing Coupon Campaign Performance

- To derive the findings of the analysis, we focused on orders that had the "breakfast" flag in the "cuisine" variable.
- The entire analysis was centered around the performance of previous coupon campaigns. We focused on customers who, at some point, had redeemed coupons.

Basic Information's: Group of Users

Users are grouped based on the following variables into the following categories:

- user_class_name: Loyal, All Star, Infrequent, One Timers, Common User, High Spenders
- City: Arta, Aegio, Drama, Xanthi, Rhodes, Rafina, Agrinio, Florina,
 Amaliada, Zakynthos, Ioannina, Didymoteicho, Alexandroupoli.
- Device: iOS, Android, Desktop, MobWeb
- paid_cash: Paid cash, Online payments

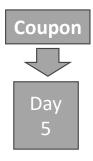
Technical idea: Analyzing the Impact of

Coupons on Customer Behavior

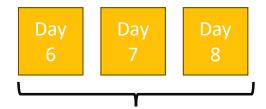


Measurements before the coupon

- 1) Median order amount before coupons.
- 2) Count of the total orders before coupons.

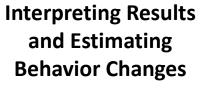


VS



Measurements after the coupon

- 1) Median order amount after coupons.
- 2) Count of the total orders before coupons.



1st KPI - Difference in Median Order Amount
This KPI measures the change in median order amounts before and after coupons.

2nd KPI - Difference in Order Frequency

This KPI quantifies the impact of coupons on order frequency by comparing counts.

Technical idea: Defining Success Criteria for the Campaign

A campaign is considered successful for a particular group of customers when, after the coupon, it is observed that:

- 1) The average order frequency of the group from efood increased.
- 2) The average order consumption amount of the group per order also increased.

Results: Analyzing the Impact of Coupons on Customer Behavior group by user_class_name

The first group of users that was analysed consists of groups based on the 'user_class_name' variable.

Group	Number of cases	Diff of average order frequency of the group before and after coupon	Diff average order consumption amount of the group per order before and after coupon
Loyal	1915	-0.084595	-0.037681
All Star	3308	-0.071644	-0.085144
Infrequent	266	<mark>0.112782</mark>	<mark>0.265602</mark>
One Timers	317	0	0
Common User	379	<mark>0.047493</mark>	<mark>0.233597</mark>
High Spenders	187	0.021390	<mark>0.088271</mark>

Results: Analyzing the Impact of Coupons on Customer Behavior group by user_class_name

- From the user_class_name variable, it appears that the following categories respond better to coupon campaigns:
 - 1) Infrequent
 - 2) Common User
 - 3) High Spenders
- Among these, Infrequent users exhibit the highest increase.
- For the next analysis, we will focus exclusively on these three user categories.

Results: Analyzing the Impact of Coupons on Customer Behavior group by device

 Since the cases within the categories iOS, Desktop, and MobWeb have been limited, a differentiation will be made into Android and not Android categories.

Group	Number of cases	Diff of average order frequency of the group before and after coupon	Diff average order consumption amount of the group per order before and after coupon
Android	448	<mark>0.131696</mark>	<mark>0.378311</mark>
not Android	384	-0.031250	-0.000851

• Among these, Android users exhibit the highest increase.

Results: Analyzing the Impact of Coupons on Customer Behavior group by city

- If we limit ourselves to Android users, the 'city' variable does not play a significant role in campaign selection.
- However, if we do not limit the analysis, the following cities appear to have the best adaptation to campaigns, with more than 50 cases:

Group	Number of cases	Diff of average order frequency of the group before and after coupon	Diff average order consumption amount of the group per order before and after coupon
Xanthi	98	0.183673	0.447619
Rhodes	188	0.021277	0.119681
Agrinio	64	0.109375	0.019792
Ioannina	221	0.076923	0.284118

Results: Analyzing the Impact of Coupons on Customer Behavior group by paid_cash

• The differentiation within this specific category appears to be irrelevant, as both groups behave in the same manner.

Recommendation

From the previous analysis, the two user groups with the most successful behavior in coupon campaigns are:

- 1. Infrequent, Common User, and High Spenders with Android devices, with a particular emphasis on the Infrequent category.
- 2. Infrequent, Common User, and High Spenders, with a particular emphasis on the Infrequent category, from the cities of Xanthi, Rhodes, Agrinio, and Ioannina.