

# Crowdrebate: Towards Getting More Rebate for Customers

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**Abstract**—With the flourishing development of e-commerce platforms, there are various promotion operations, such as offering coupons with a high threshold to stimulate customers to buy more products all at once. Nevertheless, customers always find that the price of the products they intend to order is below the coupons or delivery service threshold. They would spend lots of time looking for others with the same need to place an order together to enjoy more instant benefits. These complex processes are unreliable and time-consuming for consumers.

In this demo, we present the Crowdrebate platform that collects users' requests, groups requests into a set of orders using suitable coupons by the Crowdrebate algorithm, and then delivers products to different receivers in an order, aiming at saving users' shopping time and helping users get more rebate from online promotions. Crowdrebate is equipped with the techniques, including the Crowdrebate algorithm grouping orders for maximum rebates, web crawling, items recommendation, and data visualization. We introduce the sketch of system architecture and demonstrate various scenarios via several case analyses.

## I. INTRODUCTION

With the flourishing development of e-commerce platforms, online shopping has become part of people's lives. To further stimulate online consumption, e-commerce platforms arrange various promotion campaigns on special days, such as Black Friday [1] in Western countries and Double Eleven [2] in China [3]. Along with numerous online promotions, some intermediate platforms have emerged in the market, such as Groupon [4] and Dealmoon [5], which aim to help users sort out the hot promotions of various e-commerce platforms.

However, sometimes even users know those promotions, they find it hard to meet the spending threshold to use a coupon or to enjoy the delivery service. Therefore, users will intentionally seek for group ordering with others to meet the threshold. Because all the e-commerce platforms support only one shipping address for one order, when multiple users place an order together, one of the users will inevitably receive all the products first and then distribute them to others respectively. These complex processes are unreliable and time-consuming for consumers. What is even more damaging to consumers' motivation to shopping is that increasingly complicated e-commerce promotion activities imposes extra workloads to customers, since multiple complex coupons rules may change on a daily basis [6].

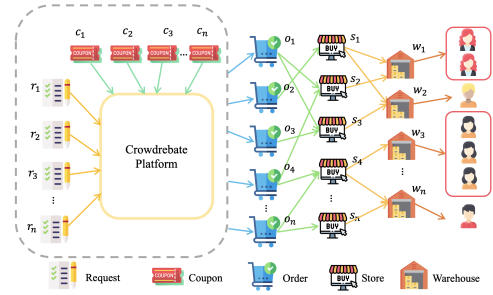


Fig. 1: The operation process of Crowdrebate

Under these circumstances, our demo-Crowdrebate firmly grasps consumer's pain points. Crowdrebate collects users' purchasing requests and group their requests into orders to maximize total benefit. Meanwhile, we give the exposure of online retailers' promotions so that more users would participate in promotions. To summarize, not only do we dedicate ourselves to help users get more benefits, we also devote ourselves to help e-commerce platforms fill more orders, increase platforms' revenue, and ultimately achieve a bilateral win-win situation.

Our platform operates as follows in Figure 1. Users can post their requirements corresponding to the information of the desired product. Based on the received requests, we make an order association and match the suitable coupons to maximize each order's rebate deducting the delivery costs, by our *crowdrebate algorithm* [7]. Because we place the orders for consumers, the products are first sent to our warehouses from the online stores, and then we will distribute the products to the according to users.

Generally, Crowdrebate relies on three components. The user interface provides service for users to post requests and track datas. The algorithm library supports our core features, group ordering and recommendation. The data manager handles business logic and data storage.

Thus, our demo has following contributions:

- We develop a platform for users to post for group ordering and track realtime information on mobile phones.
- We design two algorithms to combine orders for maximum rebates under different scenarios.
- We equip the platform with process automation and

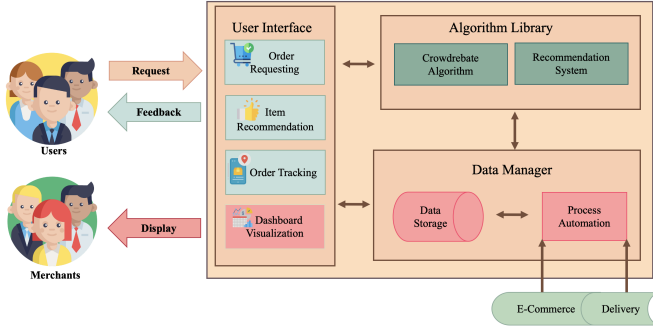


Fig. 2: The architecture of crowdbate

recommendation to handle large scale data and enrich user experience.

## II. SYSTEM ARCHITECTURE

Figure 2 depicts the architecture of Crowdbate platform. The user interface module runs on the client-side, enabling both general users and merchant users to get corresponding services. The other three core modules, process automation, crowdbate algorithm [7], and recommendation system, run on the server host and provide support for the core functionalities in Crowdbate. We introduce the technical details of these modules in the rest part of this section.

### A. User Interface

There are two types of users interacting with the user interface module, “general users” and “merchant users”. General users, who seek group buying to meet the threshold of rebates, can request and track their orders with the interface while getting a personalized recommendation. Merchant users from online shopping platforms can benefit from the data dashboard to gain a better understanding of user portraits. This module consists of the following four functions.

1) *Order Requesting*: Users with the need for group shopping can post their requirements on Crowdbate with this order requesting function. The input for this function consists of two parts, product information, and customer information. The former includes the shopping platform, product link, and promotion type, which help the web crawlers to target the specific item and its promotions. The critical part of customer information is expected time slots and delivery addresses, impacting the rounds and delivery costs for the crowdbate algorithm to run.

2) *Order Tracking*: The order tracking function enables users to manage all historical orders. No matter whether one’s item is waiting in the pool, being processed for a group order, or already on delivery, users can track real-time status for his order. Crowdbate platform utilizes the web crawler, the crowdbate algorithm, and the order bot to automate the process of group order and updates data timely for users to track.

3) *Item Recommendation*: The item recommendation functions show users products that best fit personal demands. Crowdbate has top coupons and recommended products on the home page, where all users share the same information of

top coupons while the recommended products are personalized. Both features are fed with data from Crowdbate.

4) *Dashboard Visualization*: Crowdbate opens API and provides a visualized data dashboard to cooperating online shopping platforms. With the visualized dashboard, merchant users can access to internal data depicting user portraits, product sales, and coupon popularity.

### B. Crowdbate Algorithm

Crowdbate algorithm [7] can place orders for users’ best interests and works as one of the most crucial components in our platform. Considering the combination of coupons and delivery costs, Crowdbate proposed two algorithms, the order-first algorithm and the delivery-first algorithm, to adapt to different scenarios. Users can freely choose from these algorithms which best fits their needs, and their requests will wait in two pools to process separately.

**The Order-First Algorithm.** In most cases, coupons with strict thresholds usually guarantee tempting rebates that should way cover the delivery costs. Under such a scenario, we propose the order-first algorithm to place orders. Generally speaking, this algorithm greedily packs orders for a maximum amount of total rebates while ignoring delivery costs. For each coupon, we traverse requests in the eligible pool and calculate the differential rebate amount when compared to placing an order separately, and the request with the largest amount will be integrated into one order until for any request the differential rebate amount fails to increase. However, the number of coupons used in a newly merged order may exceed the upper limit since their amounts are calculated separately before the merger. We introduce *dif procedure* to tackle extra coupons while computing for the differential rebate. After packing requests, we need to place the order and assign one single address as the delivery destination. While this algorithm doesn’t take the delivery into account before, it assigns a warehouse to the order to minimize the cost of delivery now.

**The Warehouse-First Algorithm.** The order-first algorithm performs well in most cases; however, it still suffers from two drawbacks. The first one is due to its lack of consideration for delivery costs, and once the delivery cost gets much higher, this algorithm performs poorly. The second defect results from the time complexity introduced by the error correction *dif procedure*, which can be fatal when handling peak shopping festival traffic. As a result, a more efficient approach, the warehouse-first algorithm, is proposed to tackle the large scale data and non-negligible delivery costs. Instead of directly combining requests for higher rebates, we first group those requests sharing the same warehouse. Only then will we merge requests within the same warehouse for higher rebates. The preprocessing procedure divides the integral data set into smaller groups to constrain the batch size and running time, and grouping by delivery destination guarantees relatively small delivery costs. Basically, the warehouse-first algorithm is more efficient and can produce a better result for higher delivery costs, while the order-first algorithm stands out when rebates are profitable enough to cover delivery costs. Before

users can freely choose from algorithms above, we run the order-first algorithm compulsorily for the cold start since the scale of requests is too small to be split and should handle the higher time complexity while bringing users more rebates.

### C. Process Automation

Crowdrebate uses the process automation module to complete the whole procedure from picking items to pay for our server's orders. Basically, the module comprises two sub-modules, the web crawler retrieving the information of products and coupons, and the order bot to automatically place an order.

**The Web Crawler.** Currently, two types of information will be crawled, respectively, products and coupons. For the product information, this module retrieves the data passively. Before the user can post a request, a URL linked to the product must be provided to activate the web crawler. With the pasted URL, the spider bot crawls the associated coupons along with the product information and will specifically return options like size or color for the user to choose. Besides, we crawl all the metadata of this product for further recommendation function. For the coupon information, this module actively seeks data. We take the initiative to crawl information from coupon websites like Etao [8] to enrich our available coupon pools.

**The Order Bot.** The crowdrebate algorithm runs in rounds periodically and produces final grouped orders each round. Therefore we develop the order bot to handle those orders automatically. Since some exclusive coupons in different orders may violate each other, we introduce multiple accounts to place those exclusive orders, enabling us to run the scripts in parallel.

### D. Recommendation System

The recommendation system module is one of the critical features of Crowdrebate. Considering the online shopping scenario, there will not be a massive change for items on sale. The similarity of items is relatively stable to the user's interest, so we adapt the item-based collaborative filtering [9] to recommend similar products. The recommendation system is built upon metadata from the web crawler and users' historical requests and ratings. Since users on Crowdrebate pursue higher rebates, we improved the algorithm to rank on rebate amounts instead of similarities to meet their demands better.

## III. DEMONSTRATION OVERVIEW

In this section, we describe different kinds of scenarios that users commonly encountered in the Crowdrebate platform, which include a new request submission, the view of top deals and products, status checking, and data analyses.

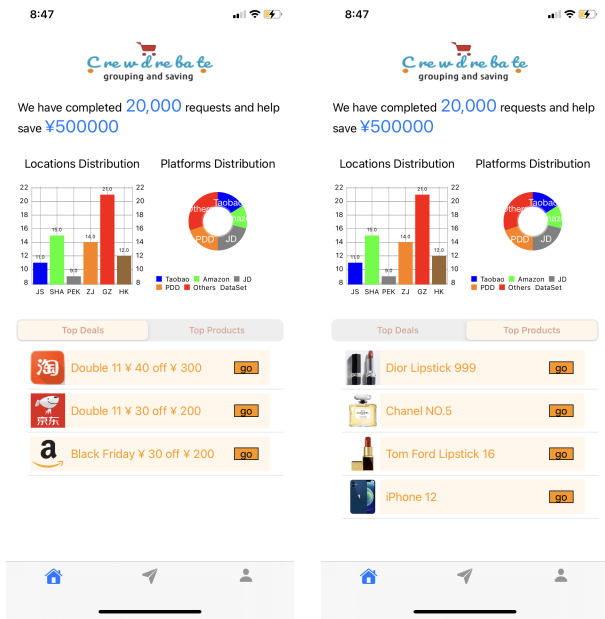
### A. Scenarios 1: Set up request

As a regular user completes the registration on our platform, has the certain product intent to buy, and clicks on the Set up new request button, as shown in Figure 3(a), she/he

Fig. 3: The flowchart of "Set up a request".

Fig. 4: View status of requests and personal data".

would enter the page which guides her/him to fill in the following information in Figure 3(b). The user needs to select the e-commerce platform she/he wishes to order from. In order to simplify the user's task and to prevent the platform from placing the wrong order of product, the user need to copy the URL link of the product he wants to buy, and the platform will automatically crawl the product information such as price, store, promotions, etc. If there is more than one Stock Keeping Unit SKU [10] for a product, we also crawl to all the SKU information and ask the user to make a choice. Next, the user needs to select the type of promotions he wants to participate in, such as whether to enjoy the coupons or to receive more free samples, and the time slot she/he wants the platform to place the order. Finally, the user fills in the delivery address. After confirming the final information twice shown in Figure 3(c), the platform will start to deal with the request at the time specified by the user.



(a) Today's deals

(b) Personalized recommendations

Fig. 5: View Today's deals and personalized recommendations.

#### B. Scenarios 2: Check request status and data

After completing to set up a request, the user can check all the requests she/he made on our platform on the personal page in Figure 4(a) and the status of each request, such as whether the order has been placed, and whether the product has been sent to the warehouse, etc.

Note that, if the user has interest in the data of order, she/he can also see specific data for each request in the My request data page (Figure 4(b)), like the original price, the price after the platform placed the order for them, the shipping cost, the platform's service charge, and the savings after deducting various costs. The user also can choose to see the sum of the data over a period of time.

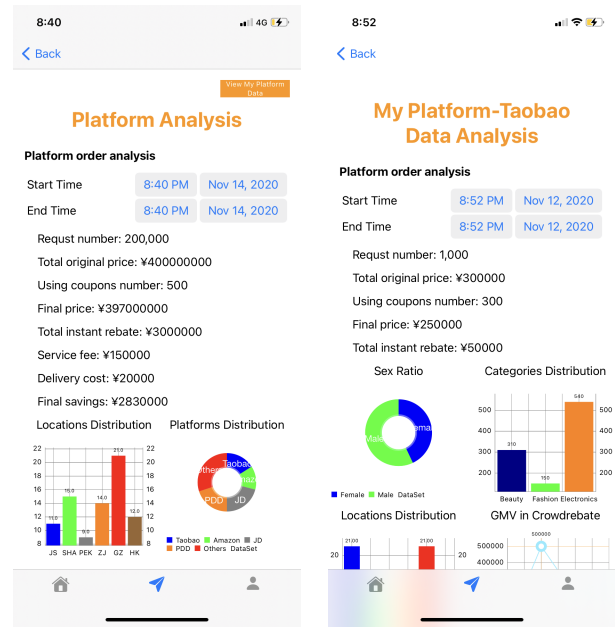
#### C. Scenarios 3: View Top deals and top products

When user are not sure about recent promotions of e-commerce platforms, she/he can use the home page in Figure 5(a) to see our daily updates of platforms' top deals to save time. While having no clear shopping target, the user can view the top products that our crowdrebate platform makes personalized recommendations based on the previous purchase data (Figure 5(b)).

#### D. Scenarios 4: Dashboard of data

**As teammates** of the Crowdrebate platform or **our cooperation partners**, in addition to the basic operations mentioned above, they can also carry out the following functions:

Without being interfered with the user privacy, our platform staffs can enter the third button Platform data (Figure 3(a)), where they will be able to see a dashboard of data related to the whole platform, such as how many orders we have received, how many coupons we have used, how much money we have helped users to save, the distribution of our users by regions,



(a) Dashboard of crowdrebate data

(b) Dashboard of cooperator data

Fig. 6: Dashboard of data.

and the distribution of platforms they like to launch request (Figure 6(a)).

As for our cooperation partners of e-commerce platforms, they can see a dashboard of data related to their platforms. For example, as Taobao cooperators, they can see how many of the requests initiated by our platform users are from Taobao, the distribution of categories they like to buy, the ratio of sex, etc shown in Figure 6(b). This dashboard of data will help our partners to gain a more practical understanding of user preferences, design coupons, and reach a win-win collaboration mechanism.

#### IV. CONCLUSION

We present the Crowdrebate platform, which collects requests from users, groups requests into a set of orders to get more rebates and delivers products to different receivers in an order. Crowdrebate is equipped with the techniques including the crowdrebat algorithm grouping orders for maximum rebates, web crawling, items recommendation, and data visualization. We also introduce the sketch of system architecture and demonstrate scenarios via several case analyses.

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