Pacing algorithm

Thomas Maurice





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1. Introduction

- Pacing algorithm
 Current algorithm
 New algorithm
 Spatial pacing
- Conclusion



Digital Out-Of-Home (DOOH) advertising is a marketing channel where an ad is dynamically and digitally displayed in outdoor public spaces.



Figure: Advertising in a mall



- ► Main intermediary between supply (publishers) and demand (advertisers)
- ► Programmatic purchasing platform
 - Booking
 - 2 RTB



- ► In open RTB: Displayce buys impressions in real time
- For each bid request, the bidder chooses whether to buy or not
- Importance of a pacing algorithm to smooth the budget expenditure

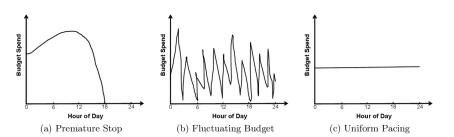


Figure: Examples of expenditures



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Data

► To build the algorithm, we will use a data set of the following form:

		ID	Imps	Price	Win	Seconds notif
2020-07-08 (06:00:00	1.0	10.0	10.0	True	365.0
2020-07-08 (06:00:00	2.0	1.0	1.0	True	361.0
2020-07-08 (06:00:01	3.0	5.0	5.0	True	555.0
2020-07-08 (06:00:01	4.0	1.0	1.0	True	645.0
2020-07-08 (06:00:03	5.0	1.0	1.0	True	310.0
2020-07-08 (06:00:03	6.0	2.0	2.0	True	355.0



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- ► The simplest algorithm is a hourly capping
- Cons:
 - All the budget is spent at the beginning of the hour

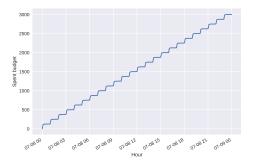


Figure: Hourly capping



- ► The simplest algorithm is a hourly capping
- ► Cons:
 - All the budget is spent at the beginning of the hour
 - 2 No catch-up possible



Figure: Hourly capping



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- ► Calculate a budget per second to have a uniform expenditure over the day
- Budget per second is recalculated each time a bid request is received
- lacktriangle We can calculate the budget per second noted b_t as follow:

$$b_{t+1} = (B - \sum_{s=1}^{t} S(s)) \frac{1}{T - t}$$

Where b_{t+1} is the budget to be allocated to the second t+1, B is the total budget for the day, S(s) is the actual expenditure to the second t and finally T-t is the estimated seconds until the end of the last bid request.

- ► We need to determine if there is a slowdown in the frequency of reception of impressions
- ▶ The budget per second can be improved by taking into account its against time variation. If b_t increases, it means that less and less impressions are bought
- ▶ We can calculate the variation $v_t = b_t b_{t-1}$ and the speed of the variation $a_t = v_t v_{t-1}$. We denote \bar{v} and \bar{a} respectively the average variation of b_t over the last 30 minutes and the average speed in the variation of b_t
- ▶ The calculation of b_t becomes:

$$b_{t+1} = \left(B - \sum_{s=1}^{t} S(s)\right) \frac{1 + (\bar{a}\bar{v})}{T - t}$$



Uniform pacing

► The expenditure over the day is much smoother and more uniform than the hourly capping

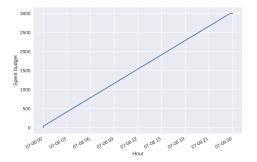


Figure: Per second budget



- ► The unspent budget from the previous hours is caught up at the end of the day contrary to the hourly capping
- ► Withdraw: the expenditure is not uniform over the day. The slowdown has to be predicted in order to smooth the expenditure over the day

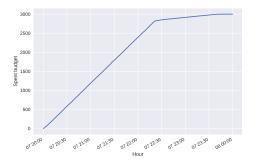




Figure: Per second budget

Predict the proportion of impressions received per hour by using historical data and performing a simple linear regression

$$Prop_i = \alpha + \beta_k \mathbf{X_k} + \varepsilon_i$$

Where β_k is the vector of coefficients associated with the vector of explanatory variables.

▶ The formula of b_t then becomes:

$$b_{t+1} = (B(h) - \sum_{s=1}^{t} S(s)) \frac{1 + (\bar{a}\bar{v})}{T - t}$$

Where B(h) is the budget allocated to the hour h, $\sum_{s=1}^{\circ} S(s)$ is the budget spent in the current hour h

Evolutive hourly budget

► The algorithm catches up with the unspent budget much more uniformly than the previous one.

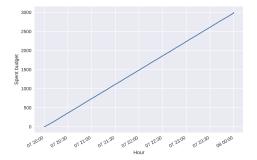


Figure: Evolutive hourly budget



Comparison

We can compare 3 algorithms over the same dataset. The third algorithm (red curve) catches-up the expenditure and do it uniformly.

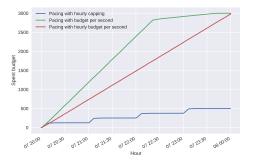


Figure: Expenditure comparison



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- ► Account for the algorithm the possibility to add a spatial dimension to the pacing
- ▶ Difficult time zones integration
- ► Initially allocate a budget per time zone and run the time pacing algorithm independently for each time zone
- ► Create a meta-class that will create its own instances of the pacing class for each time zone
- ► Problem: how much budget can we allocate for each time zone?



- ► Initialization: the algorithm distribute the budget equally among the time zones.
- ▶ The objective is to adjust in real time the budget allocated
- ► Each bid request received, the algorithm calculates the proportion of bid request bought. If it is 100%, we need to reallocate the budget.
- Numerical example: Let T_1 , T_2 , T_3 and T_4 stand for 4 time zones. Total budget: €1000. At the very beginning: $T_1 = T_2 = T_3 = T_4 = 250$. If T_1 buys 100% and we expect it to buy only half of the original objective: we reallocate 50% of €250 to the other time zones, namely, $\frac{125}{3}$ for each time zone.

Then,
$$T_1=125$$
 and $T_2=T_3=T_4=250+\frac{125}{3}$



- ► The following graph shows the pacing algorithm on 11 time zones
- ▶ The algorithm has reallocated the budget between time zones

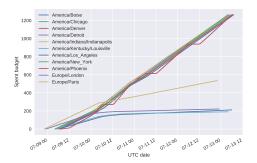


Figure: Expenditure per time zone



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- ► Implementation of a pacing algorithm that better meets the objectives:
- ► Integration of the possibility to have a pacing between time zones
- ► What's next?
 - 1 Improve the performance of the algorithm
 - 2 Provide an API to integrate the algorithm in production environment
- ► What I have gained personally:
 - Progress in research methodology
 - 2 Improvement of Python skills



END

