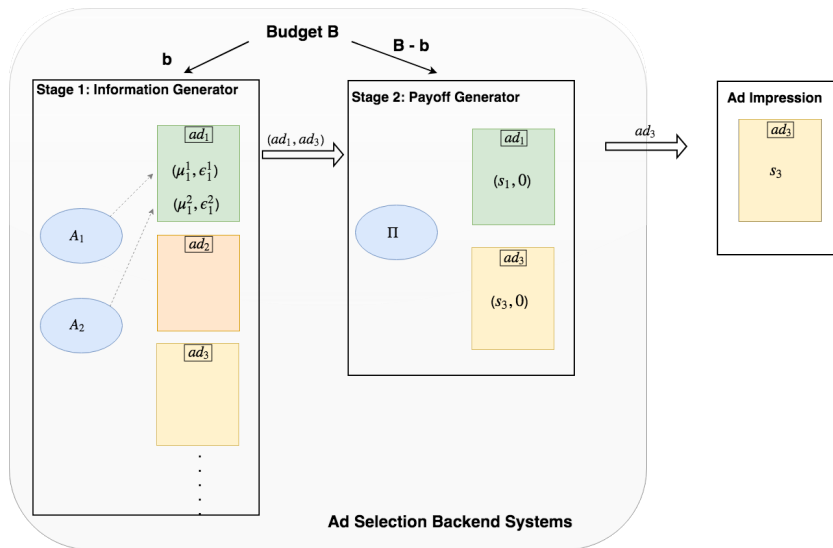


# Information Production and Analysis in Large-Scale Budget-Constrained Systems

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# Real-time ad selection system



## Some notation

**$B$** : Total budget

**$s_i$** : The value from displaying ad  $i$ ;

**$\Omega$** : Candidate set of ads ( $|\Omega| = N$ )

**$\mathcal{I}_0$** : Initial information about  $\Omega$  ( $\mathcal{I}_0 = F_0^N$ ,  $F_0$  is the prior cdf)

**$A(i, \lambda, \epsilon_i)$** : The *information generator* applied to ad candidate  $i$  with cost  $\lambda$  and error  $\epsilon_i$ <sup>1</sup>

**$\Pi(\Omega')$** : The *payoff-generator* or the final stage algorithm with unit cost and zero error evaluating  $\Omega' \subset \Omega$

**$b$** : Budget allocated to the information generator ( $A$  can evaluate  $\frac{b}{\lambda}$  elements, and  $|\Omega'| = B - b$ )

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<sup>1</sup>A reduces the uncertainty regarding payoff  $s_i$  but not perfectly, and the level of reduction is heterogeneous.

## Pay-off function

The payoff is the **highest order statistic** amongst the candidates evaluated by  $\Pi$ .


In a single slot auction:

$$\Pi(\Omega') = \max_{w \in \Omega'} s_i \quad \text{s.t.} \quad w = ad_i$$

## Some quick comments and observations

- (Trivial) One element to be evaluated by information generator
- $\lambda \in [0, 1]$ <sup>2</sup>
- The algorithm  $A$  produces a point estimate  $\gamma_i$  for ad  $i$ , such that:
  - $s_i \in [\gamma_i - \epsilon_i^1, \gamma_i + \epsilon_i^1]$ , where  $\epsilon_i$  captures the inaccuracy of the information generator, and this inaccuracy is heterogeneous across the ads
  - $\mathbb{E}[\gamma_i] = s_i$ , or the algorithm  $A$  is “calibrated” in expectation.

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<sup>2</sup> $\lambda = 0$  is updated prior and  $\lambda = 1$  is the information generator 

## Questions we'll comment on

- **Free evaluation:** Given  $I = I_0 \cup A(0, \dots)$  and  $|\Omega'| = N - 1$ , should  $ad_0 \in \Omega'_*$ , the optimal subset to be evaluated by the payoff generator.
- **Single evaluation:**  

$$\mathbb{E} \left[ \Pi(\Omega'_*(I_0)) \right] \geq \mathbb{E} \left[ \Pi(\Omega'_*({I_0 \cup A(0, \lambda, \dots)})) \right]$$
 where the latter allows only  $N - \lambda$  elements to be evaluated.
- **Optimal subset:** Given fixed  $B - b$  and  $I$ , how does one choose  $\Omega'_*$  which maximizes  $\mathbb{E} \left[ \Pi(\Omega') \right]$  over all  $B - b$  cardinality subsets.

## Questions we'll comment on

- **Budget allocation:** For a given information generator  $A(\lambda, \epsilon)$ , what should the budget allocation between the information generator ( $b_*$ ) and the payoff generator ( $B - b_*$ ).
- **Learning algorithm:** Given an array of information generators, which one would you choose?

## Quick Primer on Order Statistics

For  $X_1, X_2, \dots, X_n$  are iid continuous random variables with pdf  $f$  and cdf  $F$ , the density of the highest order statistics is:

$$P(X_{(n)} \in [x, x + \epsilon]) = \epsilon \times nf(x)F(x)^{n-1}$$

Order statistics of standard uniforms with  $n$  uniform random variables  $X_k \sim \text{Beta}(k, n - k + 1)$  and in particular  $\mathbb{E}[X_1] = \frac{n}{n+1}$



## Some simple cases

which subset to evaluate?

- Ten  $U[0, 1]$  or ten  $U[0.85 - \epsilon, 0.85 + \epsilon]$
- Case  $U[0.4, 0.7]$  where  $N = 1$  and  $N = 3$
- For the optimal solution to consider all subsets of  $\Omega$

## Monotone Sub-modular

Let  $X$  be a finite set. A function  $f : 2^X \rightarrow \mathbb{R}$  is:

- **monotone** if for all  $S \subset T$ ,  $f(S) \leq f(T)$
- **submodular** if for all subsets  $S \subset T \subset X$  and all  $x \in X \setminus T$

$$f(S \cup \{x\}) - f(S) \geq f(T \cup \{x\}) - f(T) \quad (1)$$

Does this apply to our payoff function?

## Greedy Algorithms are good enough?

**Theorem:** Let  $f$  be a monotone, submodular, non-negative function on  $X$ . The greedy algorithm, which starts with  $S$  as the empty set and at every step picks an element  $x$  which maximizes the marginal benefit ( $f(S \cup \{x\}) - f(S)$ ) provides a set  $S$  that achieves a  $(1 - \frac{1}{e})$  approximation of the optimum.<sup>3</sup>

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<sup>3</sup>M. Fischer, G. Nemhauser, L. Wolsey. An analysis of approximations for maximizing submodular set functions.

## Budget allocation

If the "value of the information" generated by  $A$  is monotonic in the  $b$ , then the optimal budget allocation should be until:

$$\mathbb{E} [\Pi[I_0 \cup A(b + \lambda)]] < \mathbb{E} [\Pi[I_0 \cup A(b)]]$$

where the cardinality of the former set to the payoff generator is  $B - b - \lambda$  while  $B - b$  for the later set.

*But... how do you compute the "marginal value of additional information?"*

## Moving to uniform distributions

- At the start of the evaluation process the  $s_i$  are treated as independent random variables, identically distributed over an interval  $U[0, 1]$  with uniform density.
- Given  $ad_i$ ,  $A$  produces an interval  $[\alpha_i; \beta_i]$  over which  $s_i$  is uniformly distributed. In particular:
  - $\mathbb{E}[s_i] = \frac{\alpha_i + \beta_i}{2}$  or the algorithm is *calibrated*

## When would you include the evaluated point?

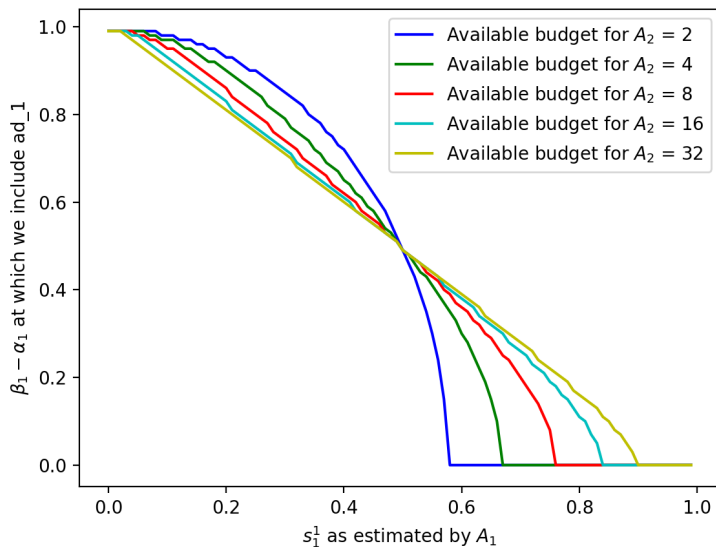
If a budget  $B$  is available for  $\Pi$  after  $A$  generates  $[\alpha_i, \beta_i]$  for ad  $i$ , and if  $\mathcal{I}' = U[0, 1]^B \cup \{U[\alpha_i, \beta_i]\}$ , then

$$\mathbb{E} [\Pi(B, \mathcal{I}')] = \underbrace{\frac{B-1}{B}}_{\mathbb{E}[\Pi] \text{ for } B-1 \text{ ads}} + \underbrace{\frac{1}{B(B+1)}}_{\text{loss from 1 less ad}} \times \underbrace{\frac{\beta_i^{B+1} - \alpha_i^{B+1}}{\beta_i - \alpha_i}}_{\text{gain due to } i}$$

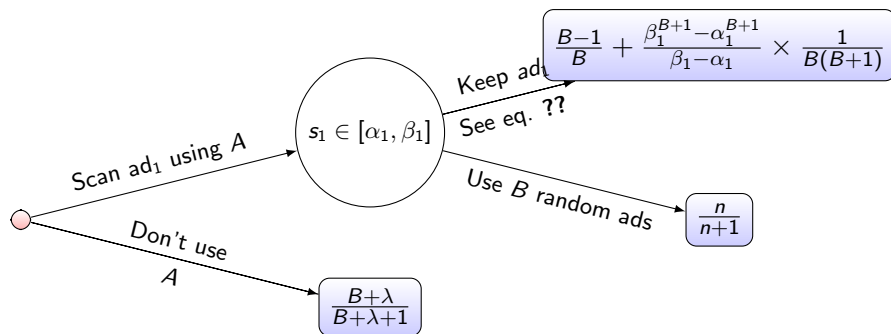
and the condition for including ad  $i$  is

$$\mathbb{E} [\Pi(B, \mathcal{I}_0)] > \mathbb{E} [\Pi(B, \mathcal{I}')] \iff \frac{\beta_i^{B+1} - \alpha_i^{B+1}}{\beta_i - \alpha_i} > 1.$$

# When would you include the evaluated point?



## Decision tree for one element?



**Figure:** Decision tree for assigning a single unit of computation to A, and then deciding how to leverage the new information  $[\alpha_i, \beta_i]$ .



## Summary

- Cost / accuracy trade-off in real-time ad selection systems.
- Simply picking elements with highest expected value might be naive, sub-optimal.
- There exist approximate algorithms even in the setting when ad candidates are processed as a stream.
- The budget allocation should balance marginal utility of the information with the marginal cost of the information.

questions and suggestions?