

6998-2 Advanced Machine Learning for Personalization

Home Work #2

va2361: Vishal Anand

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Contextual bandit: LinUCB

Max Percentage: **94.762%**

α search			
α	Cumulative-Take-Rate	Reward	Match
$1/\sqrt{t}$	0.91763	947	1032
$0.1/\sqrt{t}$	0.94589	979	1035
$0.09/\sqrt{t}$	0.94762	977	1031
$8.08/\sqrt{t}$	0.94547	971	1027
$0.085/\sqrt{t}$	0.94731	971	1025
$0.095/\sqrt{t}$	0.94762	977	1031
$10.088/\sqrt{t}$	0.60176	615	1022
$0.5088/t$	0.91185	931	1021
$5.5088/t$	0.93641	972	1038
$3.6088/t$	0.93695	966	1031
$50.6088/t$	0.91023	943	1036
$50.6088/2t$	0.91779	949	1034
$25.6088/(\sqrt{t} + 2t)$	0.93030	961	1033
$10.6088/(\sqrt{t} + 2t)$	0.93743	974	1039
$18.6088/(\sqrt{t} + t^3)$	0.92210	947	1027
$0.085/t^{0.25}$	0.92843	960	1034
$0.0085/t^{0.25}$	0.94680	979	1034
$0.0085/t^{0.125}$	0.94731	971	1025
$0.0085/t^{0.2975}$	0.94498	979	1036
$0.0085/t^{0.2999}$	0.94083	970	1031

Table 1: Alphas and Cumulative-Take-Rates

Algorithm 1 LinUCB with disjoint linear models **modified for current dataset**

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1: Inputs:  $\alpha \in \mathbb{R}_+$ 
2: for  $t = 1, 2, 3, \dots, T$  do
3:   Observe features of all arms  $a \in \mathbb{A}_t : x_{t,a} \in \mathbb{R}^d$ 
4:   for all  $a \in \mathbb{A}_t$  do
5:     if  $a$  is new then
6:        $A_a \leftarrow I_d$  ( $d$ -dimensional identity matrix)
7:        $b_a \leftarrow 0_{d \times 1}$  ( $d$ -dimensional zero matrix)
8:     end if
9:      $\hat{\theta} \leftarrow A_a^{-1} b_a$ 
10:     $p_{t,a} \leftarrow \hat{\theta}^T x_t + \alpha \sqrt{x_t^T A_a^{-1} x_t}$ 
11:  end for
12:  Choose arm  $a_t = \arg \max_{a \in \mathbb{A}_t} p_{t,a}$  with ties broken arbitrarily, and observe a real-
    valued payoff  $r_t$ 
13:   $A_{a_t} \leftarrow A_{a_t} + x_t x_t^T$ 
14:   $b_{a_t} \leftarrow b_{a_t} + r_t x_t$ 
15: end for

```

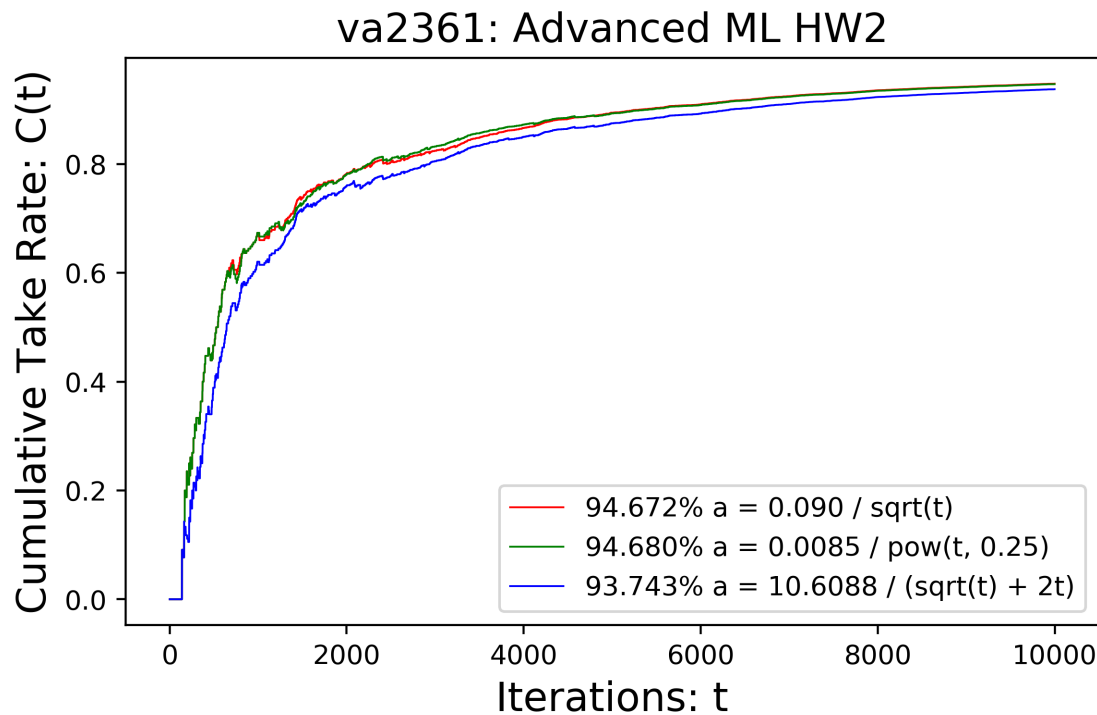
Algorithm 2 Policy evaluator (with finite data stream)

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1: bandit algorithm A; stream of events  $S$  of length  $L$ 
2:  $h_0 \leftarrow \emptyset$  An initially empty history
3:  $\hat{\mathbb{G}}_A \leftarrow 0$  An initially zero total payoff
4:  $T \leftarrow 0$  An initially zero counter of valid events
5: for  $t = 1, 2, 3, \dots, T$  do
6:   Get the  $t$ -th event  $(x, a, r_a)$  from  $S$ 
7:   if  $A(h_{t-1}, x) = a$  then
8:      $h_t \leftarrow \text{CONCATENATE}(h_{t-1}, (a, a, r_a))$ 
9:      $\hat{\mathbb{G}}_A \leftarrow \hat{\mathbb{G}}_A + r_a$ 
10:     $T \leftarrow T + 1$ 
11:     $b_a \leftarrow 0_{d \times 1}$  ( $d$ -dimensional zero matrix)
12:  else
13:     $h_t \leftarrow h_{t-1}$ 
14:  end if
15: end for
16: Output:  $\hat{\mathbb{G}}_A / T$ 

```

Graphs



Algorithms adapted from [1] and [2]

Code is attached with the zip file, and the same is also at <https://github.com/vishalanand/6998-HW2-linUCB>

References

- [1] Lihong Li, Wei Chu, John Langford and Robert E. Schapire *A Contextual-Bandit Approach to Personalized News Article Recommendation*
<https://arxiv.org/abs/1003.0146>
- [2] Lihong Li, Wei Chu and John Langford *An Unbiased, Data-Driven, Offline Evaluation Method of Contextual Bandit Algorithms*
<https://arxiv.org/abs/1003.5956>