Seminar in Deep RL

Multi Armed Bandits-Contextual Bandits

Orhan Saeedi 20.04.2020





What we saw so far!

k-armed Bandits (in this case k= 2)



Exploration vs.



William R. Thompson's Thompson Sampling







One-Armed Bandit



Cost: \$

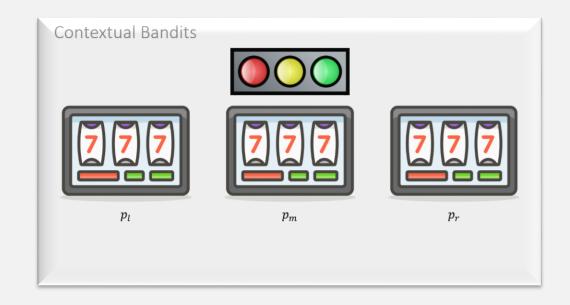






Context Free vs. Context Based/Contextual Bandits







Context Free Bandits







 p_l

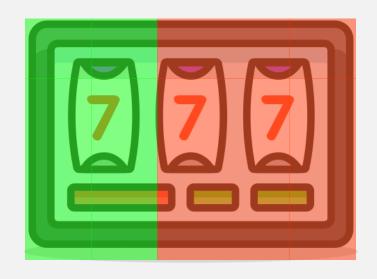
 p_m

 p_r

Context Free Bandits



$$p_l = 0.5$$



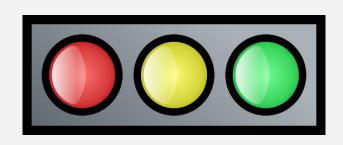
 $p_m = 0.4$



 $p_r = 0.55$



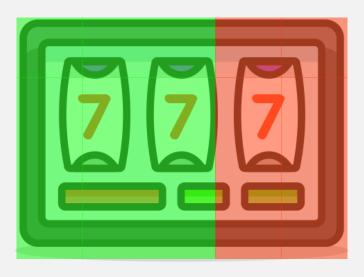


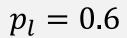


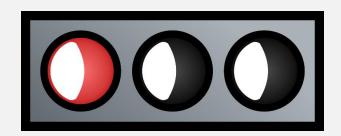


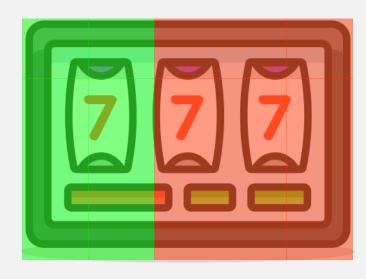


 p_l p_m

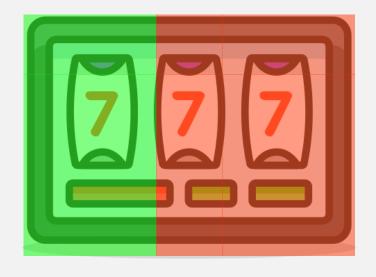






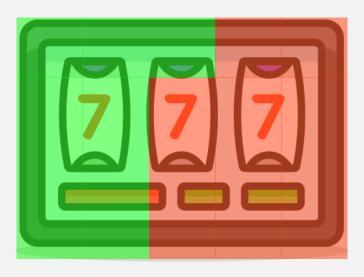


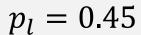
$$p_m = 0.4$$

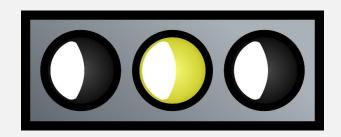


$$p_r = 0.4$$



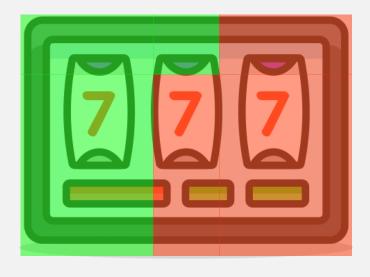








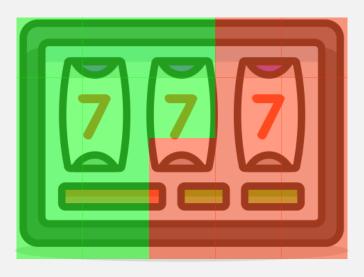
$$p_m = 0.45$$

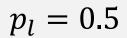


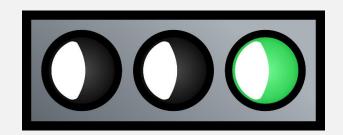
 $p_r = 0.45$

11



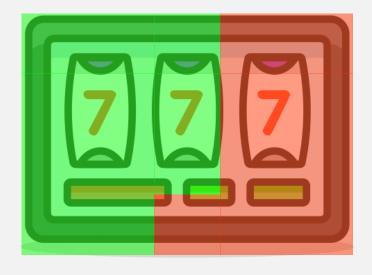








$$p_m = 0.4$$



$$p_r = 0.55$$

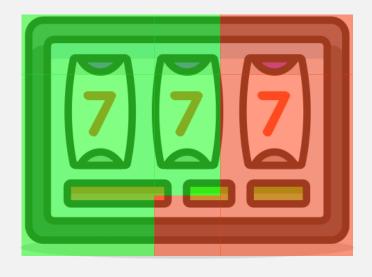




$$p_l = 0.5$$



 $p_m = 0.4$



 $p_r = 0.55$

LOG IN

Today's Paper

With a Doctor in a

Disaster Mode'

Brooklyn E.R.: 'We're in

Test kits and protective

gear have been in short

sick, and every day gets

keeps showing up.

413 comments

supply, doctors are falling

more difficult. But the staff

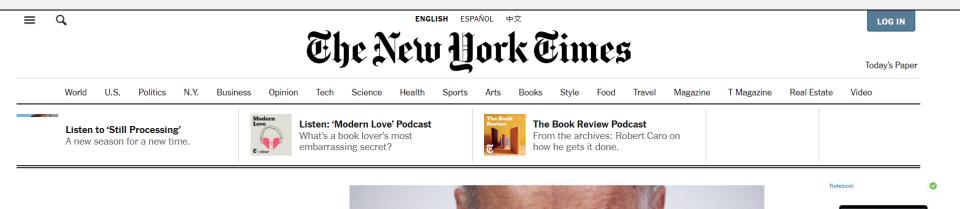


Personalized News Article Recommendation





Personalized News Article Recommendation







Biden Is Losing the Internet. Does That Matter?

The coronavirus has forced the Democrats' presumptive presidential nominee into an all-digital campaign, and he's struggling to break through.

21h ago · By KEVIN ROOSE



Apple Pro Display XDR -

Standard glass

Apple Pro Display XDR -

Standard glass

Welcome to the 'Rabbit Hole'

Introducing an audio series about how the internet is changing, and how it's changing us.

14h ago · By KEVIN ROOSE



Facebook-Backed Libra Cryptocurrency Project Is Scaled Back

After months of criticism, the cryptocurrency is moving ahead with significant revisions.

18h ago * By NATHANIEL POPPER and MIKE ISAAC



20h ago · By SHIRA OVIDE

FOMO Has Survived the Coronavirus

No one is going anywhere cool, and yet I still feel left out.







Personalized News Article Recommendation







Stocks Surge as Washington's Aid Package Advances: Live

Updates

Live updates on stock market and business news during the coronavirus outbreak. 5m ago



How the Fed's Magic Money Machine Will Turn \$454 Billion Into \$4 Trillion

The central bank takes Treasury Department loan guarantees and uses them to stand up huge programs. Here's how that works. 7h ago · By JEANNA SMIALEK

'It's a Wreck': 3.3 Million File **Unemployment Claims** as Economy Comes Apart

The weekly figure is among the first data on the economic toll of the vast disruption of normal life and commerce caused by the coronavirus pandemic.

3h ago * By BEN CASSELMAN, PATRICIA COHEN and TIFFANY HSU

Surging Traffic Is Slowing Down Our Internet

With people going online more in the pandemic, internet traffic has exploded. That's taking a toll on our download speeds and video

9h ago * By CECILIA KANG, DAVEY ALBA and ADAM SATARIANO



Personalized News Article Recommendation





ON BASEBAL

It's Opening Day. Baseball Is Closed.

Instead of red, white and blue bunting and crowds flooding ballparks, we have players holed up in homes and endless questions about when baseball will return. When it does, it will be more important than ever.

7h ago * By TYLER KEPNER

Coronavirus Protective Masks to Be Made From M.L.B. Uniform Material



The deal will send masks to health professionals in the Northeast with the distinctive pinstripes of the Yankees and the Philadelphia Phillies.

2h ago · By DAVID WALDSTEIN

W.N.B.A. Will Hold 'Virtual' Draft in April



The league is keeping its scheduled date of April 17 but adapting the format to allow draftees to have their moment in the spotlight no matter where they are.

7h ago · By HOWARD MEGDAL

Baseball Season Opens With a Dash of Imagination



Complaints of baseball's slow pace yielded to the universal wish that this game would last forever.

4h ago · By DAN BARRY



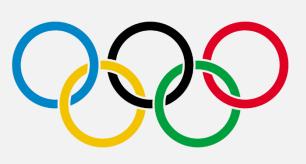


Context















Problem Protocol: Context Free Bandits

```
For each round t \in [T]:

ALG picks an Arm/Action a_t \in A.

Reward r_t is realized.
```



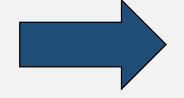
One Round: Context Free Bandits

Arm/Action $a_t \in A$

ALG







Reward r_t

Exp. Reward $\mu(a_t)$



Problem Protocol: Contextual Bandits

```
For each round t \in [T]:

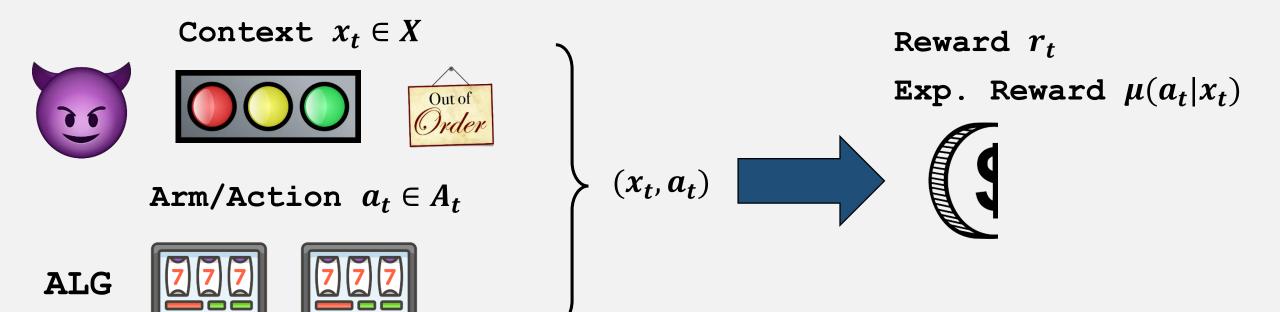
ALG observes a Context x_t \in X.

ALG picks an Arm/Action a_t \in A_t.

Reward r_t is realized.
```

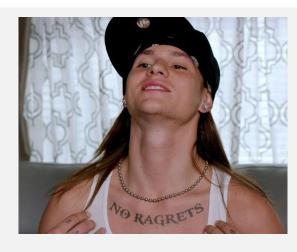


One Round: Contextual Bandits





Expected Regret: Context Free Bandits

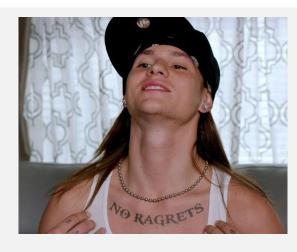


$$\mathbb{E}[R(T)] = \mu^* \cdot T - \sum_{t=1}^{l} \mu(a_t)$$

with
$$\mu^* \coloneqq \max_{a \in A} \mu(a)$$



Expected Regret: Contextual Bandits



$$\mathbb{E}[R(T)] = \text{REW}(\pi^*(x)) - \sum_{t=1}^{r} \mu(a_t|x_t)$$

with Best Response Policy:
$$\pi^*(x) = \max_{a \in A} \mu(a_t|x_t)$$

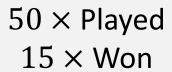


Recap: Upper Confidence Bound

$$r_t(a) = \sqrt{\frac{2 \ln T}{n_t(a)}} = \sqrt{\frac{2 \ln 170}{n_t(a)}}$$

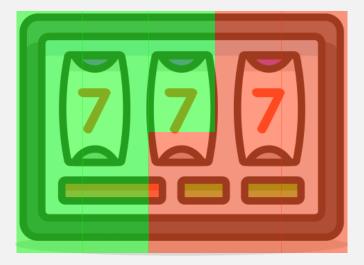
$$100 \times Played$$

 $60 \times Won$



$$20 \times Played$$

 $9 \times Won$





$$p_{l} = 0.5; \mu(a_{l}) = 0$$

$$\bar{\mu}(a_{l}) = 0.2$$

$$r_{t}(a_{l}) \approx 0.32$$

$$0.52$$

$$p_m = 0.4; \mu(a_m) = -0.2$$

$$\bar{\mu}(a_m) = -0.4$$

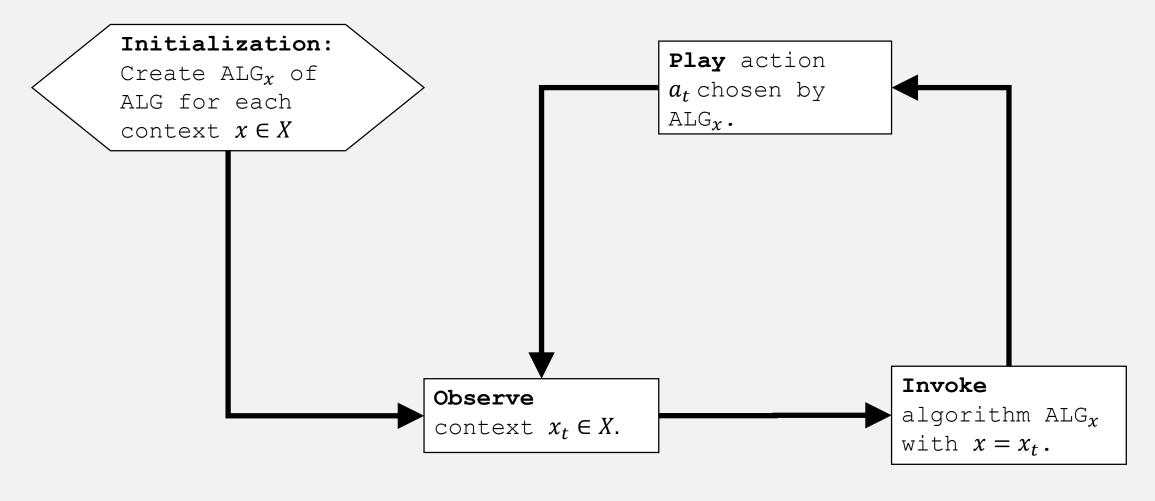
$$r_t(a_m) \approx 0.45$$
0.05

$$p_r = 0.55; \mu(a_r) = 0.1$$

 $\bar{\mu}(a_r) = -0.1$
 $r_t(a_r) \approx 0.71$ 0.61



Run separate copy of ALG for each context



Example for ALG: Upper Confidence Bound/UCB1



Regret Analysis with ALG=UCB1

$$\mathbb{E}[R(T)] = \sum_{x \in X} \mathbb{E}[R_x(T)] = \sum_{x \in X} O\left(\sqrt{Kn_x \log T}\right) \le O\left(\sqrt{KT|X|\log T}\right)$$

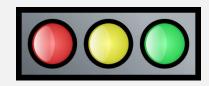
Regret of UCB1



Regret Analysis with ALG=UCB1

$$\mathbb{E}[R(T)] = O(\sqrt{KT|X|logT})$$

Good for small number of contexts:



Bad for **large** number of contexts:

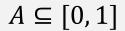








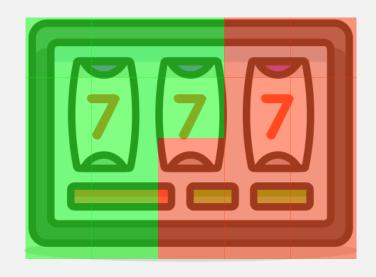
Quick Recap: Lipschitz Bandits





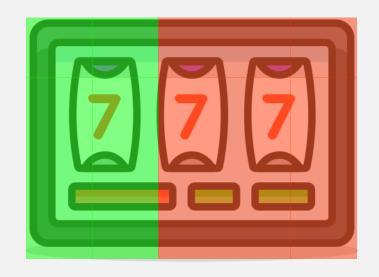
 $\in [0,1]$

OAB₁



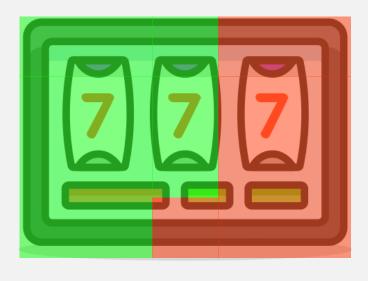
 $p_l = 0.5$

OAB2



 $p_m = 0.4$

OAB1



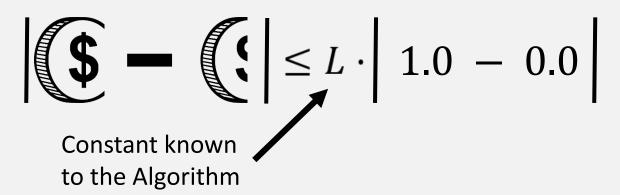
$$p_r = 0.55$$



Quick Recap: Lipschitz Bandits

Rewards satisfy a Lipschitz condition:

$$|\mu(x) - \mu(y)| \le L \cdot |x - y|$$
 for any two arms $x, y \in A$,





Quick Recap: Lipschitz Bandits

Simple Solution: **Discretization**Solve for $S \subseteq A$ with off-the-shelf MAB algorithm
(i.e. UCB1)



$$X \subseteq [0,1]$$















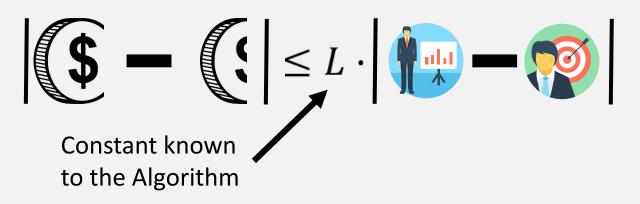






Rewards satisfy a Lipschitz condtion:

 $|\mu(a|x) - \mu(a|x')| \le L \cdot |x - x'|$ for any two contexts $x, x' \in X$, where L is the Lipschitz constant known to the ALG





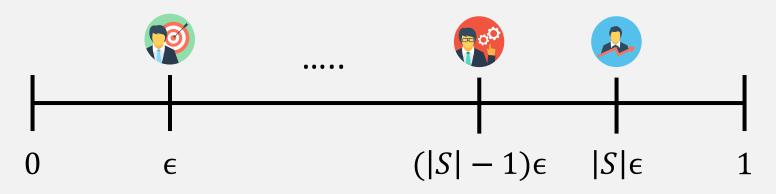
Simple Solution: **Discretization (Context)** Solve for $S \subseteq X$ with off-the-shelf MAB algorithm (i.e. UCB1)



Context Free vs. Contextual Lipschitz Bandits





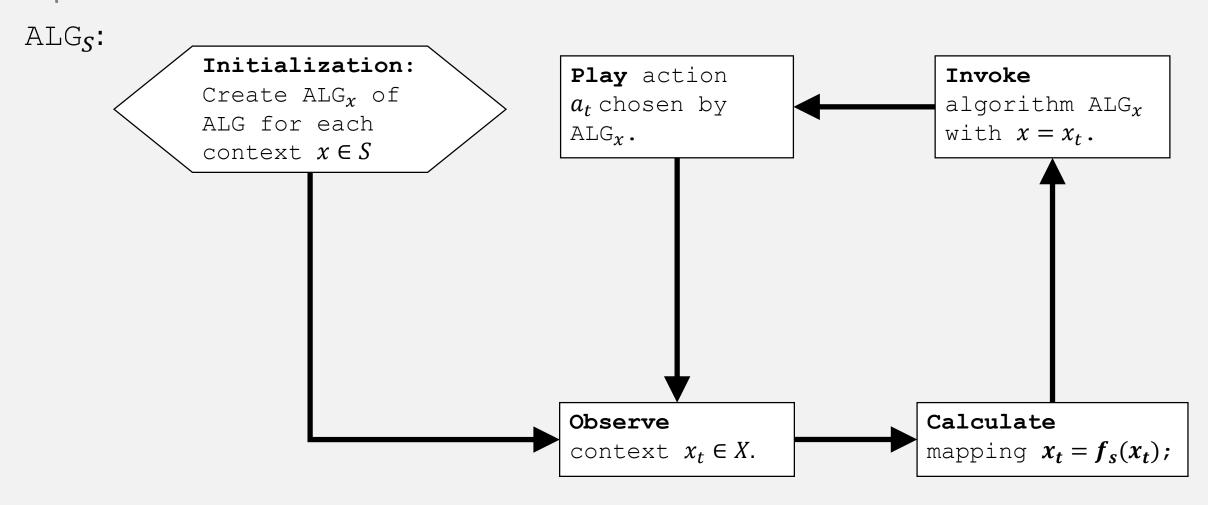


Let S be the ϵ -uniform mesh with $\epsilon = \frac{1}{|S|+1}$ and apply ALG for S

Let $f_S(x)$ be a mapping from context x to the closest point in S: $f_S(x) = \min (\arg \min_{x' \in S} |x - x'|)$



Lipschitz Contextual Bandits - Discretization



Example for ALG: Upper Confidence Bound/UCB1

Regret Analysis – Lipschitz (ALG = UCB1)

Let us define the **Discretized Best Response** $\pi_S^*: X \to A$ $\pi_S^*(x) = \pi^*(f_S(x))$ for each context $x \in X$

Regret of ALG_S : $R_S(T) = REW(\pi_S^*) - REW(ALG_S)$:

Discretization Error: $DE(S) = REW(\pi^*) - REW(\pi_S^*)$

Overall Regret: $R(T) = R_S(T) + DE(S)$

Overall Expected Regret: $\mathbb{E}[R(T)] = \mathbb{E}[R_S(T)] + \mathbb{E}[DE(S)]$

Regret Analysis – Lipschitz (ALG = UCB1)

Overall Expected Regret: $\mathbb{E}[R(T)] = \mathbb{E}[R_S(T)] + \mathbb{E}[DE(S)]$

Expected Regret: $\mathbb{E}[R_S(T)] = O\left(\sqrt{KT|S|\log T}\right) = O\left(\sqrt{\frac{1}{\epsilon}KT\log T}\right)$

Exp. Discretization Error: $\mathbb{E}[\mathrm{DE}(S)] = \mathrm{REW}(\pi^*) - \mathbb{E}[\mathrm{REW}(\pi_S^*)] \le \epsilon \mathrm{LT}$

 $\mu(\pi_S^*(x)|f_S(x)) \ge \mu(\pi^*(x)|f_S(x))$ Optimality of $\pi_S^*(x)$

 $\geq \mu(\pi^*(x)|x) - \epsilon L$

Lipschitz property

 $\mathbb{E}[\text{REW}(\pi_S^*)] \ge \text{REW}(\pi^*) - \epsilon LT$



Regret Analysis – Lipschitz (ALG = UCB1)

$$\mathbb{E}[R(T)] \le O\left(\sqrt{\frac{1}{\epsilon}KT\log T}\right) + \epsilon LT = O(T^{\frac{2}{3}}(LK\log T)^{\frac{1}{3}})$$

Linear Contextual Bandits: LinUCB algorithm

WWW 2010 • Full Paper

April 26-30 • Raleigh • NC • USA

A Contextual-Bandit Approach to Personalized News Article Recommendation

Lihong Li[†], Wei Chu[†],

[†]Yahoo! Labs

lihong,chuwei@yahooinc.com

John Langford[‡]

[‡]Yahoo! Labs
jl@yahoo-inc.com

Robert E. Schapire

+Dept of Computer Science
Princeton University
schapire@cs.princeton.edu



Contextual Bandits in Personalized News Article Recommendation



Context:

 $x_{t,1}$

 $x_{t,2}$

 x_{t,K_t}



Dancing With a Black Hole

Astronomers described the strange orbit of a star that loops the monster in the Milky Way, offering more evidence for one of Einstein's ideas.

Science April 17



Is the virus on my clothes? My shoes? My hair? My newspaper?

We asked the experts to answer questions about all the places coronavirus lurks (or doesn't). You'll feel better after reading this.

Live April 18 1085 comments



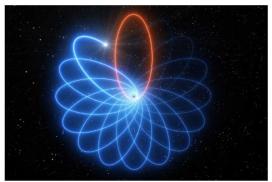
Global: Dozens Test Positive in Afghan Presidential Palace

The Afghan president is isolating. About 100,000 people defied a lockdown in Bangladesh. In Tokyo, our correspondent finds a city seemingly under a spell.

Live just now



Contextual Bandits in Personalized News Article Recommendation



I. Calcada/Furonean Southern Observator

Dancing With a Black Hole

Astronomers described the strange orbit of a star that loops the monster in the Milky Way, offering more evidence for one of Einstein's ideas.

Science April 17



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 $a_t \in A_t$



Global: Dozens Test Positive in Afghan Presidential Palace

The Afghan president is isolating. About 100,000 people defied a lockdown in Bangladesh. In Tokyo, our correspondent finds a city seemingly under a spell.

Live just now

Reward r_{t,a_t} : Clicked (1) or Not Clicked (0)



Linear Contextual Bandits

Expected reward is linear in:

$$\mu(a|x_{t,a}) = x_{t,a}^T \theta_a^*$$
 for all arms a and contexts x

$$\theta_a = (D_a^T D_a + I_d)^{-1} b_a$$

Training input:

$$D_a \in \mathbb{R}^{m \times d}$$









Is the virus on my clothes? My shoes? My hair? My newspaper?

We asked the experts to answer questions about all the places coronavirus lurks (or doesn't). You'll feel better after reading this.

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Response vector:

$$b_a \in \mathbb{R}^{\mathrm{m}}$$

Click

Click

No-Click



Confidence Region

When components in b_a are independent to rows of D_a , it can be shown [1] that with probability at least $1 - \delta$:

$$|x_{t,a}^T \theta_a - \mu(a|x_{t,a})| = \alpha \sqrt{x_{t,a}^T (D_a^T D_a + I_d)^{-1} x_{t,a}}$$

for any
$$\delta>0$$
 and $x_{t,a}\in\mathbb{R}^{\mathrm{d}}$ where $\alpha=1+\sqrt{\frac{\ln\left(\frac{2}{\delta}\right)}{2}}$

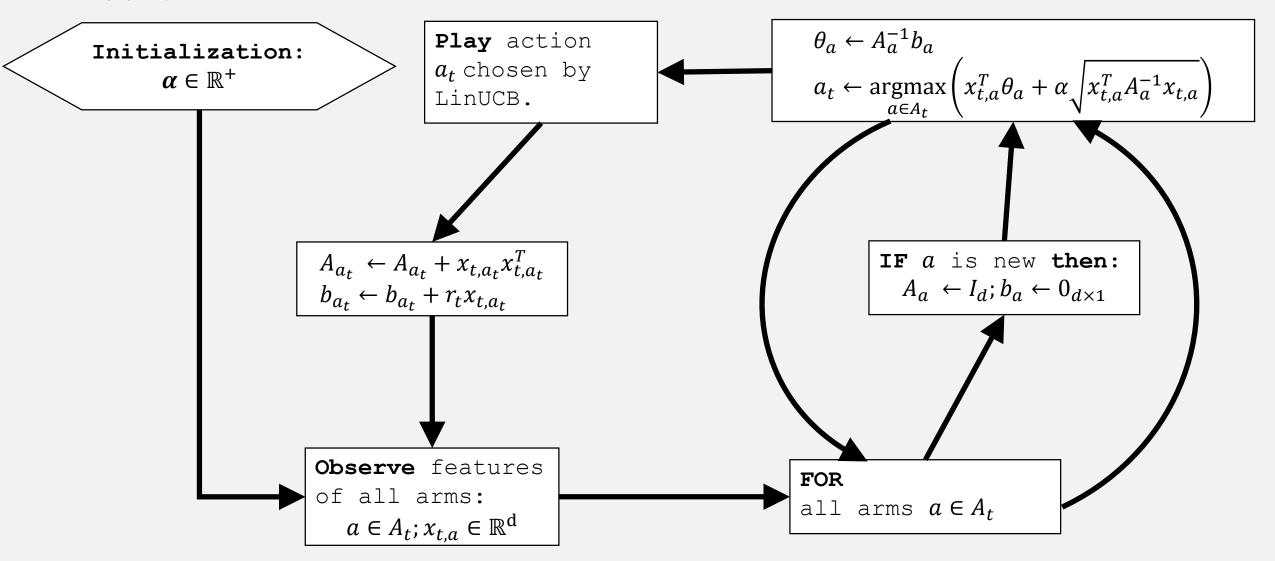
Which gives us following **UCB arm selection policy**:

$$a_t = \operatorname*{argmax}_{a \in A_t} \left(x_{t,a}^T \theta_a + \alpha \sqrt{x_{t,a}^T A_a^{-1} x_{t,a}} \right)$$
 where $A_a = D_a^T D_a + I_d$



Linear Contextual Bandits

LinUCB:





Regret Analysis

$$\mathbb{E}[R(T)] = O\left(d\sqrt{T\ln\frac{1+T}{\delta}}\right)$$



Disjoint vs. Hybrid Linear Models

Disjoint Linear Model:

Expected reward is linear in:

$$\mu(a|x_{t,a}) = x_{t,a}^T \theta_a^*$$

for all arms a and contexts x

Hybrid Linear Model:

Expected reward is linear in:

$$\mu(a|x_{t,a}) = \mathbf{z}_{t,a}^T \boldsymbol{\beta}^* + x_{t,a}^T \theta_a^*$$

for all arms a and contexts x

References

Slivkins, Aleksandrs. "Introduction to multi-armed bandits." Foundations and Trends® in Machine Learning 12.1-2 (2019): 1-286.

Li, Lihong, et al. "A contextual-bandit approach to personalized news article recommendation." *Proceedings of the 19th international conference on World wide web.* 2010.

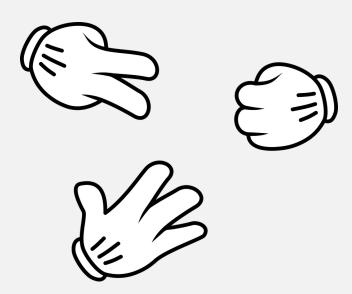
Zhou, Li. "A survey on contextual multi-armed bandits." arXiv preprint arXiv:1508.03326 (2015



Other Application for Bandit Algorithms

Bandit & Games







Bandit & Games

Matrix *M*:























ALG chooses row i_t of M

ADV chooses col j_t of M









```
For each round t \in [T]:
    Simultaneously:
         ALG chooses row i_t of M;
         ADV chooses col j_t of M;
    ALG incurs cost M(i_t, j_t)
    ALG observes feedback F_t = F(t, i_t, j_t, M)
```