

Bandits on Stochastic Blockmodel Graphs

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Reinforcement Learning

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Problem definition

N arms, T steps.

- ① The environment chooses losses for every arm noted $l_{t,i}$ for the arm i at the step t .
- ② Following the algorithm we hope would minimize as much as possible the regret the player draws an arm I_t .
- ③ The player receives the loss l_{t,I_t} .
- ④ We define $(O_t)_{i \in [N]}$ as the indicative function of observed loss at step t . We have:

$$O_{t,I_t} = 1 \quad \forall i \neq I_t, O_{t,i} \sim B(r)$$

$(O_t)_{i \in [N]}$ corresponds to the value of the logic expression *i is neighbor of I_t* in the Erdős-Rényi random graph drawn at step t .

- ⑤ For all i such that $O_{t,i} = 1$ the player can observe the loss $l_{t,i}$.

Duplex tricks

Definition geometrical random variables

$$M_t^* = \min\{1 \leq i < N : O'_{t-1,i} = 1\} \cup \{N\}$$

$$G_{t,i} = \min(K_{t,i}, M_t)$$

Independence

$$p_{t+2,i} \propto w_{t+2,i} = \frac{1}{N} \exp\left(-\eta_{t+2} \hat{L}_{t,i}\right)$$

Generalizing

ESTIMATE_R gives a safe lower bounding on r .

If $\underline{r} = 0$, we run vanilla EXP3.

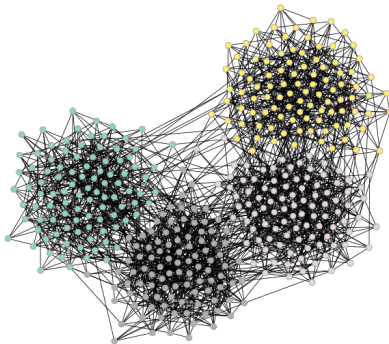
If $\underline{r} \geq \frac{\log T}{N}$, we run vanilla DUPLEX.

If $0 < \underline{r} < \frac{\log T}{N}$, $A = \left\lceil \frac{\log T}{N\underline{r}} \right\rceil$ and run DUPLEX grouping A steps.

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Random graph with Stochastic Block-Model



Neighbour communities

Probability of side information

- N clusters
- Labelled vertices
- $R = (r_{ij})_{1 \leq i, j \leq n}$ the matrix representing the probability of having side information.
- r_{ij} represents the probability that a vertex of the cluster j reveals his loss to a given vertex of the cluster i

Adapt M_t sampling

Unbiased estimator

$$\hat{l}_{t,i}^* = \frac{O_{t,i} l_{t,i}}{p_{t,i} + (1 - p_{t,i}) r_{C(l_t), C(i)}}$$

Using data...

$$\hat{l}_{t,i} = G_{t,i} O_{t,i} l_{t,i} \tag{1}$$

- Estimator $G_{i,t} \sim G(p_{t,i} + (1 - p_{t,i}) r_{l_t, i})$
 - M_t^i truncated geometric law of parameter $p_{t,i}$
- $G_{t,i} = \min(K_{t,i}, M_t^i)$

Sampling M_t^i

Algorithm adaptation

- Using previous observations
- Ensure independence of M_t and O_t
- Find the last $O_{t'}$, with $t' \equiv t - 1[2]$ in which $C(I(t')) = C(I(t))$
- Average the distance between two observation within cluster $C(j)$.

Keeping track of last time we picked I in cluster i

Lower bounding R

Generalized algorithm

If $\underline{R} = 0$,

we run vanilla EXP3.

If $\min_{i,j, \underline{r}_{i,j} > 0} (\underline{r}_{ij} N_j) \geq \log T$,

we run adapted DUPLEX.

Else,

$$A = \left\lceil \frac{\log T}{r_{N^*}} \right\rceil \text{ where}$$

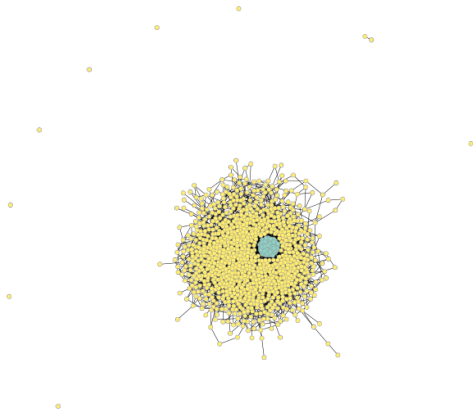
$$r_{N^*} = \min_{i,j, \underline{r}_{i,j} > 0} (\underline{r}_{ij} N_j)$$

and run DUPLEX grouping A steps.

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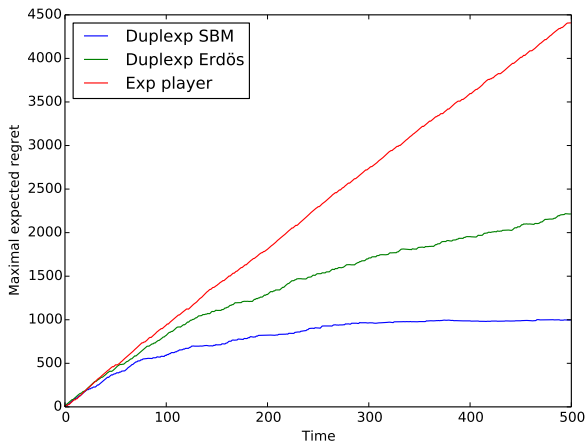
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Heavily unbalanced graph



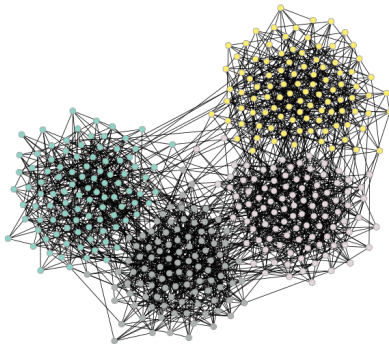
Neighbour communities

Graph 'close' to Erdős-Rényi



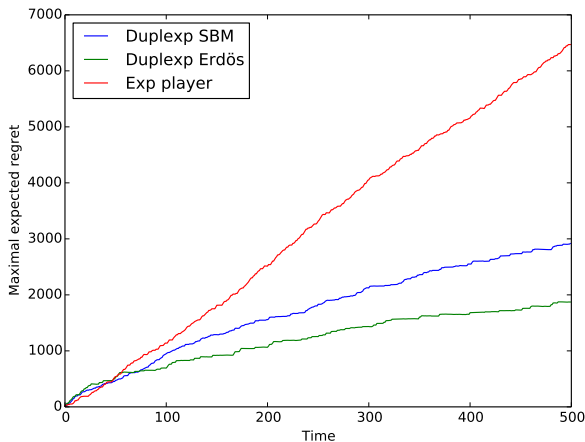
Original algorithm outperforms our algorithm

Graph 'close' to Erdős-Rényi



Neighbour communities

Graph 'close' to Erdős-Rényi



Original algorithm outperforms our algorithm