

BTP- Summary

A1

Algo:

for t in 1 to T :

- Select encoder from λ (ixen)
- Adversary selects channel
- get b from E (error matrix) (enxch)
- calculate $tprob[c]$ - prob. of selecting channel c from b
- Now update $temp[i,j]$ - matrix of temporary transition probabilities using eqn.
$$\Delta temp[i,j] = \log(sprob[i] tprob[j])$$

↓
steady state prob. from
target transition matrix (t)
- if $t \% step = 0$:
 - update P - target trans. prob. with $temp$
 - calculate $sprob$
 - Do binprog using this new $sprob$, E , state and update λ

Notes:

→ Two transition matrices are used $P(\text{target})$, $\text{temp}(\text{temporary})$ where target is updated for every few steps to reduce variance

A2

Algo:

for t in 1 to T :

→ Select encoder from $\lambda_{(1 \times n)}$

→ Adversary selects channel

→ get b from $E(\text{error matrix})_{(n \times ch)}$

→ if $t-1 \text{ step} = 0$:

→ model a multinomial HMM with
transition matrix - initialized to $P_{(\text{last estimate})}$
start prob. - $\text{steadyprob}(P)$

emission prob: $\lambda_{(1 \times ch)} * E_{(n \times ch)}_{(\text{last estimate})}$

feed the HMM with b and train
transition matrix

→ update P with ls and trans_mat

$$\Delta P = ls(\text{trans_mat} - P)$$

→ calculate s_{prob}

→ Do vinprog using this new
 s_{prob} , E , date and update
 λ

Notes:

- Do eps-greedy selection of encoders and decrease learning rate over time to get better estimates on p

A3:

Non Markov OPLB - Appron:

Algo: Same as the one in the paper with few sidebines:

- Instead of solving quadratic constrain used π_{t-1} in the square term
- used $\alpha_c = 0.001$ but given $\alpha_c \geq 1$ as $\alpha_c \geq 1$ is making equations infeasible for some initial values of $\pi (= \lambda a^T)$
- Also the constrain part can be loosened as our θ^* is known
- used pseudo inverse in some cases instead of inverse as it is blowing up for almost singular matrices which is often the case

Non Markov OPMD - Appron:

Algo: from the same paper
Sidebines:

- Consider each encoder as an arm
now $\pi (= \lambda a^T)$
- even here used $\alpha_c = 0.001$ otherwise

linprog. equations are becoming infeasible

NonMarkov OPLB - No approx - using cvxopt modelling

Algo: As it is mentioned in the paper with no approximation

Notes: \rightarrow Rate is converging to lower value than optimal / any of the above algorithms.

Markov OPLB - Approx

Algo:

for t from 1 to T :

\rightarrow Select encoder from $\lambda_{(1 \times n)}$

\rightarrow Adversary selects channel

\rightarrow get b from $E(\text{error matrix})_{(n \times ch)}$

\rightarrow if $t-1 \text{ step} = 0$:

\rightarrow model a multinomial HMM with transition matrix - initialized to $P_{(\text{last estimate})}$

start prob. - $\text{steadyprob}(P)$

emission prob: $\lambda_{(1 \times ch)} * E_{(\text{last estimate})}(n \times ch)$

feed the HMM with b and train transition matrix

\rightarrow update P with ls and trans_mat

$$\Delta P = ls(\text{trans_mat} - P)$$

\rightarrow Now calculate cost as

some heuristic of

$$y = P_{(ch \times ch)} * E'_{(ch \times en)} * x_{(en \times 1)}$$

* here $y[i]$ is the cost if previous channel is i

* using $\max(y)$ will get safe n

→ Now apply DP/B as usual using this cost formulation.