

Q-Learning Algorithm for VoLTE Closed-Loop Power Control in Indoor Small Cells

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MOTIVATION

Downlink VoLTE Power Control (PC)

- Network impairments impact SINR and data rates
- LTE and 5G NR do not have downlink power control
- Link adaptation is instead used in the downlink
- Network intelligence enables clearer communications
- Current practice:**
 - Diversity and more robust modulation and coding
- Proposed:**
 - Reinforcement learning is used to maximize SINR

Goal

- Enable downlink power control for indoor voice bearer
- Data is 60% likely to be on Wi-Fi!

Approach

- Q-learning improves the downlink SINR through:
 - Performing exploration and exploitation
 - Running until it finds a near-optimal policy

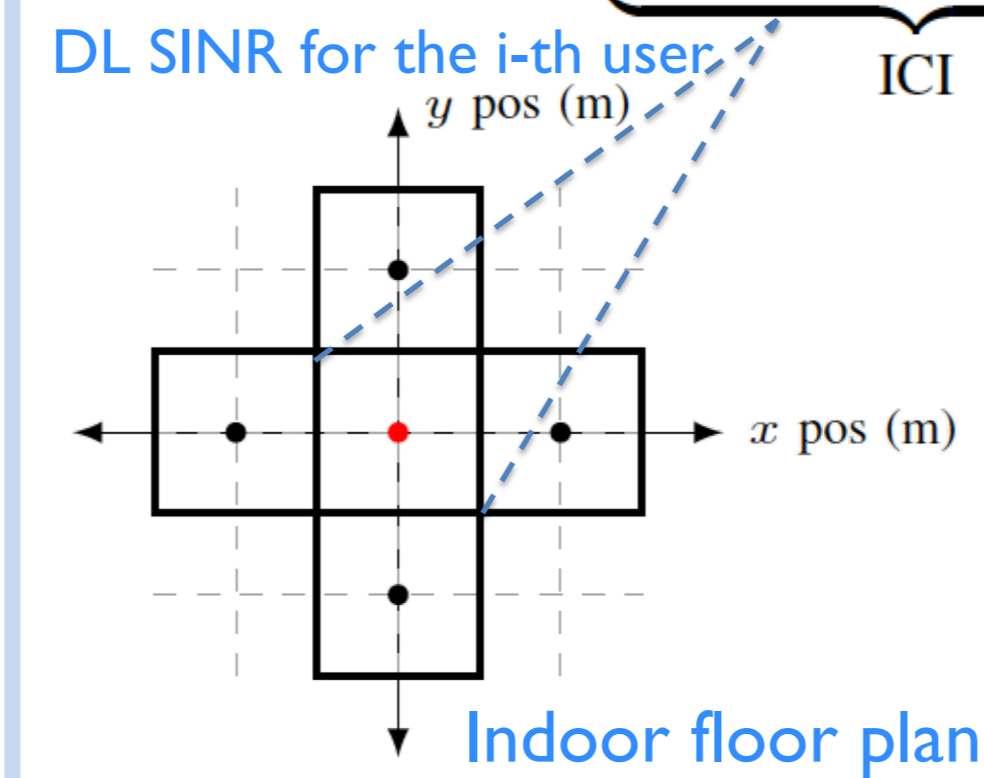
NETWORK MODEL

- An indoor OFDM-based cellular cluster in the FDD mode of operation
- Multi-user multiple access with one tier of neighboring cells
 - Square geometry and inter-site distance of length L
- Network impairments regularly occurring at random
- PC keeps track of forward link SINR and potential radio impairments

Forward link budget for the i -th user

$$P_{UE,i}[t] = P_{TX}[t] + G_{TX} - L_m - L_{a,i}[t] + G_{UE}$$

$$\gamma_{DL,i} \triangleq \frac{|y_i|^2}{\sigma_n^2 + \underbrace{\sum_{j: o_j \in \mathcal{C} \setminus \{o_0\}} k_j |y_j|^2}_{ICI}}$$



$$P_{TX}[t] = \min(P_{BS}^{max}, P_{TX}[t-1] + \eta[t]c[t])$$

Power control repetition (up to 3 times per TTI)

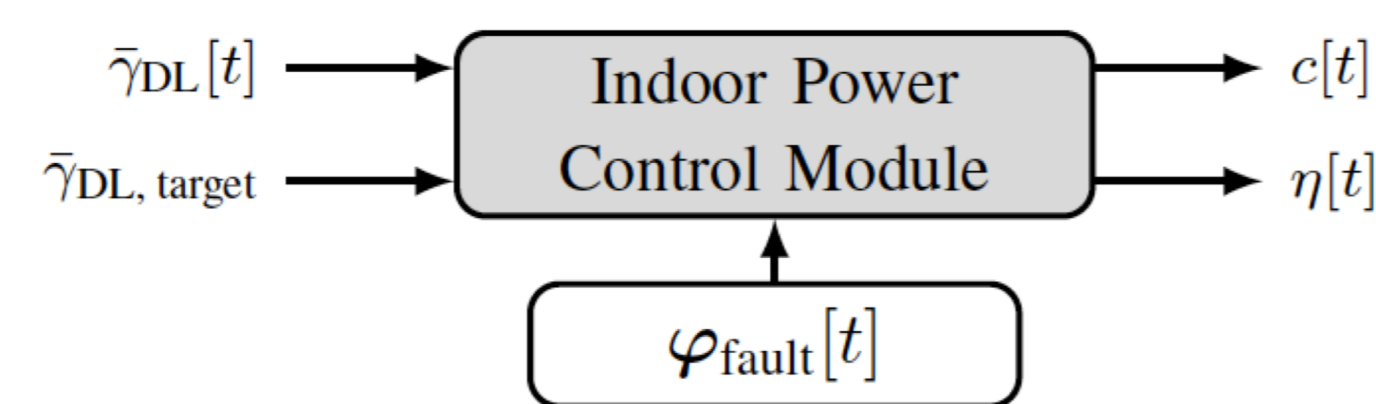
POWER CONTROL (PC) ALGORITHM ACTIONS AT TIME t

| Action a | Definition |
|------------|---|
| 0 | Nothing (this is a transient action). |
| 1 | Three (PC = -1) executed (i.e., $\eta[t] = 3$). |
| 2 | Single (PC = -1) executed (i.e., $\eta[t] = 1$). |
| 3 | Single (PC = +1) executed. |
| 4 | Three (PC = +1) executed. |

PARAMETERS

Radio Network

| Radio Network Parameters | | | |
|--------------------------|----------|-----------------|-----------------|
| Cell radius | 10 m | Geometry | Square |
| f_c | 2600 MHz | Antenna pattern | Omnidirectional |
| Bandwidth | 20 MHz | Prop. model | COST231 |
| Tx Power | 33 dBm | Tx Ant. Height | 10 m |
| Tx, Rx branches | 2,2 | Rx Ant. Height | 1.5 m |

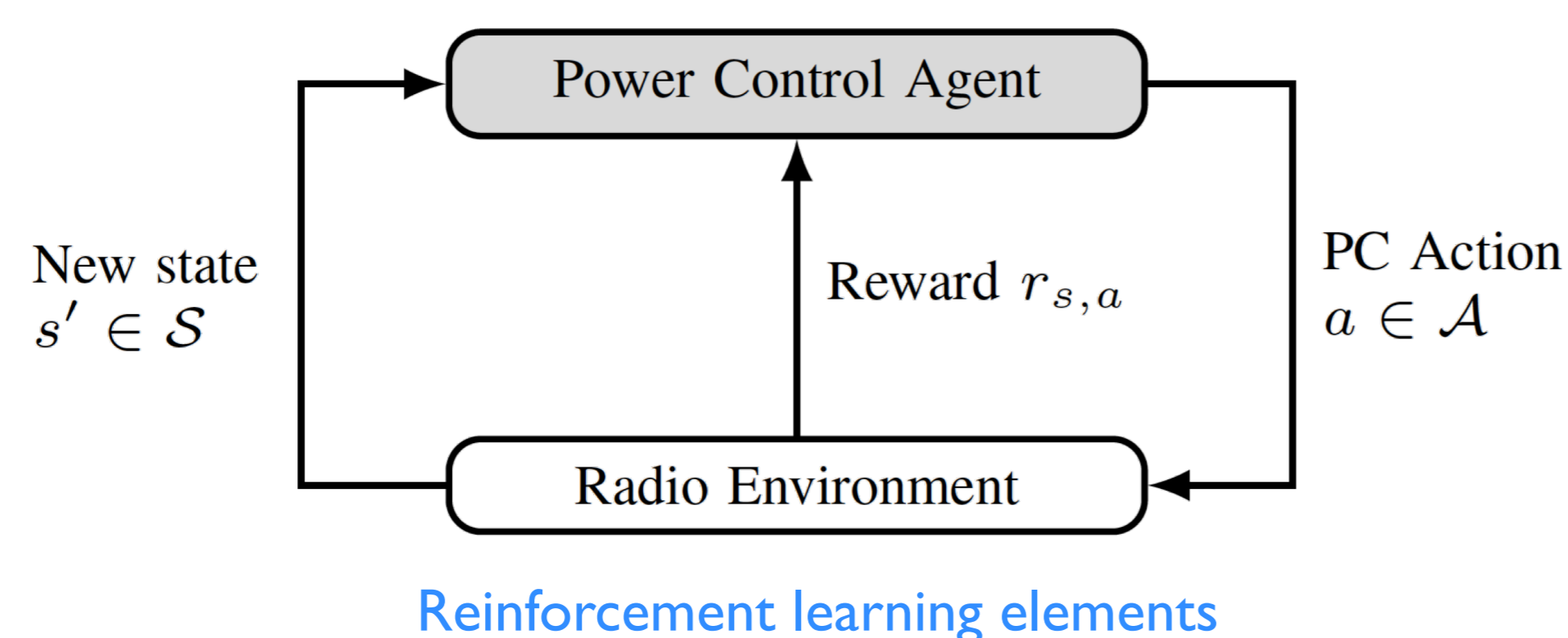


$$Q(s, a) \triangleq (1 - \alpha)Q(s, a) + \alpha [r_{s,a} + \gamma \max_{a'} Q(s', a')]$$

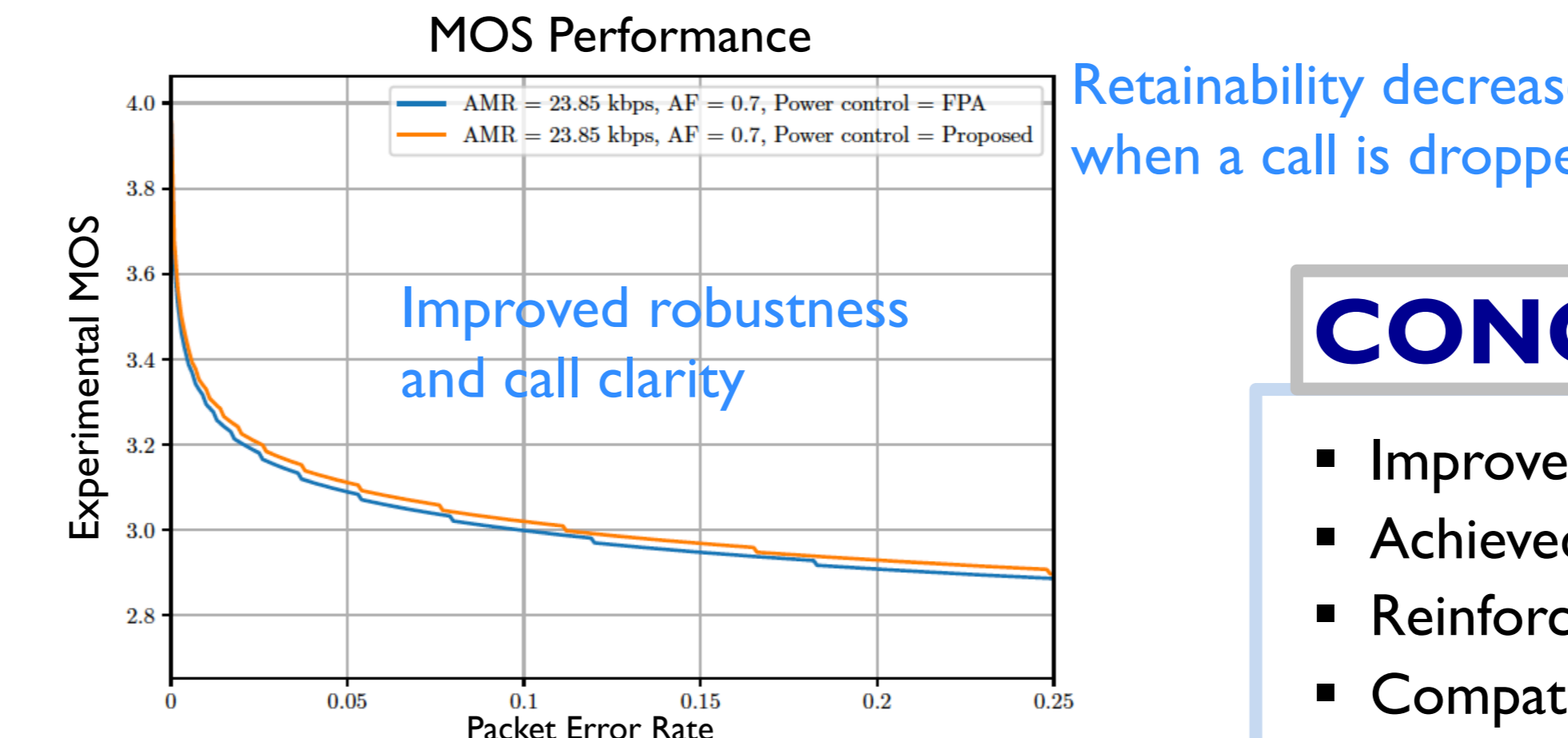
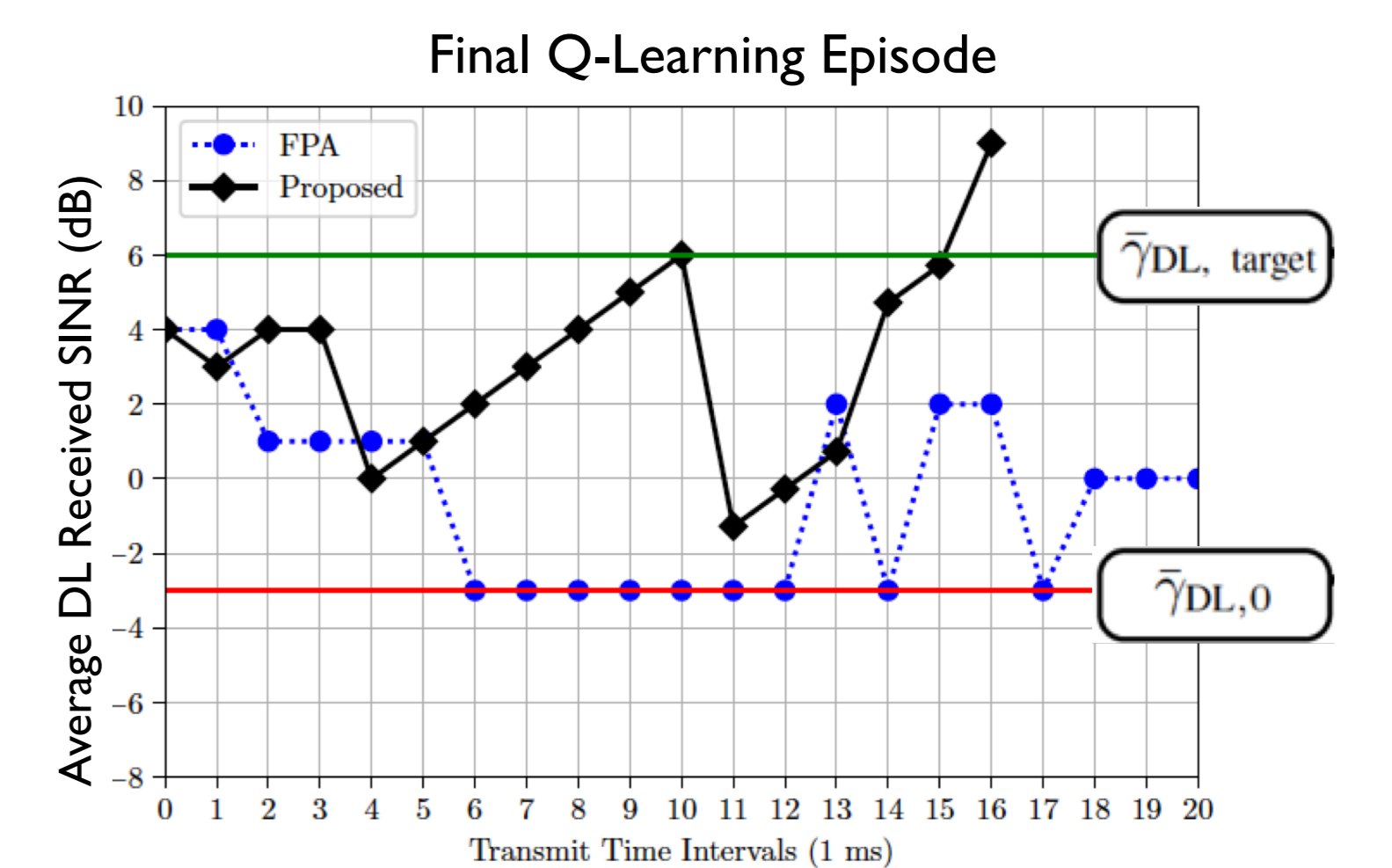
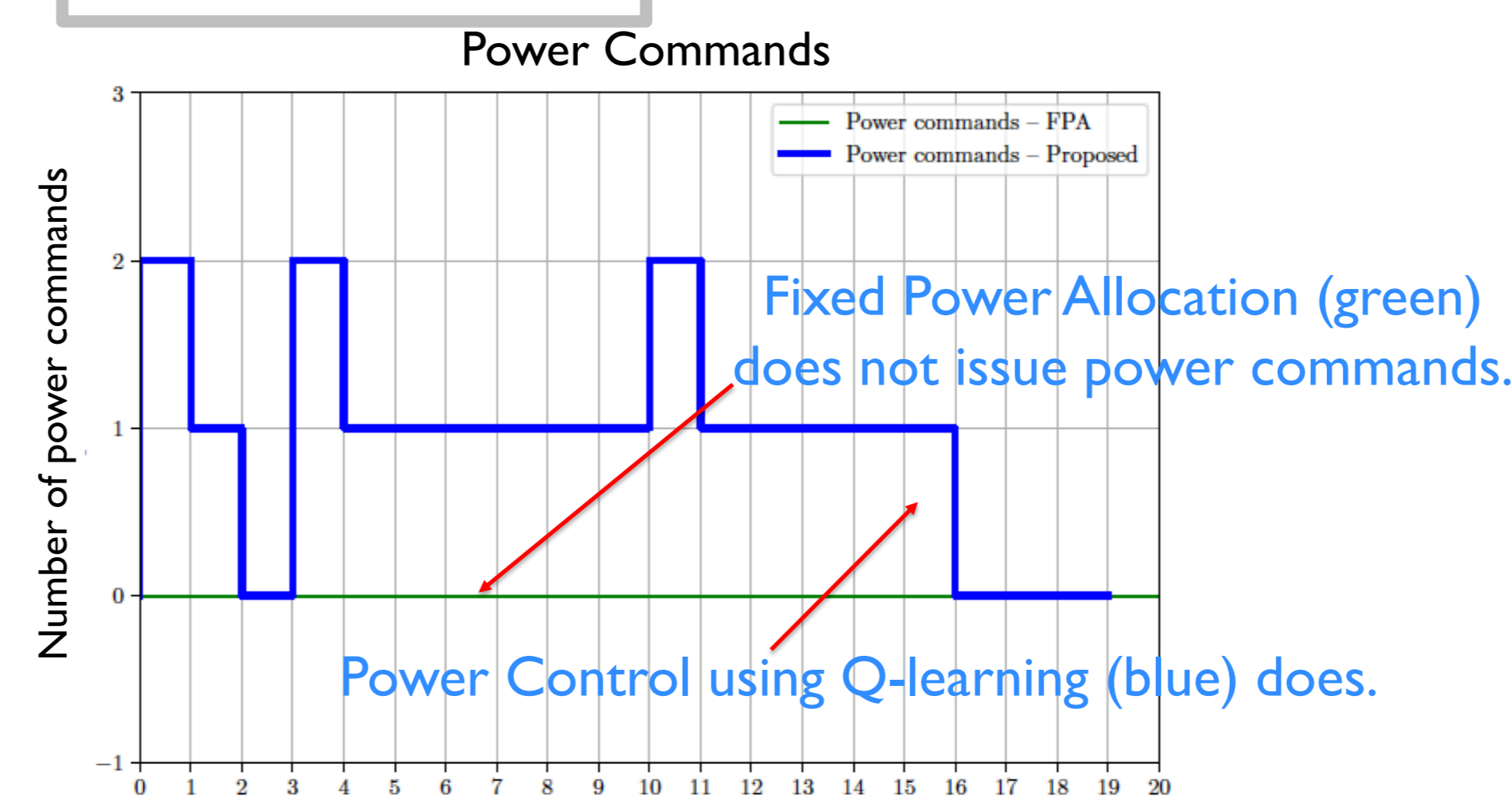
The discounted future-looking reward

Q-Learning Parameters

| MACHINE LEARNING PARAMETERS | |
|---|-------|
| Parameter | Value |
| Number of episodes ζ | 707 |
| One episode duration τ (ms) | 20 |
| Discount factor γ | 0.950 |
| Exploration rate ϵ | 1.000 |
| Minimum exploration rate ϵ_{min} | 0.010 |
| Exploration rate decay d | 0.99 |
| Learning rate α | 0.001 |
| Number of states | 3 |
| Number of actions | 5 |



RESULTS



| | Fixed Power Allocation | Proposed |
|---------------|------------------------|----------|
| Retainability | 55.00% | 78.75% |

CONCLUSIONS

- Improved downlink SINR
- Achieved higher voice call retainability and clarity
- Reinforcement learning enables power control
- Compatible with LTE and 5G NR standards

