



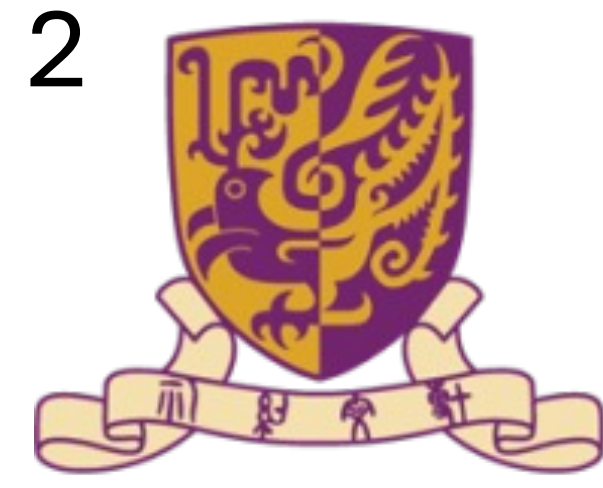
Stochastic Bandits Robust to Adversarial Attacks



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Multi-armed bandits & Adversarial attacks

- K arms, each with a stochastic reward X_k with **unknown** mean μ_k
- $\Delta_k := \mu_{k^*} - \mu_k$ where $k^* := \arg \max_k \mu_k$
- T sequential decision rounds.
- **Regret**: $R_T := T\mu_{k^*} - \sum_{t=1}^T \mu_{I_t}$
- **Attack budget**: $\mathcal{C} := \sum_{t=1}^T |X_{k,t} - \tilde{X}_{k,t}|$

Attack (above) vs. Corruption (below)

Robust algorithm designs & Regret bounds

Known Attack Budget

$$O\left(\sum_{k \neq k^*} \frac{\log T}{\Delta_k} + KC\right) \xrightarrow{\text{Stop Cond.}} O(\sqrt{KT}(\log T + C))$$

Multi-Phase

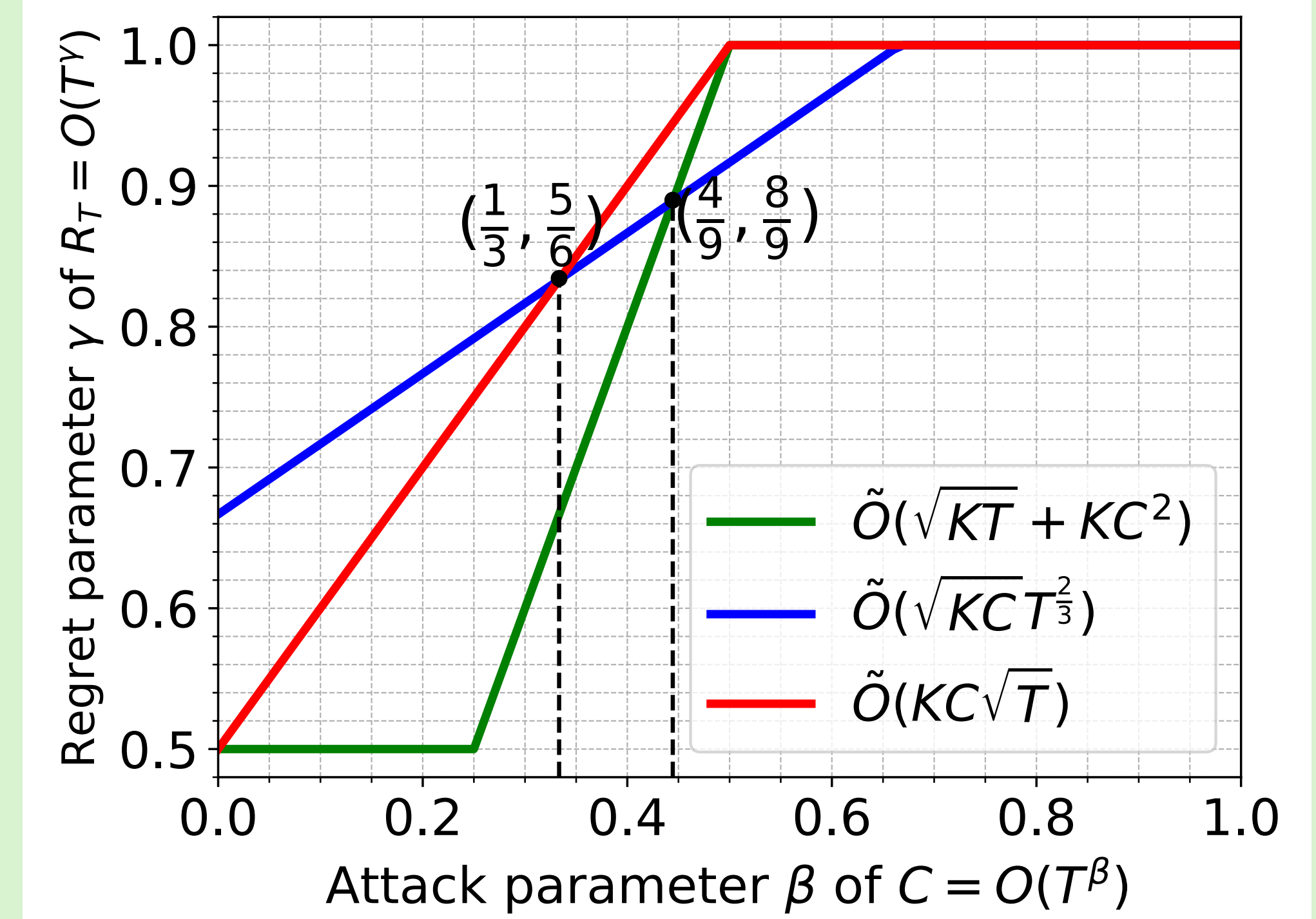
Model Selection

$$\text{PE-WR} \\ \tilde{O}(\sqrt{KT} + KC^2)$$

$$\text{MS-SE-WR} \\ \tilde{O}(\sqrt{KCT}^{\frac{2}{3}}), \tilde{O}(KC\sqrt{T})$$

Unknown Attack Budget

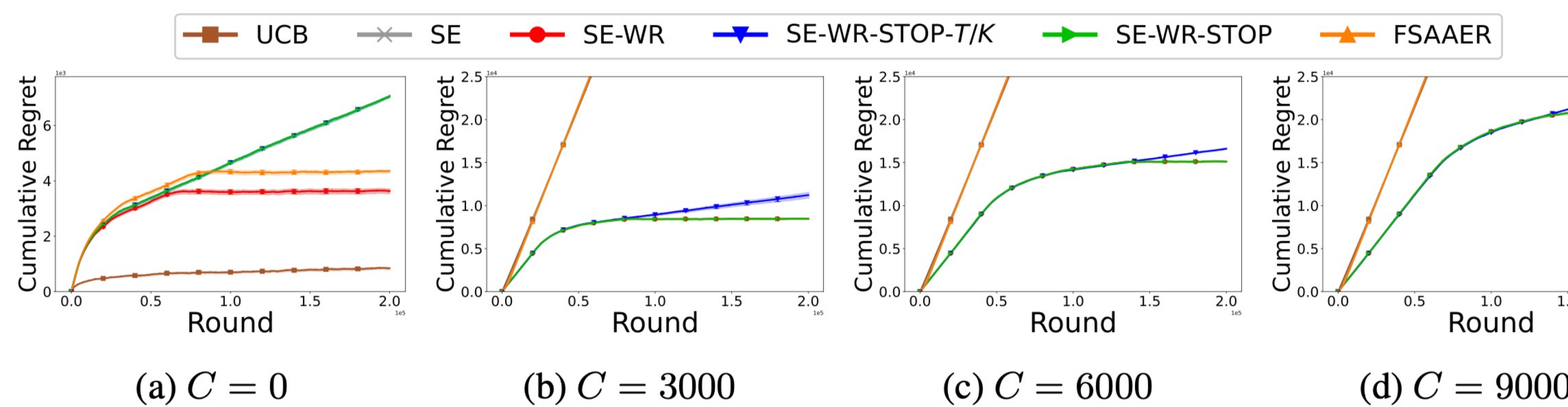
Additive vs. Multiplicative



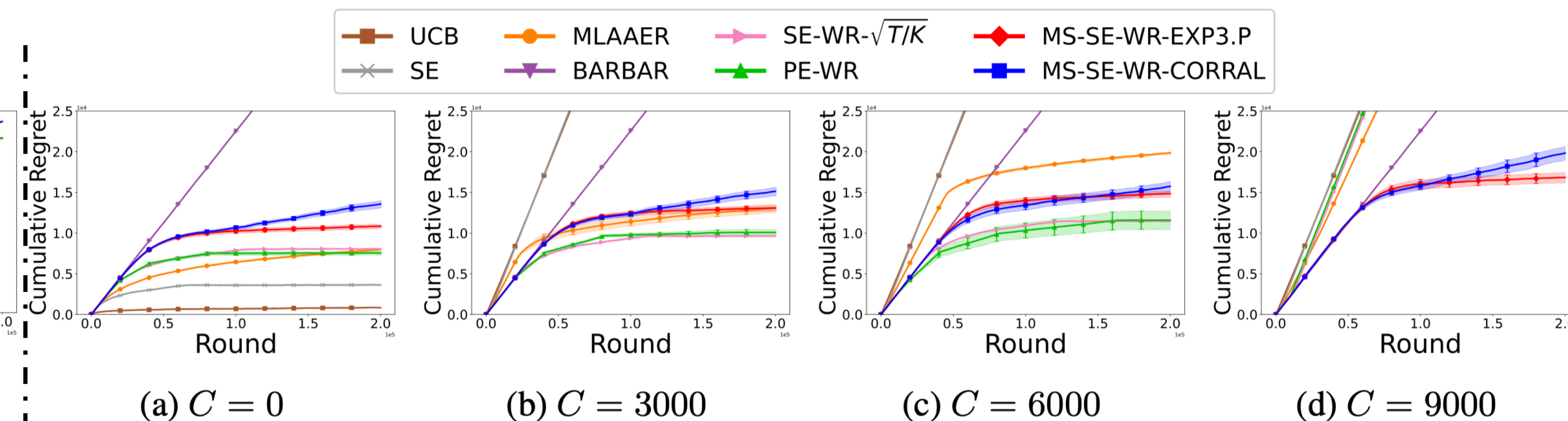
*All upper bounds are tight to some logarithmic factors (see lower bounds in the paper).

- Successive elimination with wide confidence radius (SE-WR) improves the $\sum_k C/\Delta_k$ of Lykouris et al. (2018, Theorem 1) to KC .
- SE-WR-STOP introduces a stopping condition to transfer the gap-dependent to independent (worse-case) bound.
- Phase Elimination (PE-WR) uses multiple phases, each dealing with one possible attack budget.
- Model Selection (MS-SE-WR) uses $\log_2 T$ base algorithms (SE-WR-STOP) with different attack budgets and pick instances as “arms” via either CORRAL or EXP3.P to defense adaptively.

Simulations: Regret comparison



Algorithms with **known** attack budget C



Algorithms with **unknown** attack budget C