

No-Regret is not enough! Bandits with General Constraints through Adaptive Regret Minimization

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0. Abstract

- Problem description:
 - In the standard bandits with knapsacks (BwK) problem, the learner has m resource-consumption constraints.
 - In this paper, a general BwK problem is considered where the learner has a set of general long-term constraints.
 - **Goal** - maximize cumulative reward while limiting cumulative constraint violations.
- **Problem** - conventional methods fail to yield sublinear violations of constraints.
- **Solution** - circumvent the issue by requiring the primal and dual algorithm to be *weakly adaptive*.
 - **Key Idea** - Without information on Slater's parameter ρ , the interplay between weakly adaptive primal and dual regret minimizers yields a "self-bounding" property of two variables.
- **Results**
 - Stochastic constraints - show that the algorithm guarantees sublinear regret.
 - Adversarial constraints - establish a tight competitive ratio of $\rho/(1 + \rho)$.
 - Both cases - constraint violations are guaranteed to be sublinear in time.
 - Contextual bandits with linear constraints - provides the first no- α -regret guarantees for adversarial contexts.