

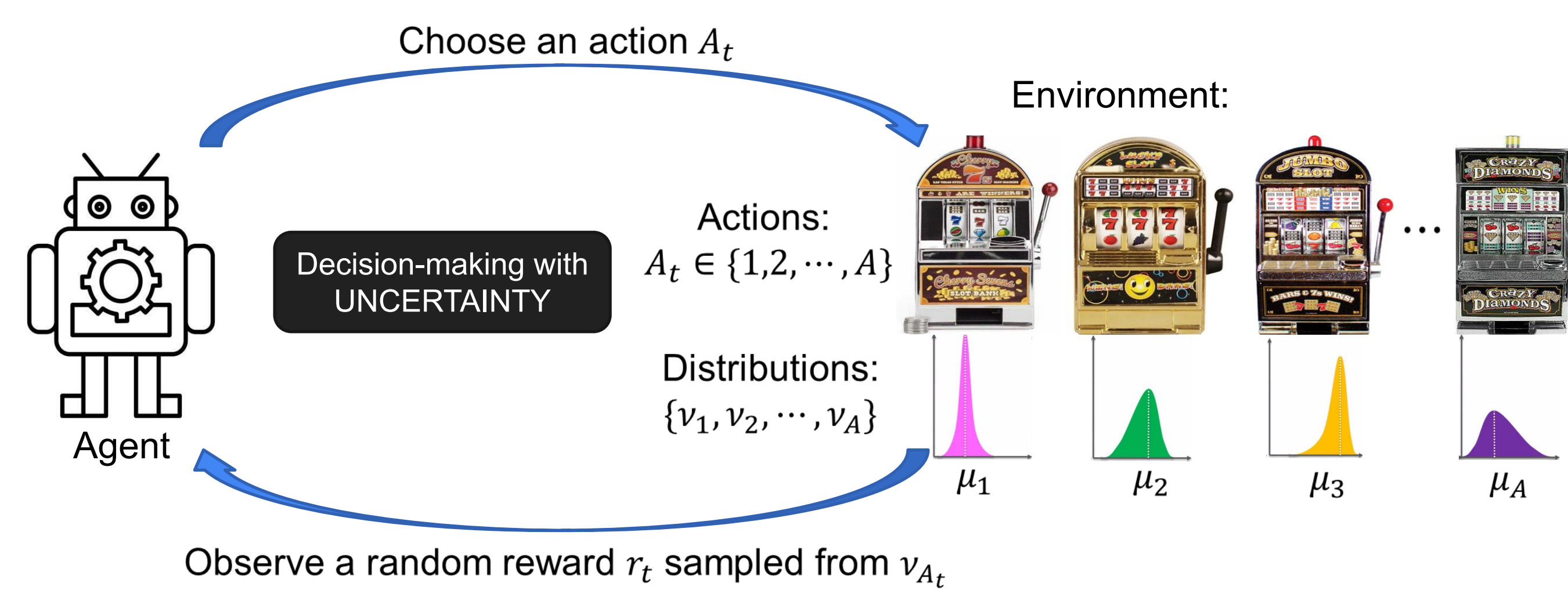
Cost Aware Best Arm Identification

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<https://arxiv.org/abs/2402.16710>

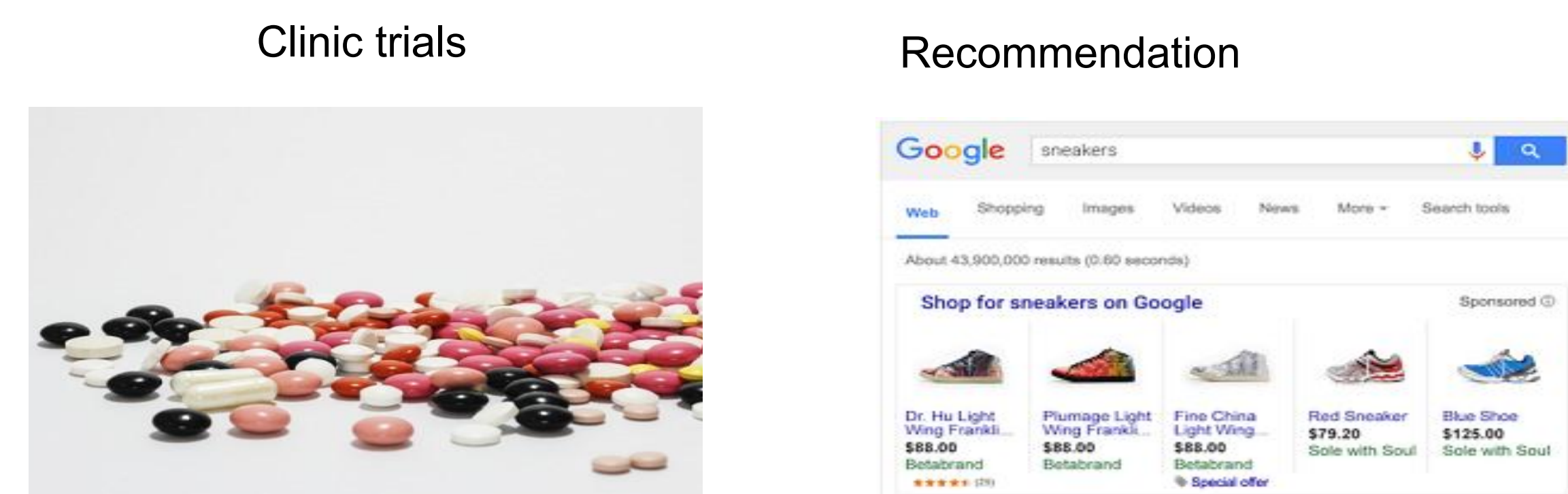
Multi Armed Bandits

- Online sequential decision making.
- Many real world applications.



Learning / Implementation Phase

- Real world problems often have different objectives between learning and implementation



- Learning: Minimize cost of trials
- Implementation: Cost is subsidized want most effective.
- Learning: Minimize cost to display ads
- Implementation: Click through rate.

Cost Aware BAI

- Best Arm Identification (Garivier, et al. 2016) ← Each arm has same cost

BAI

$$\min \tau$$

$$\text{s.t } \Pr(\hat{a}_\tau = a^*) \geq 1 - \delta$$

CBAI

$$\min \sum_a \mathbb{E}[c_a N_a(\tau)]$$

$$\text{s.t } \Pr(\hat{a}_\tau = a^*) \geq 1 - \delta$$

Cost Optimal Best Arm Identification

- Asymptotic optimality in incurred cost.
- Asymptotic lower bound on the cost for fixed confidence.

$$\mathbb{E} \left[\sum_a c_a N_a(\tau_\delta) \right] \geq T^*(\mu) \log(1/\delta) + o(\log(1/\delta))$$

$$T^*(\mu)^{-1} = \sup_{w \in \Sigma_K} \inf_{\lambda \in \{a^*(\lambda) \neq a^*(\mu)\}} \sum_a \frac{w_a}{c_a} d(\mu_a, \lambda_a).$$

Our Algorithm: CTAS

- Cost Aware Track and Stop
- Each arm makes up an optimal proportion of the final cost.
- Idea: track optimal proportion and stop as soon as possible.

| Algorithm | Optimal? | $w_1(t)$ (1.5, 1) | $w_2(t)$ (1, 0.1) | $w_3(t)$ (0.5, 0.01) |
|-----------|----------|-------------------|-------------------|----------------------|
| TAS | × | 0.46 | 0.46 | 0.08 |
| CTAS | ✓ | 0.23 | 0.72 | 0.05 |

- Optimal proportions for heterogeneous cost.
(mean reward, mean cost) for each arm.

- Stopping is same as BAI: **ASAP**
- Use Generalized Likelihood Ratio.

Theoretical Results

Asymptotically optimal in expectation:

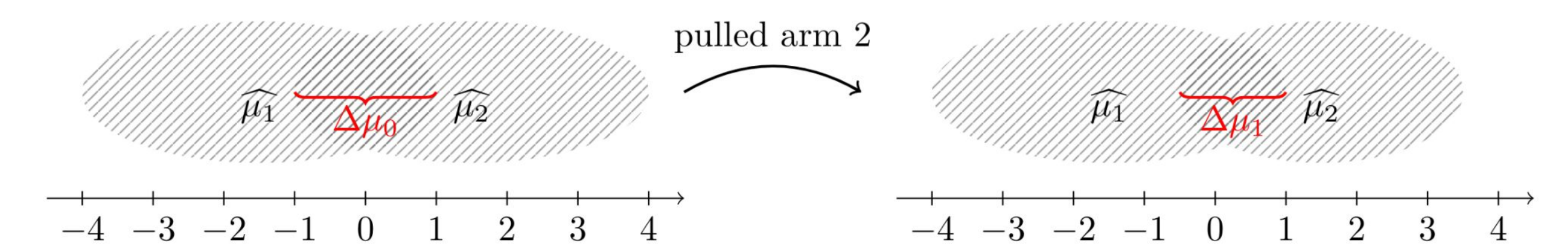
- Under CTAS (for suitable choices)

$$\limsup_{\delta \rightarrow 0} \frac{\mathbb{E}[\sum_a c_a N_a(\tau_\delta)]}{\log(1/\delta)} \leq T^*(\mu)$$

Low Complexity Algorithm

Idea: pull arm $a_t \in \arg \min_{a \in \mathcal{R}} \sqrt{c_a} N_a(t)$

- Asymptotically optimal in two armed Gaussian bandits.



| | CO | CTAS | TAS | d-LUCB |
|-----------|----|------|------|--------|
| Gaussian | 85 | 1712 | 2410 | 82 |
| Bernoulli | 58 | 1995 | 2780 | 60 |
| Poisson | 96 | 3260 | 4633 | 101 |

Time (sec) of each algorithm over 1000 trajectories.

Numerical Results

