

Resolution Limits of 20 Questions Search Strategies for Moving Targets



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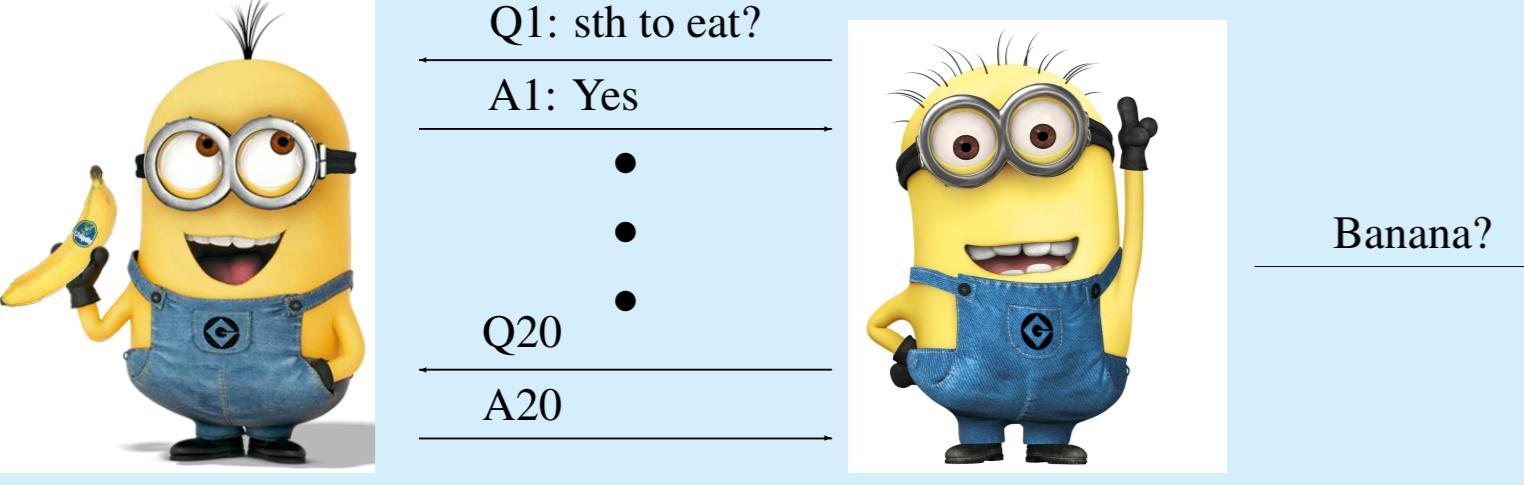
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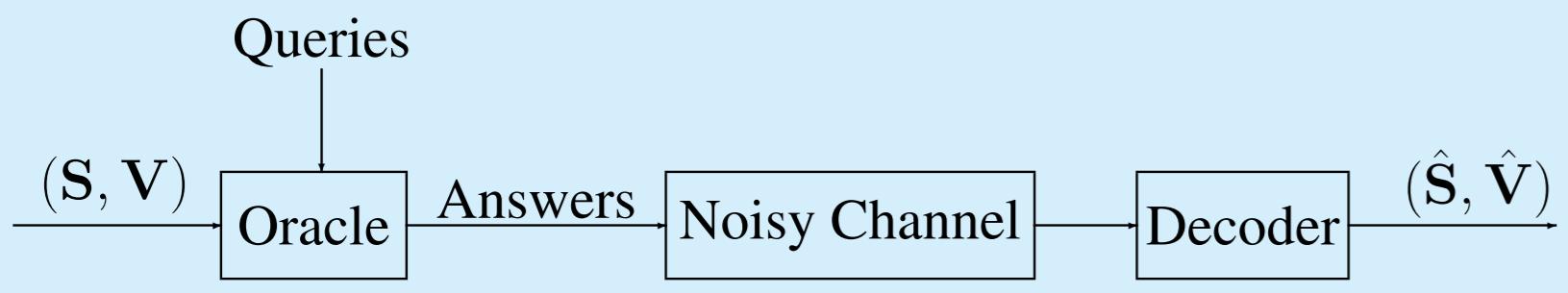
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Origin: 20 Questions Game



- Two players: oracle and questioner
- Querying and answering game with a budget (cf. [1]-[3])
- Noiseless: oracle is always honest
- Key: design smart queries to extract the secret as soon as possible

Problem Formulation

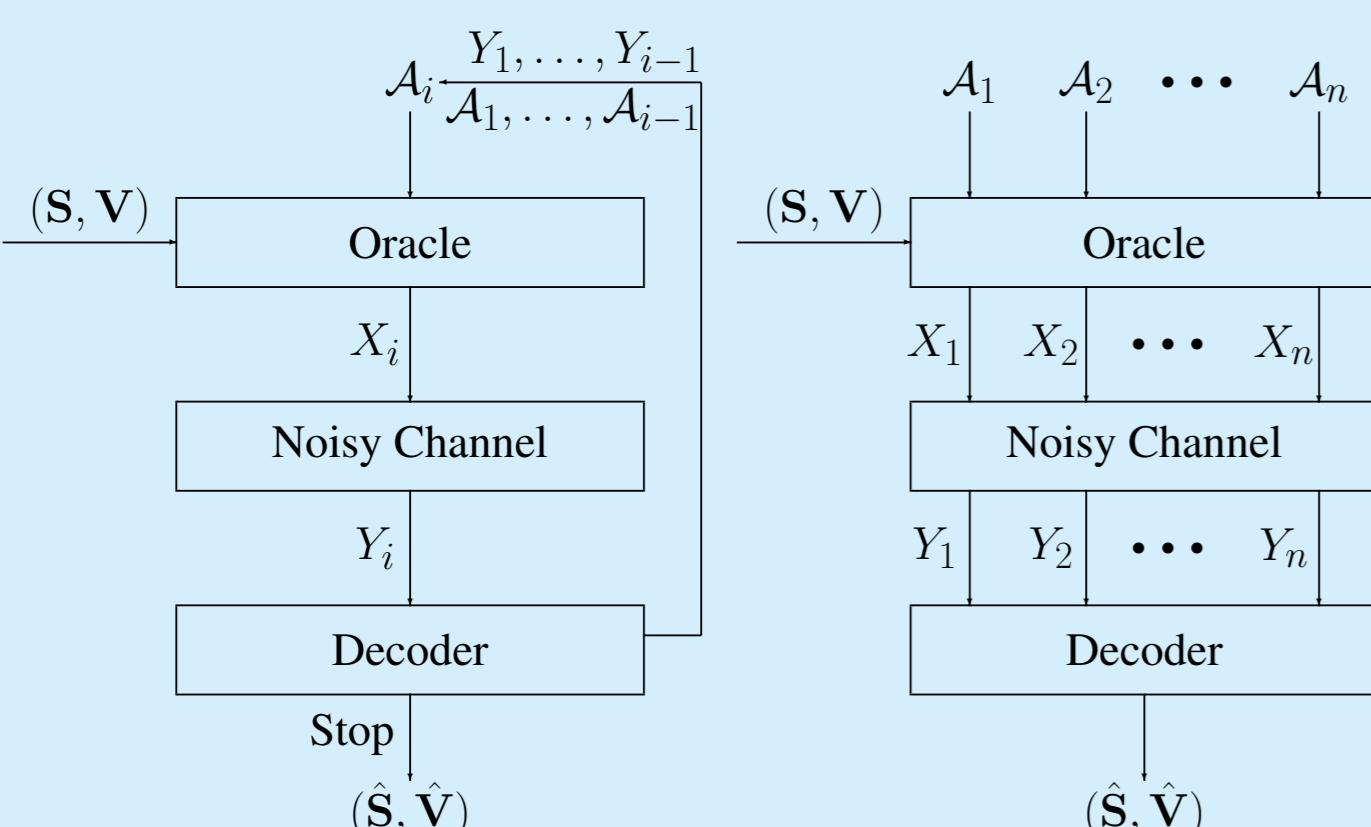


- Target: accurately estimate the trajectory of a moving target with unknown initial location $S = (S_1, \dots, S_d) \in \mathcal{S}^d$ and moving velocity $V = (V_1, \dots, V_d) \in \mathcal{V}^d$
- $\mathcal{S} = [0, 1]$, $\mathcal{V} = [-v_+, v_+]$ and $(S, V) \sim f_{SV}$.
- Task: design queries and decoder (scheme/strategy/policy)
- Motivation: applications such as localizing a moving target using a sensor network

The Torus Model

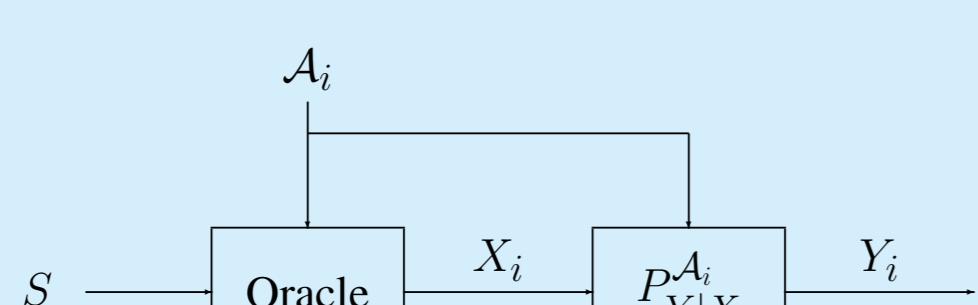
- Consider a target with initial location $s = (s_1, \dots, s_d) \in [0, 1]^d$ and velocity $v = (v_1, \dots, v_d) \in \mathcal{V}^d$
 - Torus model: at each time point $t \in [n]$ and in each dimension $i \in [d]$, the location of the target is
- $$l(s_i, v_i, t) := \begin{cases} 1 & \text{if } \text{mod}(s_i + tv_i, 1) = 0, \\ s_i + tv_i - \lfloor s_i + tv_i \rfloor & \text{if } s_i + tv_i \in \bigcup_{h \in \mathbb{N}} [2h, 2h+1], \\ \lceil s_i + tv_i \rceil - (s_i + tv_i) & \text{otherwise,} \end{cases}$$
- the target moves within a unit length interval
 - similar to the unit circle model studied by Kaspi, Shayevitz and Javadi (T-IT 2018)
 - The location of the target, denoted as $l(s, v, t)$, is a d -dimensional vector $(l(s_1, v_1, t), \dots, l(s_d, v_d, t))$. At each dimension,

Adaptive and Non-adaptive Querying



- A query A_i asks whether the target lies in a certain region $A_i \subseteq [0, 1]^d$ at a discrete time $i \in [n]$ and X_i is the noiseless answer
- **Non-adaptive querying**
 - queries are designed simultaneously offline
 - lower cost, faster execution time due to parallelizability

Measurement Dependent Noisy Channel



- Binary Input alphabet \mathcal{X} and arbitrary output alphabet \mathcal{Y}
- For any $\mathcal{A} \subseteq [0, 1]^d$, $P_{Y|X}^{\mathcal{A}}$ depends on \mathcal{A} through its size via a bounded Lipschitz continuous function $f : [0, 1] \rightarrow \mathbb{R}_+$
- Example: for any parameter $\zeta \in (0, 1)$ and any $\mathcal{A} \subseteq [0, 1]^d$
 - Measurement dependent binary symmetric channel (BSC)

$P_{Y X}^{\mathcal{A}}$	$X = 0$	$X = 1$
$Y = 0$	$1 - \zeta f(\mathcal{A})$	$\zeta f(\mathcal{A})$
$Y = 1$	$\zeta f(\mathcal{A})$	$1 - \zeta f(\mathcal{A})$

Definition of Query Procedure

Given any $n \in \mathbb{N}$, $\delta \in \mathbb{R}_+$ and $\varepsilon \in [0, 1]$, to search for a moving target over a d -dimensional torus, an $(n, d, \delta, \varepsilon)$ -non-adaptive query procedure consists of

- n queries \mathcal{A}^n , where for each $i \in [n]$, we query whether the target locates inside a Lebesgue measurable subset $\mathcal{A}_i \in [0, 1]^d$
- and a decoder $g : \mathcal{Y}^n \rightarrow [0, 1]^d \times \mathcal{V}^d$

such that the excess-resolution probability satisfies

$$P_e(n, d, \delta) := \sup_{f_{SV}} \Pr \left\{ \max_{t \in [0:n]} \|l(\hat{S}, \hat{V}, t) - l(S, V, t)\|_\infty > \delta \right\} \leq \varepsilon,$$

Accurate estimation of the trajectory implies accurate estimate of the initial location and velocity, and vice versa

- $|\hat{S}_i - S_i| < \frac{\delta}{2}$ and $n|\hat{V}_i - V_i| < \frac{\delta}{2}$ implies accurate estimation of the trajectory, i.e., $\max_{t \in [0:n]} \|l(\hat{S}, \hat{V}, t) - l(S, V, t)\|_\infty \leq \delta$
- $|\hat{S}_i - S_i| > \delta$ or $n|\hat{V}_i - V_i| > 2\delta$ implies poor estimate

Fundamental Limit

- Given any number of queries $n \in \mathbb{N}$ and $\varepsilon \in [0, 1]$,

$$\delta^*(n, d, \varepsilon) := \inf \{ \delta \in \mathbb{R}_+ : \exists \text{ an } (n, d, \delta, \varepsilon)\text{-non-adaptive query procedure} \}.$$

- minimal non-asymptotic resolution achievable by any non-adaptive query procedure with n queries and excess-resolution probability ε

- Dual quantity (sample complexity):

$$n^*(d, \delta, \varepsilon) = \inf \{ n \in \mathbb{N} : \delta^*(n, d, \varepsilon) \leq \delta \}$$

Preliminaries

- $P_X = \text{Bern}(p)$ denotes the Bernoulli distribution
- $P_{Y|X}^q$ denotes $P_{Y|X}^{\mathcal{A}}$ when $f(|\mathcal{A}|) = q$
- $P_Y^{p,q}$ denotes the distribution on \mathcal{Y} induced by P_X and $P_{Y|X}^q$
- For any $(x, y) \in \mathcal{X} \times \mathcal{Y}$, define the mutual information density

$$i_{p,q}(x; y) := \log P_{Y|X}^q(y|x) - \log P_Y^{p,q}(y).$$

- “Capacity” of measurement dependent channels $\{P_{Y|X}^q\}_{q \in [0,1]}$

$$C := \max_{p \in [0,1]} \mathbb{E}[i_{p,f(p)}(X; Y)], (X, Y) \sim \text{Bern}(p) \times P_{Y|X}^{f(p)}$$

- “Dispersion” of measurement dependent channels

$$V_\varepsilon := \begin{cases} \min_{p \in \mathcal{P}_{ca}} \text{Var}[i_{p,f(p)}(X; Y)] & \text{if } \varepsilon < 0.5, \\ \max_{p \in \mathcal{P}_{ca}} \text{Var}[i_{p,f(p)}(X; Y)] & \text{if } \varepsilon > 0.5. \end{cases}$$

Main Result

Theorem 1 For any $\varepsilon \in (0, 1)$ and finite $d \in \mathbb{N}$, the minimal achievable resolution $\delta^*(n, d, \varepsilon)$ satisfies the following properties

- if $nv_+ = O(n^t)$ for $t \in [0, 0.5]$,

$$-2d \log \delta^*(n, d, \varepsilon) = nC + O(nv_+);$$

- if $nv_+ = O(n^t)$ for $t \in [0, 0.5]$

$$-2d \log \delta^*(n, d, \varepsilon) = nC + \sqrt{nV_\varepsilon} \Phi^{-1}(\varepsilon) + O(\max\{nv_+, \log n\});$$

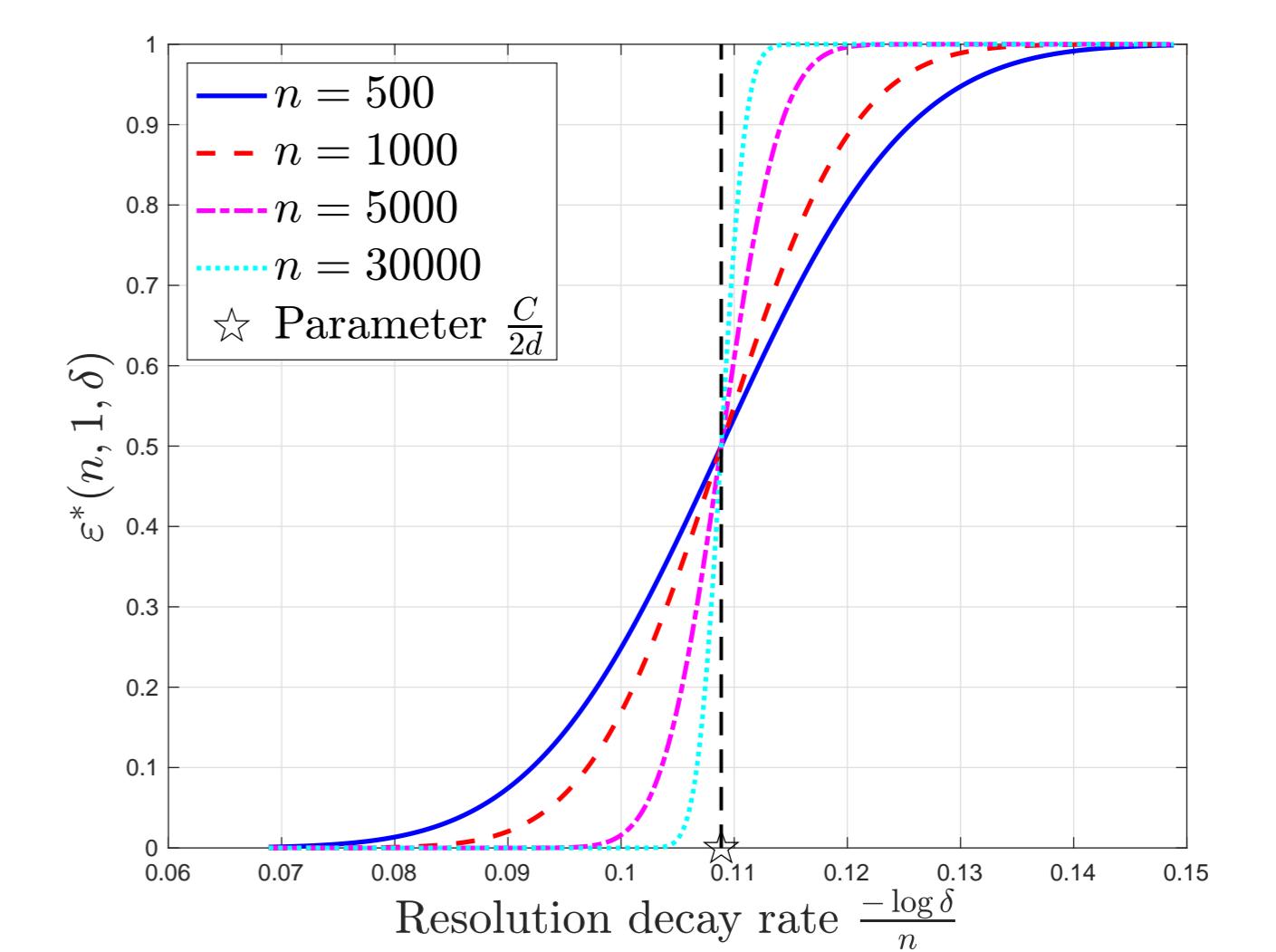
Phase Transition

- Minimal excess-resolution probability $\varepsilon^*(n, d, \delta)$

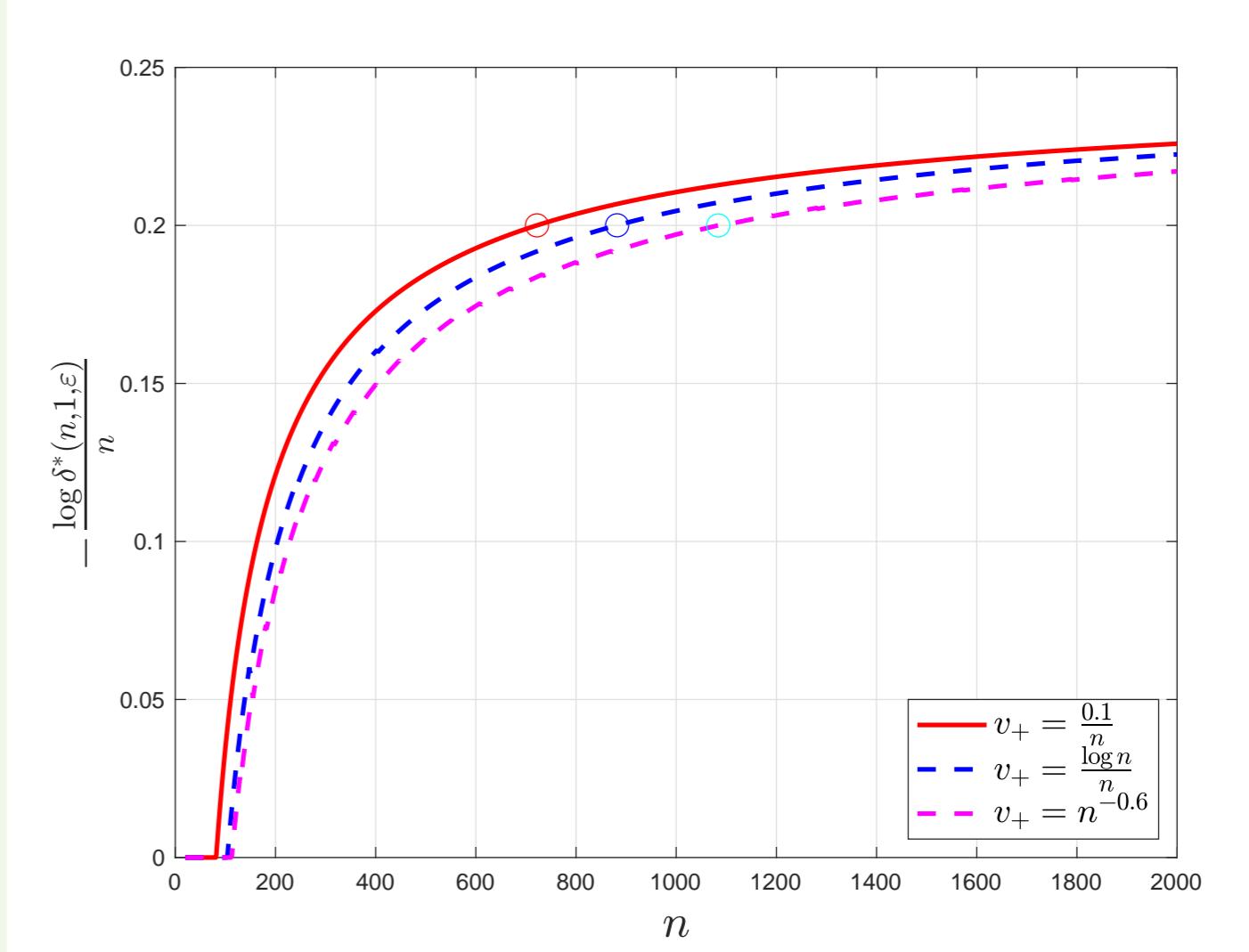
- For maximal speed such that $nv_+ = o(\sqrt{n})$

$$\varepsilon^*(n, d, \delta) = \Phi \left(\frac{-d \log \delta - nC}{\sqrt{nV_\varepsilon}} \right) + o(1)$$

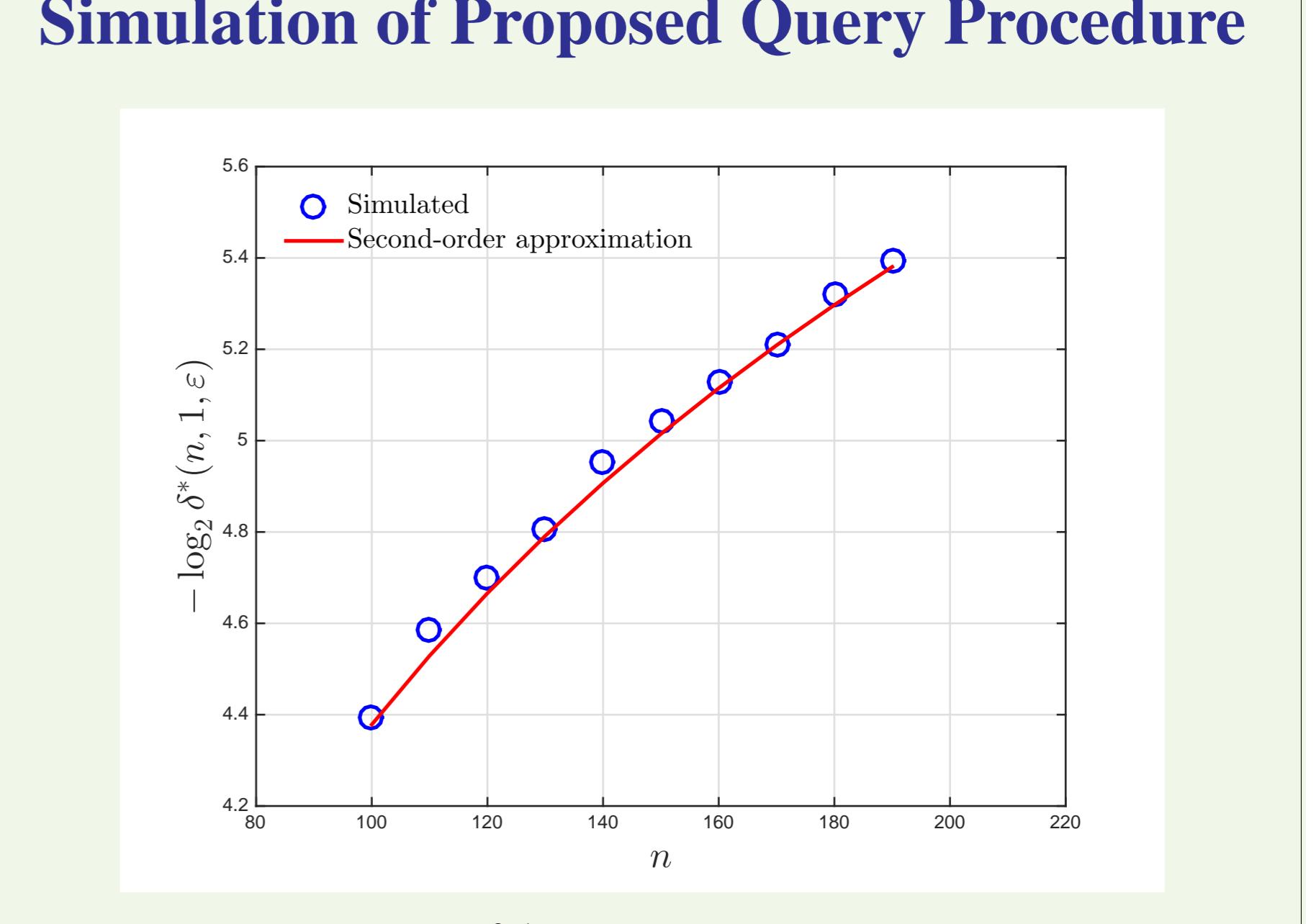
- Numerical illustration for a measurement dependent BSC with parameter $\zeta = 0.2$ and the Lipschitz continuous function $f(q) = q + 0.5$ when $d = 1$



Impact of Maximal Speed



- Measurement dependent BSC with $\zeta = 0.05$ and $f(q) = 0.5 + q$
- The larger the maximal speed, the harder to search



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Discussions

- Theorem 1 is tight under maximal speed constraint v_+
- Refines the result by Kaspi et al., TIT 2018 (Theorem 3):
 - Non-asymptotic, non-vanishing vs asymptotic, vanishing
 - Any measurement dependent channel vs a measurement dependent BSC
 - Multidimensional vs one-dimensional
- Strong converse holds

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log \delta^*(n, d, \varepsilon) = \frac{C}{2d}.$$

- Proof ideas: finite blocklength channel coding + analysis of the number of quantized trajectories