

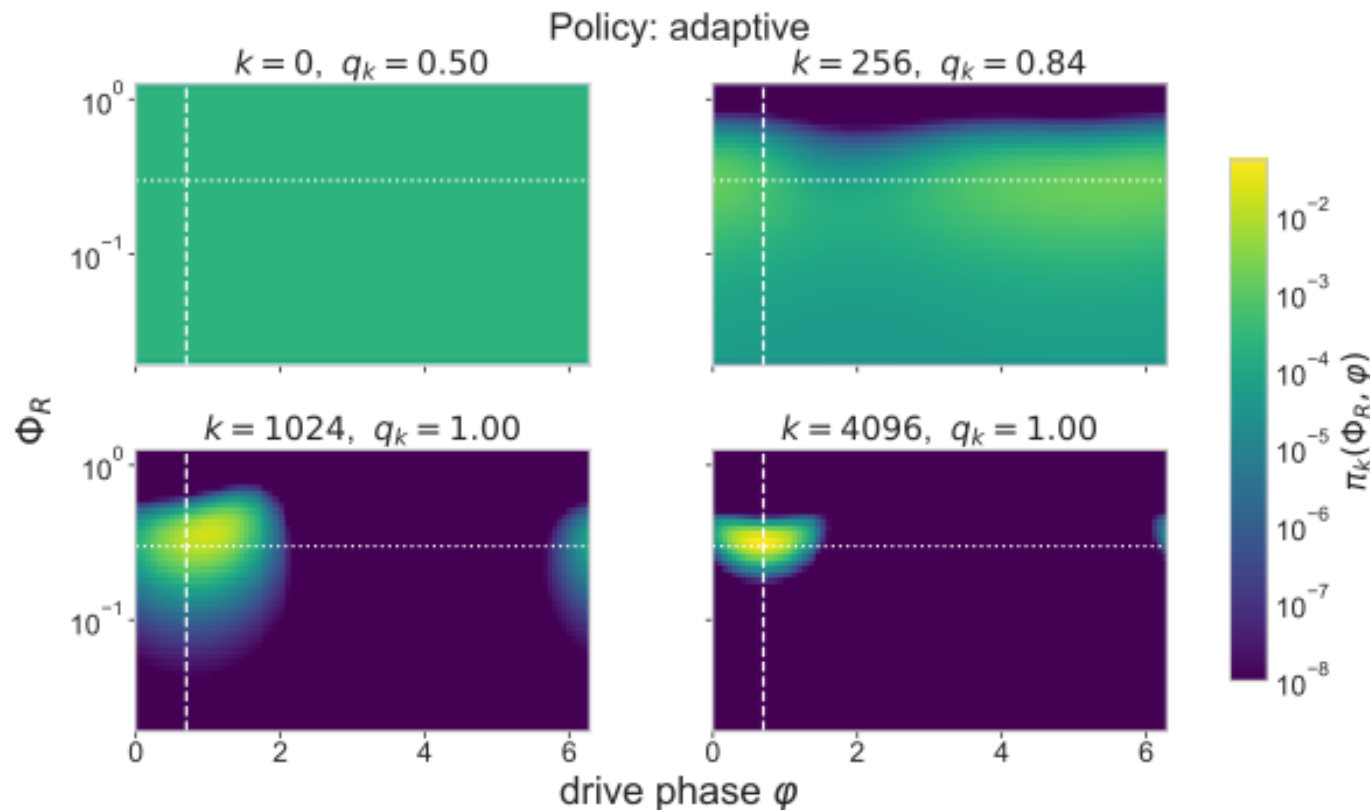


Massachusetts Institute of Technology



MASSACHUSETTS INSTITUTE OF TECHNOLOGY

MIT CENTER FOR THEORETICAL
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Adaptive quantum sensing for wave-like dark matter searches

arXiv: [2608.xxxxx]
In collaboration with



Codex

So Chigusa (Postdoctoral Fellow)

Light Dark World 2026 @ Carleton (7/28/2026)

Wave dark matter signals at direct detection

❖ Coherent oscillation of classical wave dark matter (DM)

$$a(t) = a_0 \sin(m_{\text{DM}}t - m_{\text{DM}}\vec{v}_{\text{DM}} \cdot \vec{x} + \delta)$$

❖ Unknown properties of DM signals

- Frequency = m_{DM}
- Amplitude $\propto a_0$
- Oscillation phase = δ
- Direction

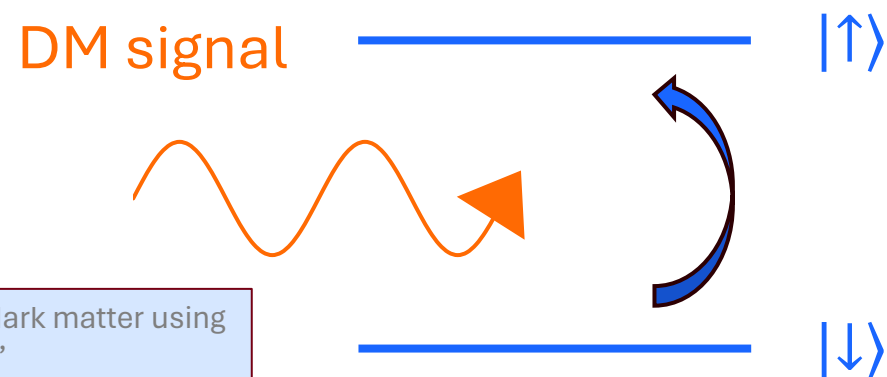
ex) Dark photon polarization

Axion-induced $\vec{B}_{\text{eff}}$

Example: “Detection of hidden photon dark matter using the direct excitation of transmon qubits”
[Chen, et al. PRL '23]

❖ Resonant drive

as example of quantum sensing



$$m_{\text{DM}} \simeq \omega_{\text{qubit}}$$

Classical interpretation and projective measurements

❖ Classical interpretation

- “DM signal flips the qubit spin at probability p ”

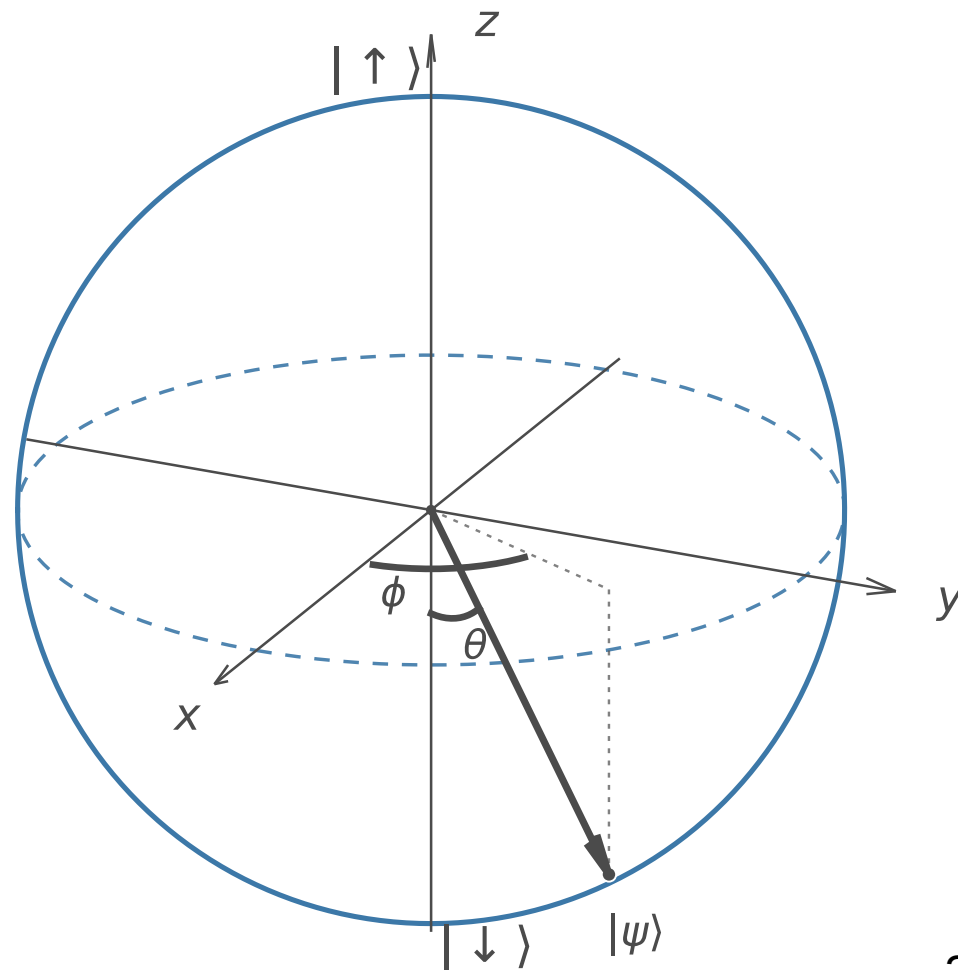
❖ Quantum description

$$|\psi\rangle = \cos\frac{\theta}{2}|\downarrow\rangle + \sin\frac{\theta}{2}e^{i\phi}|\uparrow\rangle$$

- $p = \sin^2\frac{\theta}{2}$
- Additional information from ϕ

❖ Projective measurement

- DOF of readout-axis choice
- “Classical” = z -projective measurement



Longitudinal vs transverse readout

❖ z-readout

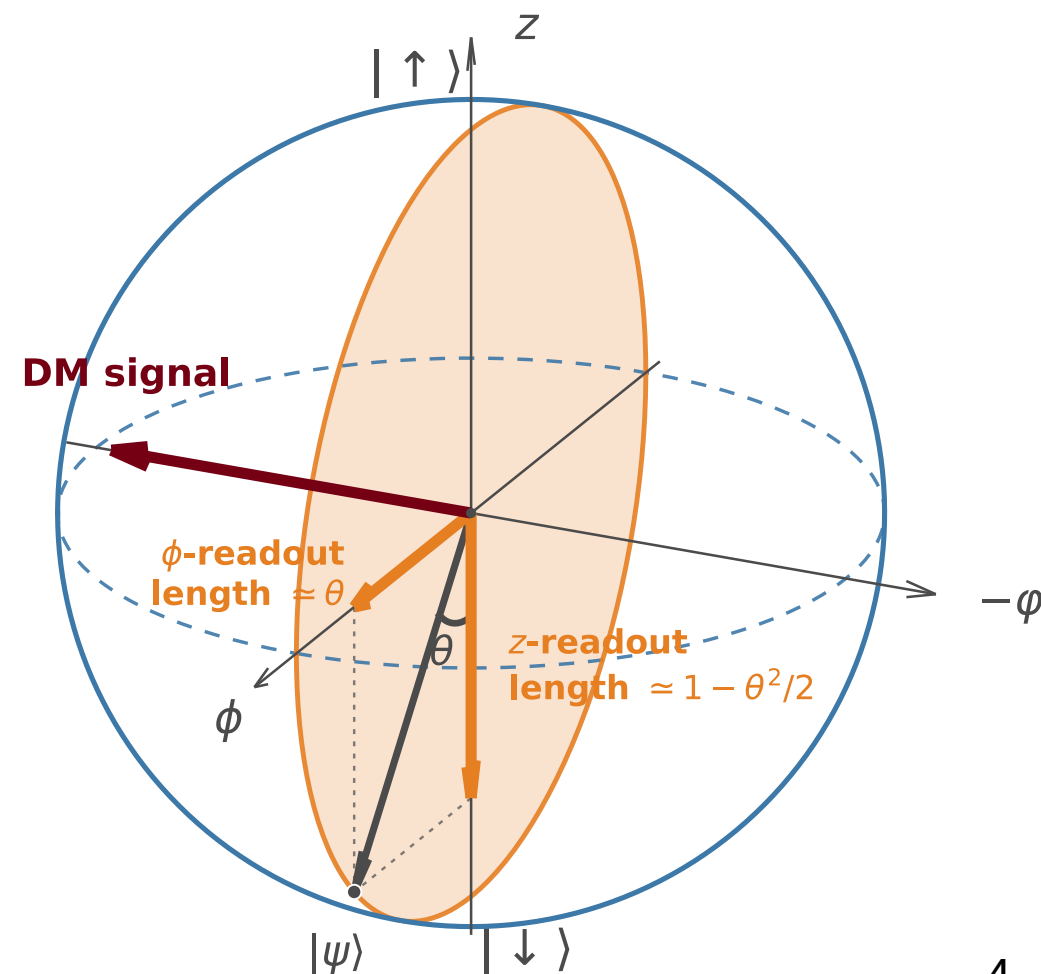
- 👍 Insensitive to unknown phase ϕ
- 👎 Quadratic response $2\sin^2(\theta/2) \propto \theta^2$

❖ Transverse oracle ϕ -readout

- 👍 Linear response $\propto \theta$
- 👎 Need to know ϕ

❖ ϕ is determined by signal direction

- For $\vec{E}_{\text{sig}} = \cos \varphi \hat{x} + \sin \varphi \hat{y}$
- $\phi = \varphi + \pi/2$ is unknown $\Rightarrow$ learn it!



Adaptive strategy

❖ During repeated measurements with n_{shots} total shots,

Start with z-readout

- Phase-insensitive signal search

Phase learning

- Insert trial shots (**information gain**) to learn the signal phase φ

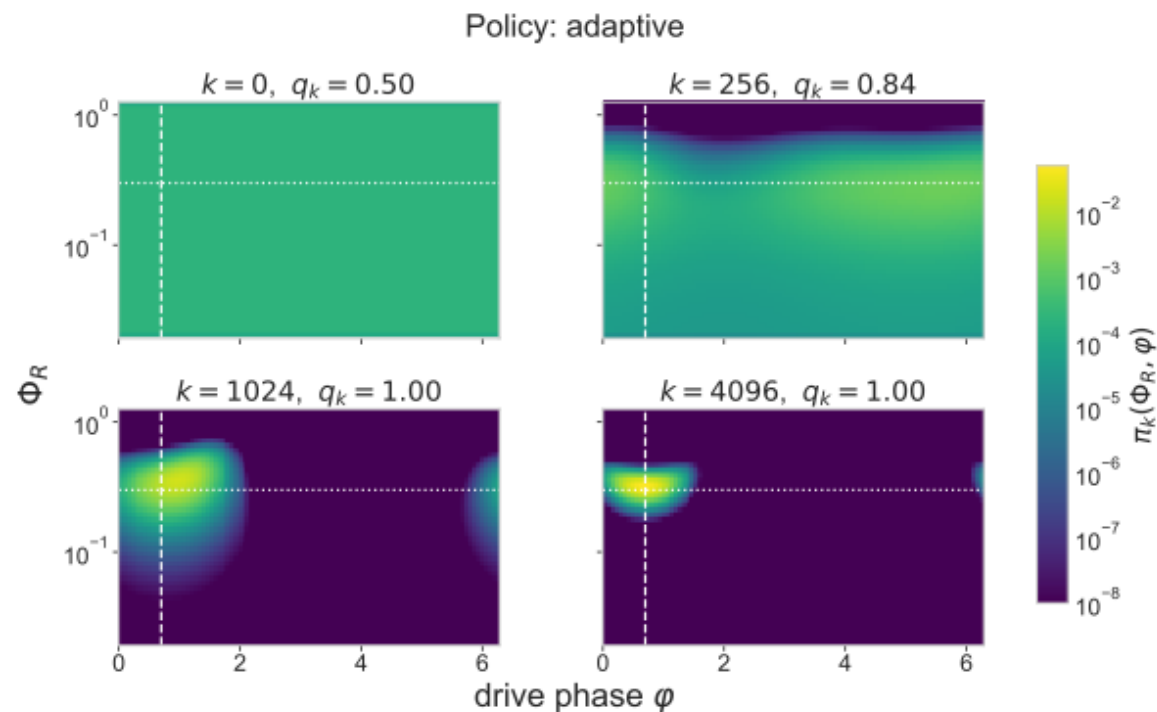
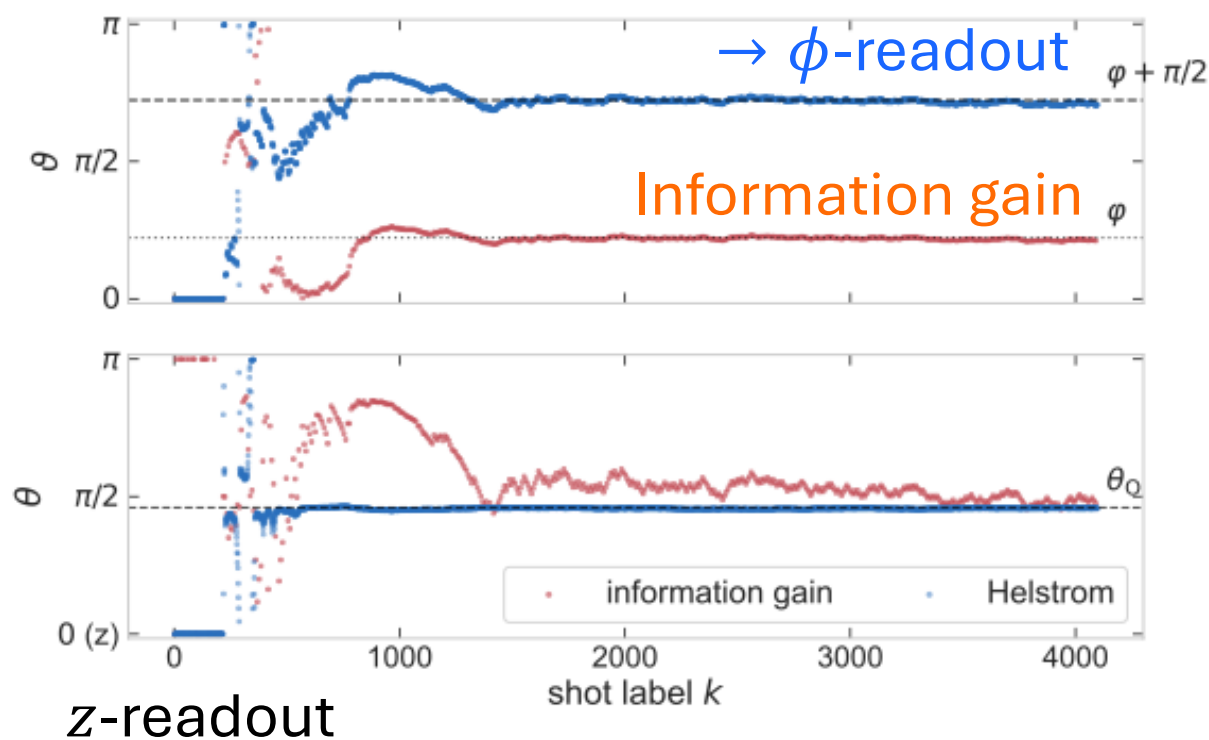
Converge to transverse oracle

- The optimal **ϕ -readout** for signal detection

❖ Resilient to fake positives/look-elsewhere effects

Adaptive trajectory

- ❖ Noisy environment simulation: Relaxation (T_1, T_2), Readout contrast & error
- ❖ Bayesian analysis tracking prior/posterior distribution of nuisance parameters



Sensitivities

❖ Statistical formulation

- Composite hypothesis test

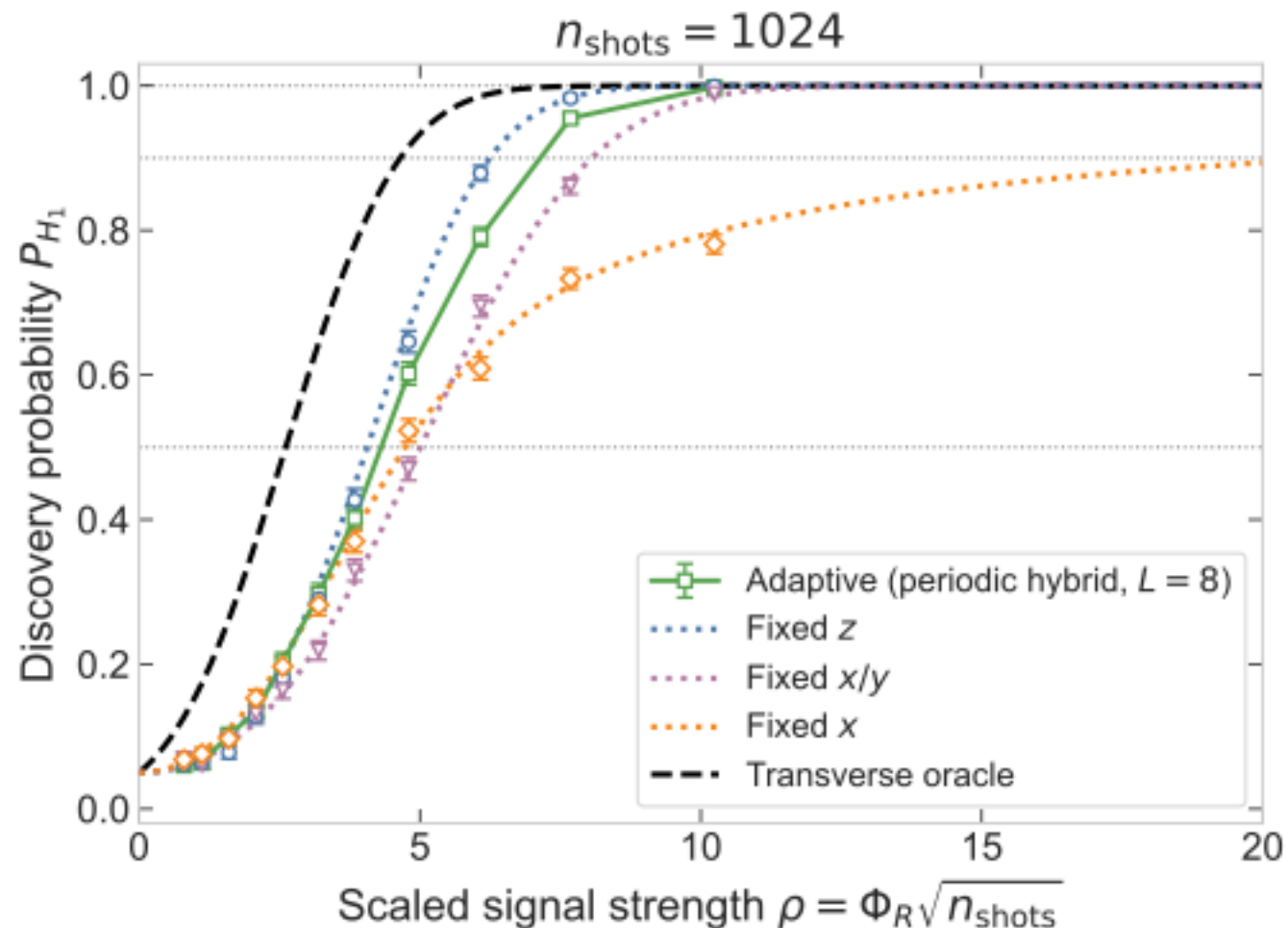
👎 Noise only vs 👍 Noise+DM

- Asymmetric testing

Keep false-positive rate $< 5\%$

❖ Best strategy for increasing n_{shots}

1. z-readout for small n_{shots}
2. Adaptive for large n_{shots}



Sensitivities

❖ Statistical formulation

- Composite hypothesis test

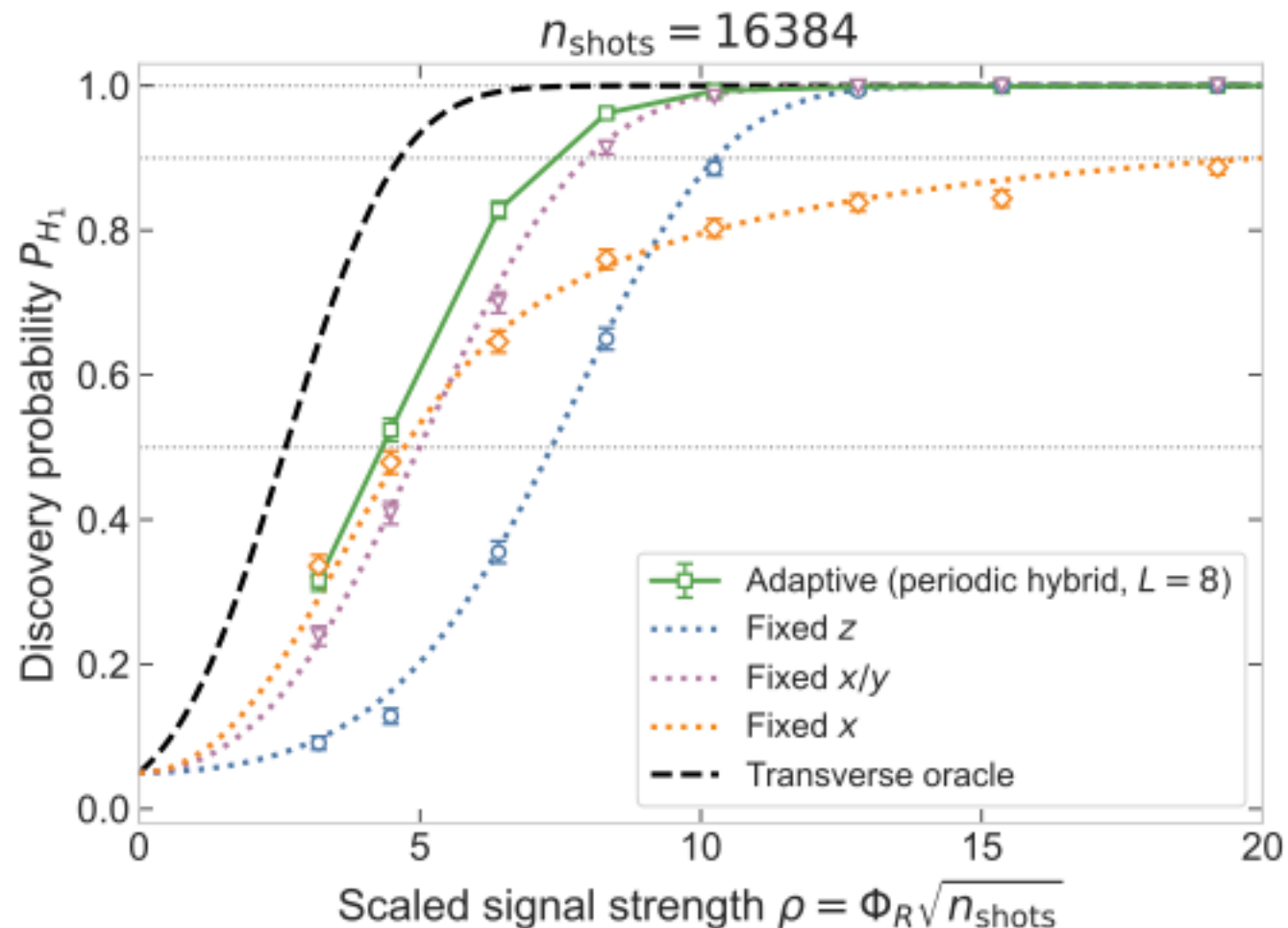
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❖ Best strategy for increasing n_{shots}

1. z-readout for small n_{shots}
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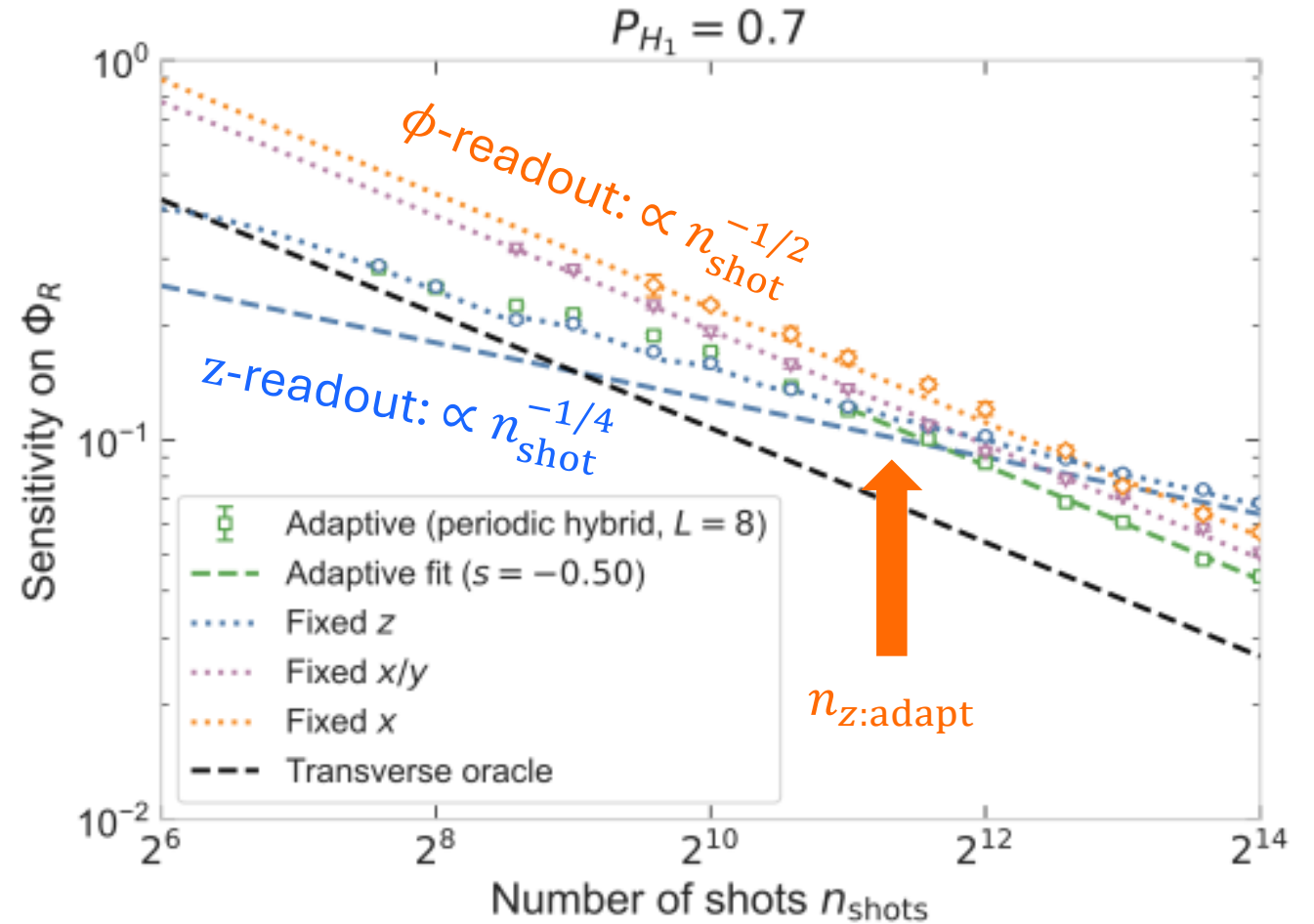
Sensitivity scaling and comparison

❖ Crossings due to different scaling

P_{H_1}	$n_{z:O_1}$	$n_{z:adapt}$	$n_{z:xy}$	$n_{z:x}$
0.50	2.97×10^2	1.53×10^3	3.99×10^3	3.20×10^3
0.70	5.17×10^2	1.99×10^3	5.54×10^3	9.50×10^3
0.90	9.41×10^2	3.12×10^3	8.22×10^3	3.21×10^5
0.95	1.19×10^3	3.51×10^3	9.69×10^3	3.97×10^6

❖ Phase information efficiency

$$\begin{aligned}
 \bullet f_{\text{pol}} &\equiv 2 \left\langle \left\{ \hat{n}_{\text{sig}} \cdot (\hat{n}_{\text{meas}} \times \hat{z}) \right\}^2 \right\rangle \\
 &= \begin{cases} 1 & \text{(fixed x/y)} \\ 1.15 - 1.32 & \text{(adaptive)} \\ 2 & \text{(oracle)} \end{cases}
 \end{aligned}$$



Takeaways and discussions

❖ Implication to decision making

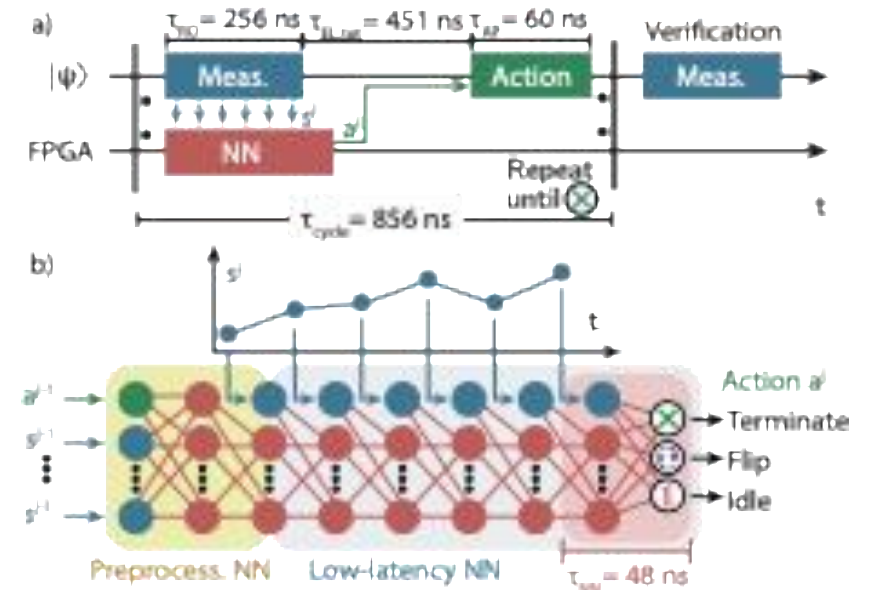
- Adaptive is best for $n_{\text{shots}} \gtrsim n_{\text{z:adapt}}$
- $n_{\text{z:adapt}}$ is independent of DM properties

❖ Online update, NOT offline analysis

- Info discarded by measurement never recovered

❖ Prospects

- Global optimization $\Rightarrow$ DRL
- Optimize initial state, interrogation time, etc. $\Rightarrow$ VQA



“Realizing a deep reinforcement learning agent discovering real-time feedback control strategies for a quantum system”

[Reuer, et al. Nat. Comm. '23]

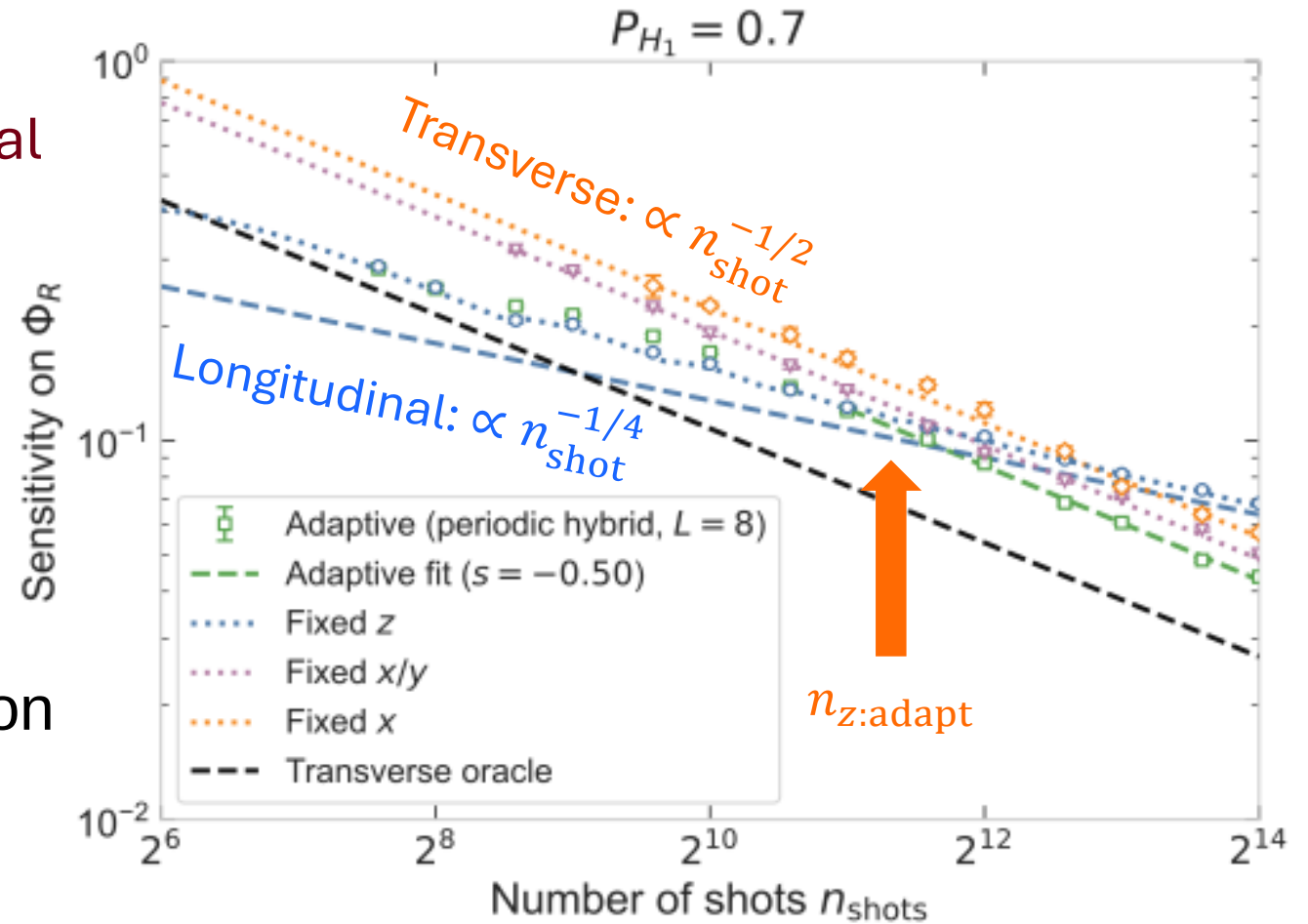
Conclusion

❖ Main message

- Adaptive strategy can be beneficial for DM searches
- The best search strategy determined solely by detector setups; deserves a consideration during experimental design

❖ Fascinating interdisciplinary direction

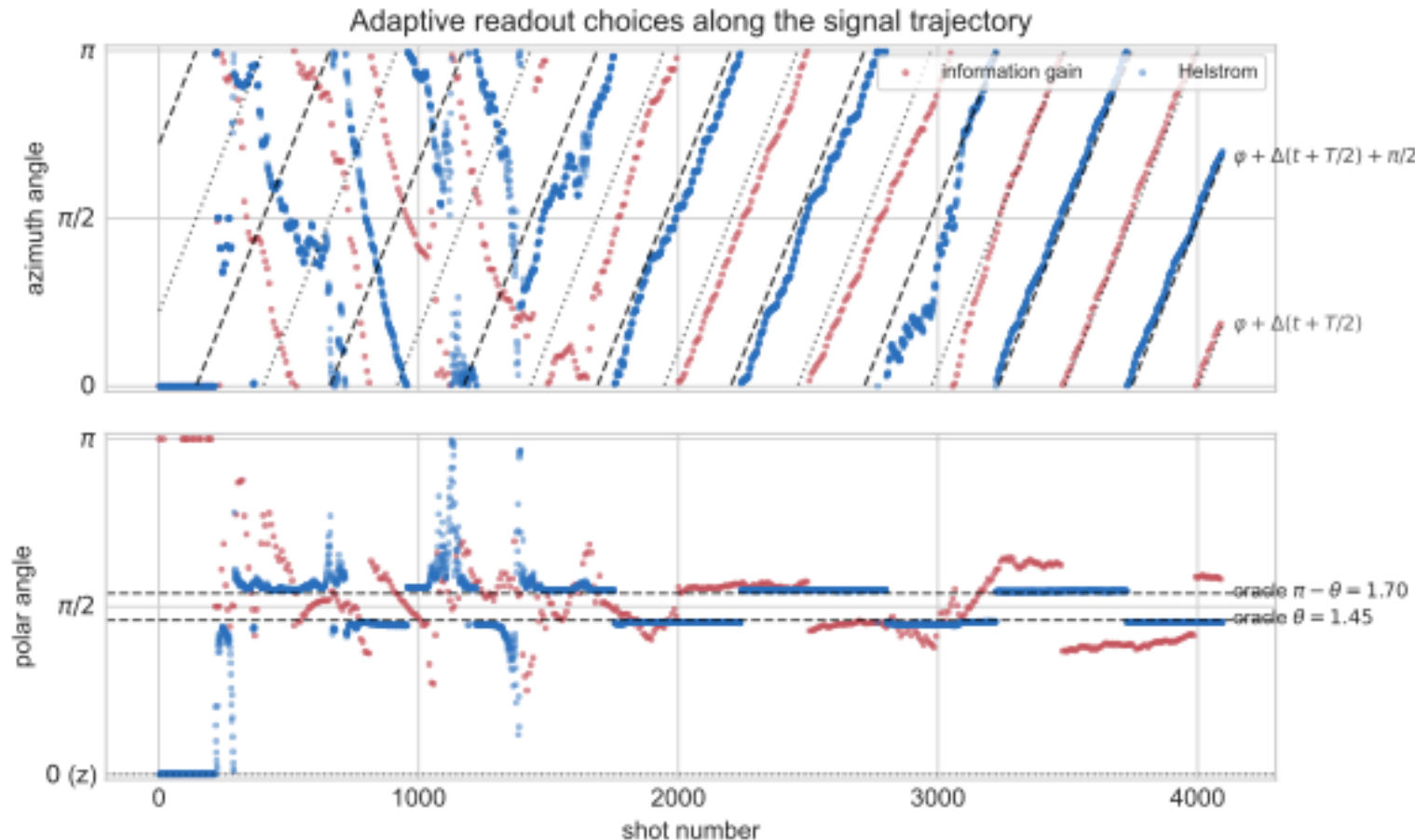
Particle × Quantum × ML



Backup slides

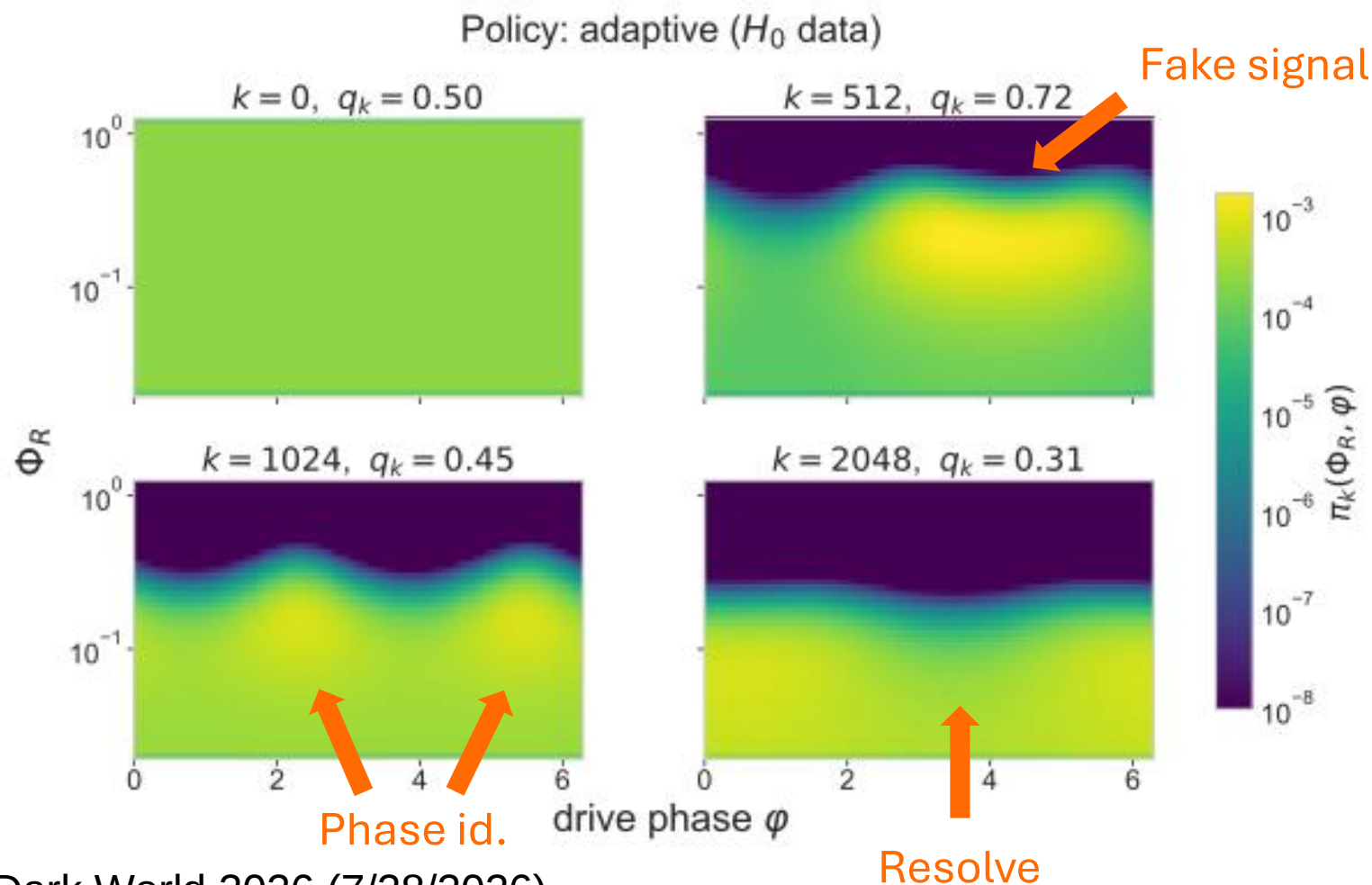
Detuning effects

❖ Detuning Δ makes optimal readout time-dependent: $\phi(t) = \varphi + \pi/2 + \Delta t$



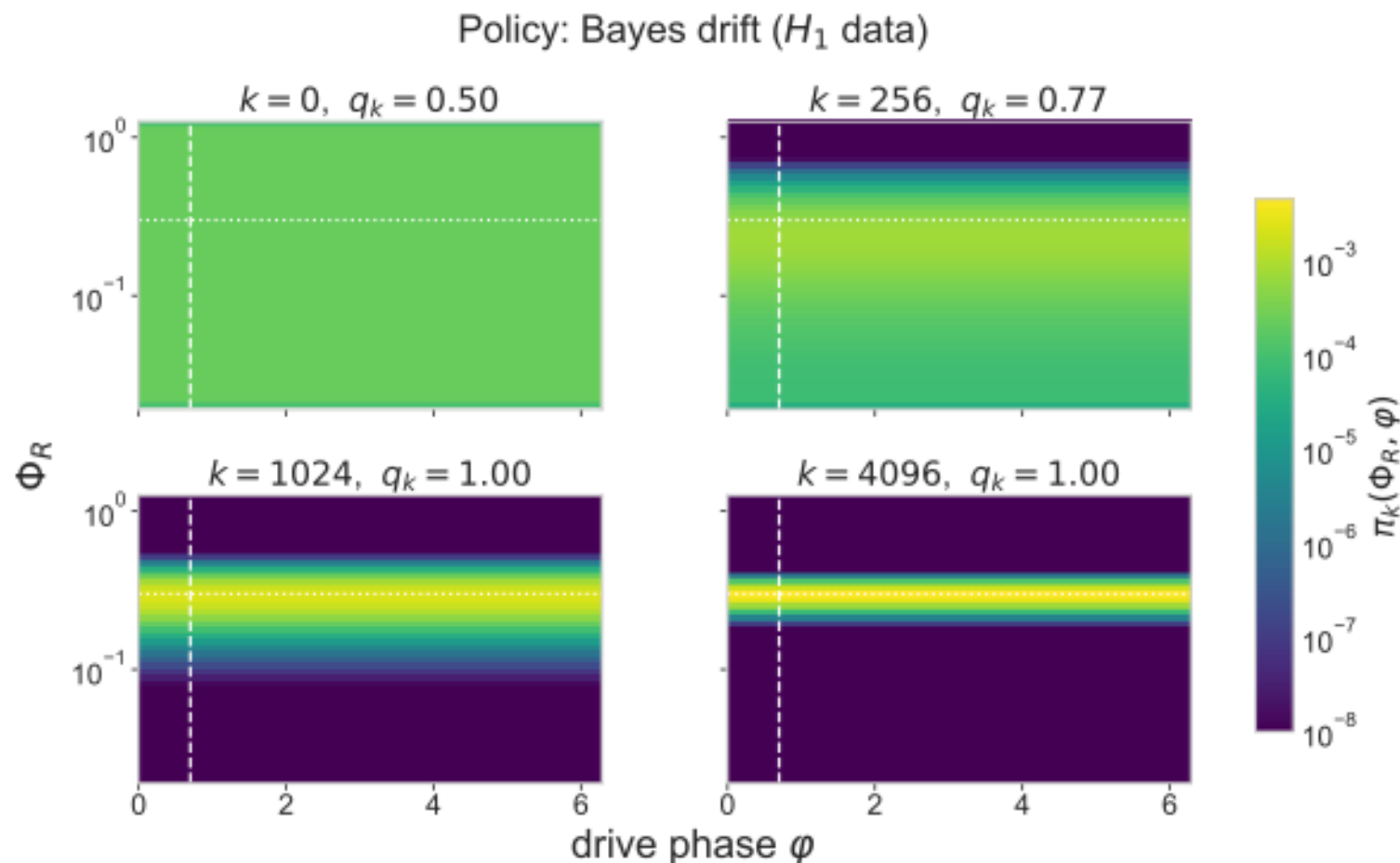
Visualizing look-elsewhere effects

❖ No signal simulation



Repeated local optimization $\neq$ global optimization

❖ Fixed- z is the best with unknown phase $\Leftrightarrow$ No phase information is learned



Information theory for hypothesis test

❖ Type-II error (false negative) rate decays exponentially

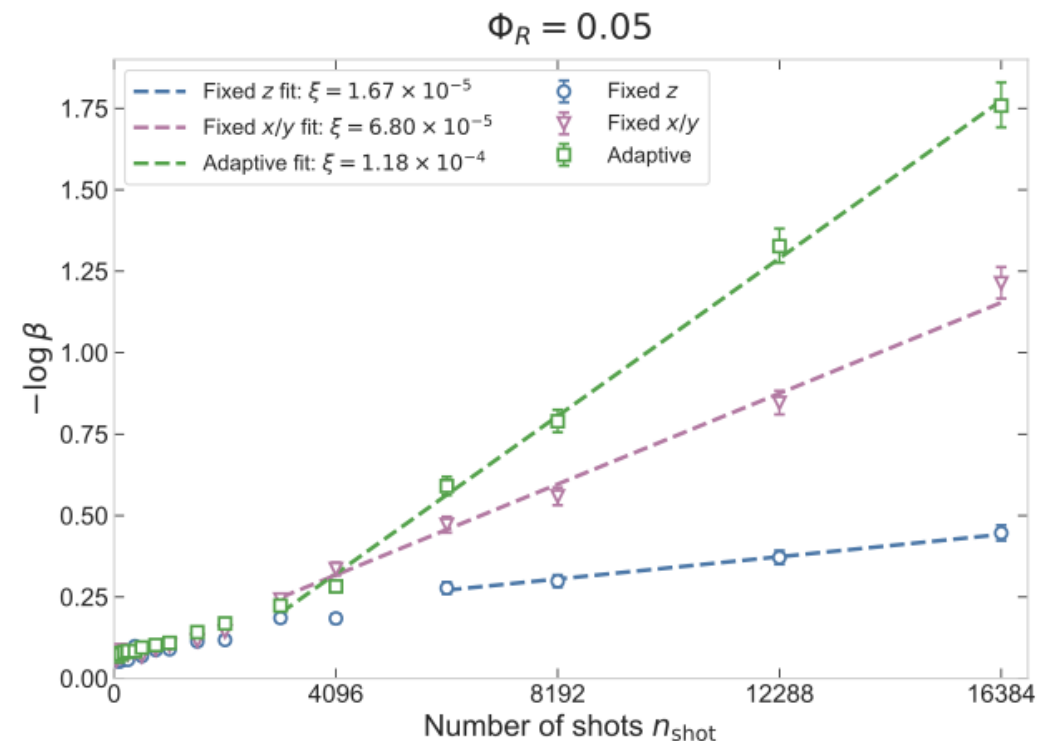
- $\beta \rightarrow \exp[-n\xi + o(n)]$
- $\xi = D_{\text{KL}}(P_0||P_1)$: Kullback–Leibler divergence

❖ Relevant QI quantity: **relative entropy** $D(\rho_0||\rho_1)$

- **Max extractable information:** $D_{\text{KL}} \leq D$

cf: Fisher information for measurement

cf: Chernoff bound for symmetric testing



Assumed experimental setups

❖ Detector performance (optimistic transmon qubit)

- $T_1 = T_2 = T$: interrogation time
- Measurement contrast $C = 0.999$
- Readout error rate $\epsilon = 0.05 \%$

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❖ For false-positive rate $\alpha = 5 \%$ and detection power $P = 1 - \beta = 95 \%$

$$n_{z:\text{adapt}} \simeq \frac{19.2749}{f_{\text{pol}}^2((1 - C) + 2\epsilon)} [1 + O(1 - C, \epsilon)]$$