Re-calibration of quantum devices by RL

An application to coherent-state receiver

arXiv:2404.10726

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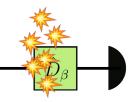
Joint work w/ Tomás Crota Fernando Vilariño Lorena Rebón Mauricio Matera

This talk in a nutshell

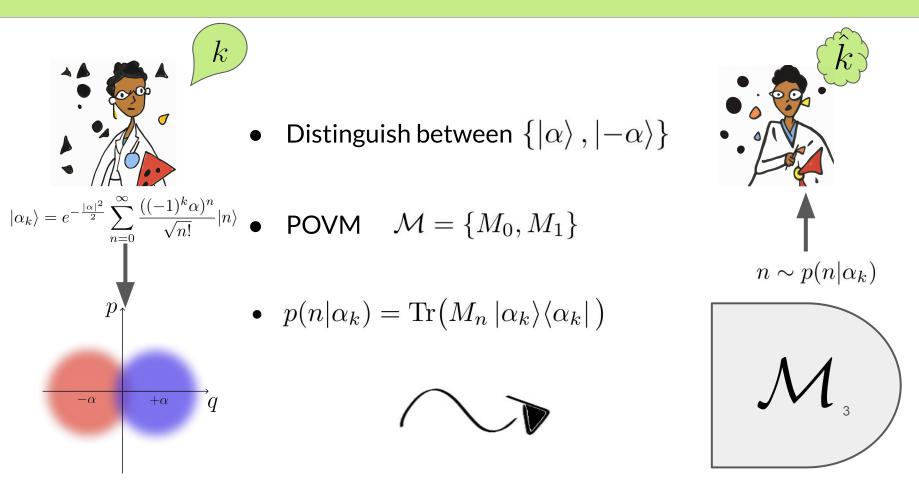
- What happens if the environment changes?
- Detect the change ?
 Adapt the policy ?
 What happens w/ Q-values?
 Which strategies can we adopt ?
- Binary quantum-state discrimination







Coherent-state discrimination



Success probability & Helstrom bound

• Success probability; g(n) juessing rule

$$P_s(\mathcal{M},g) = \sum_n p(n|\alpha_{\hat{k}}) \operatorname{pr}(\alpha_{\hat{k}})|_{\hat{k}=g(n)}$$

Helstrom bound

k

 $|\alpha_k\rangle = e^{-\frac{|\alpha|^2}{2}} \sum_{n=1}^{\infty} \frac{((-1)^k \alpha)^n}{\sqrt{n!}} |n\rangle$

p

 $-\alpha$

 $+\alpha$

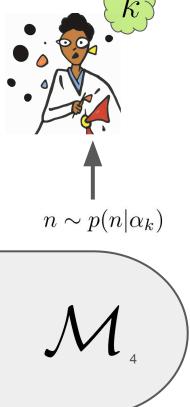
q

$$P_{s}(\mathcal{M},g) \leq P_{s}(\mathcal{M}^{*},g^{*})$$
$$= \frac{1}{2} \Big(1 + ||p_{0}\rho_{0} - p_{1}\rho_{1}||_{1} \Big)$$

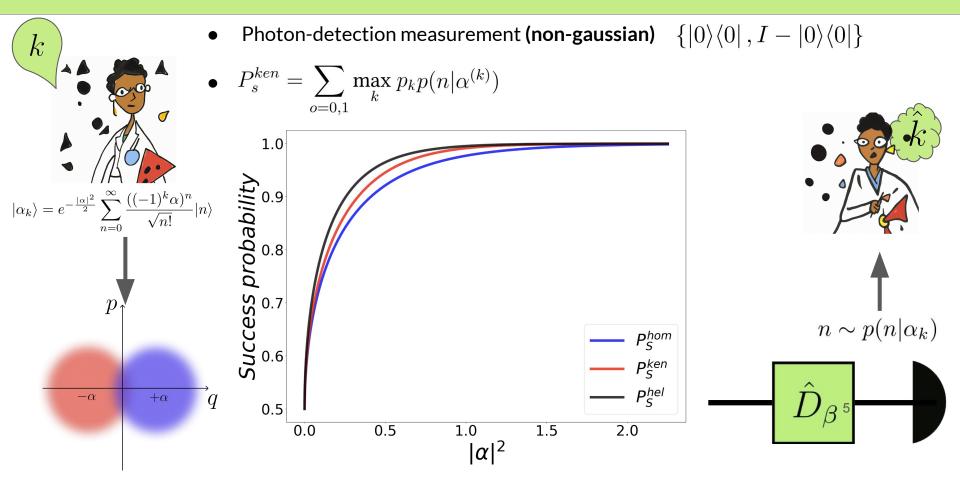
• Optimal: $|\text{cat}
angle \propto |lpha
angle + |-lpha
angle$

Can be realized w/ lineal optics + on/off detectors:

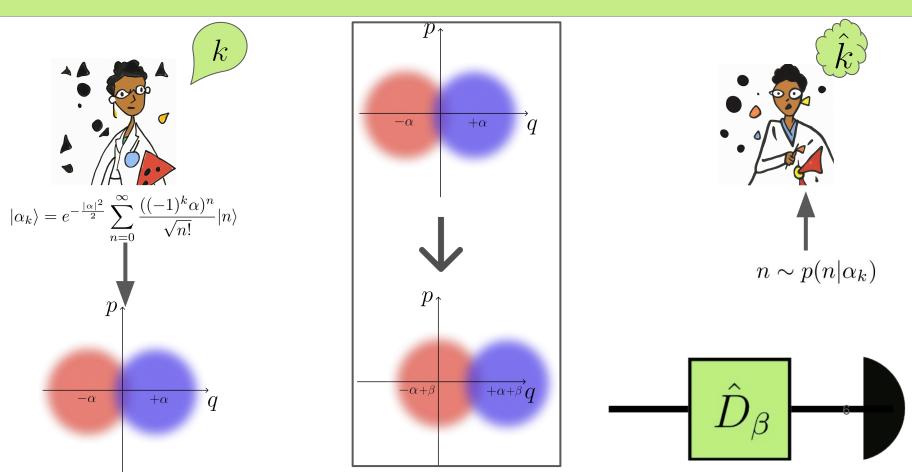
• Feed(forward) protocol on consecutive <u>Kennedy</u> receivers



Kennedy receiver



Kennedy receiver

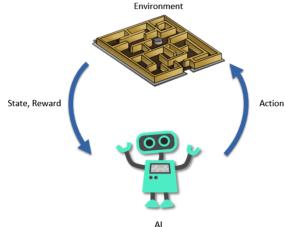


RL calibration

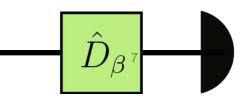
Estimate of how good a displacement is by trying it

 $\hat{Q}($

This is done by Q-learning, which updates these estimates

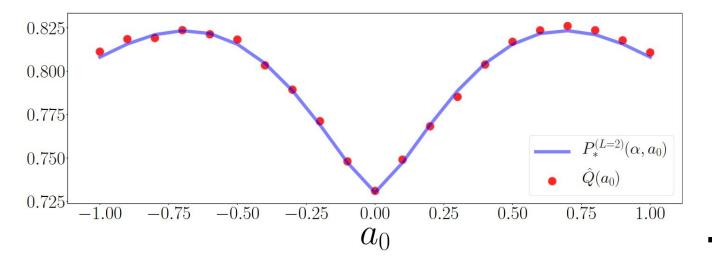


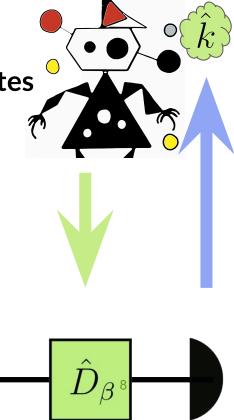
$$s_{\ell}, a_{\ell}) \leftarrow (1 - \tilde{\alpha})\hat{Q}(s_{\ell}, a_{\ell}) \\ + \tilde{\alpha} \left(r_{\ell+1} + \gamma \max_{a' \in \mathcal{A}(s_{\ell+1})} \hat{Q}(s_{\ell+1}, a') \right)$$



RL calibration

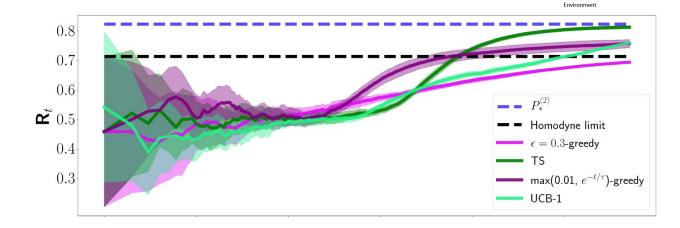
- Estimate of how good a displacement is by trying it
- This is done by Q-learning, which updates these estimates

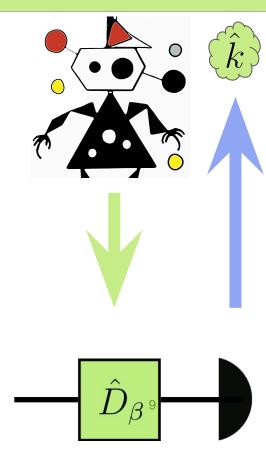




RL calibration

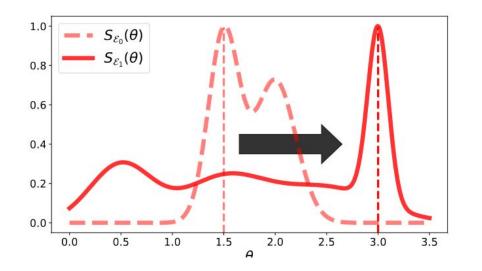
- Agent guesses, reward 1/0 if ok/wrong
- Noise robustness ([PRR, 033295 (2020)])





But if the environment changes?





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De-calibration witness

• How to realize a change occurs?

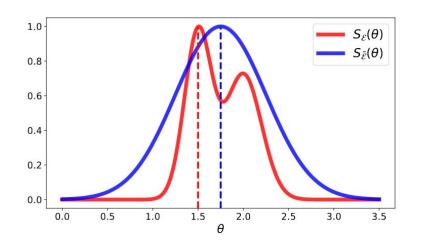
• Monitor the reward \rightarrow inaccessible when "deploying"

• Example: monitor empirical outcome probabilities



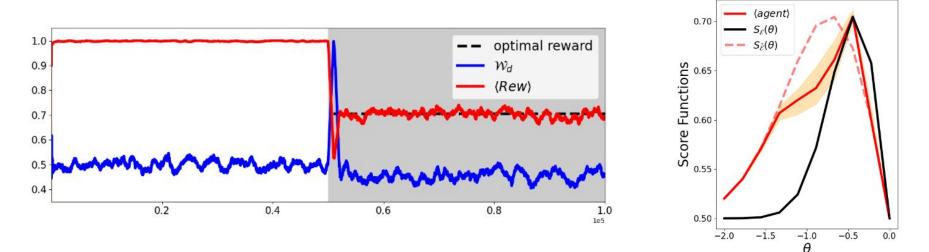
Q-values

- Would them adapt? → Standard Q-learning yes & no (granted exploration)
- Can we do some sort of approximation?
- [Other approaches \rightarrow learn map] $\mathcal{E}_0 \rightarrow \mathcal{E}_1$
- Here: effective re-initialization of the Q-values + fine-tune w/ RL



Case-study

- The displacements are faulty $\beta \rightarrow \lambda \beta$ D_{β} and intensity $|\alpha|^2$ changes
- Monitor empirical outcome probabilities (decalibration witness) $\,\hat{p}(n=1)\,$
- Effective model: success probabilities (of un-faulty device!)



Conclusion & questions

- We study how RL adapts to a sudden change in environment, and develop simple strategies to complement Q-learning
- We tackle a very simple case, the Kennedy receiver
- Alternative scenarios & problems to try this machinery? *e.g.* quantum control

I'm happy to talk :) mbilkis@cvc.uab.cat



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