

Parallelization of Sequential Quantum Channel Discrimination in the Non-Asymptotic Regime

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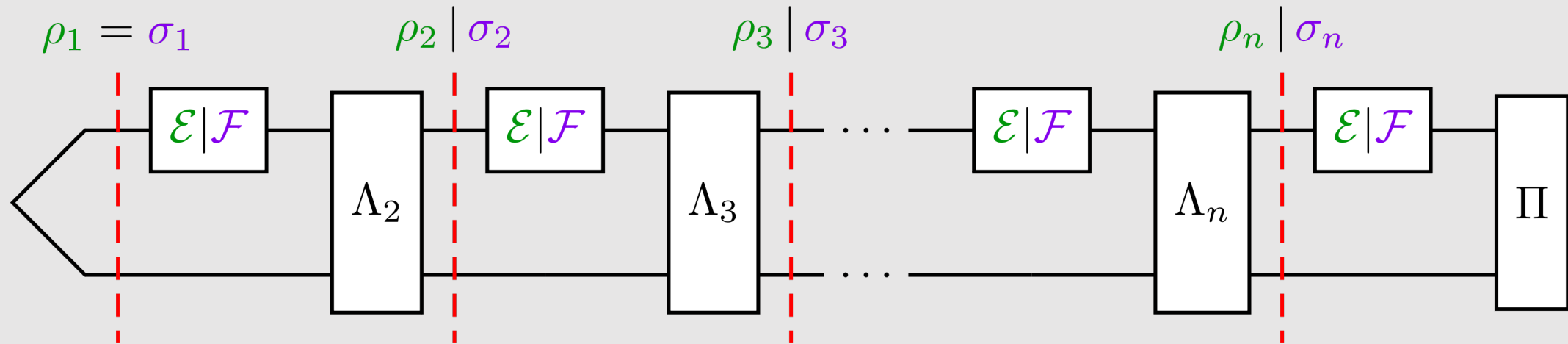
The task

- Black-box promised to be one of two possible channels



- Find out which one
- You are allowed to use the black-box a finite number of times (n)

The most general setup



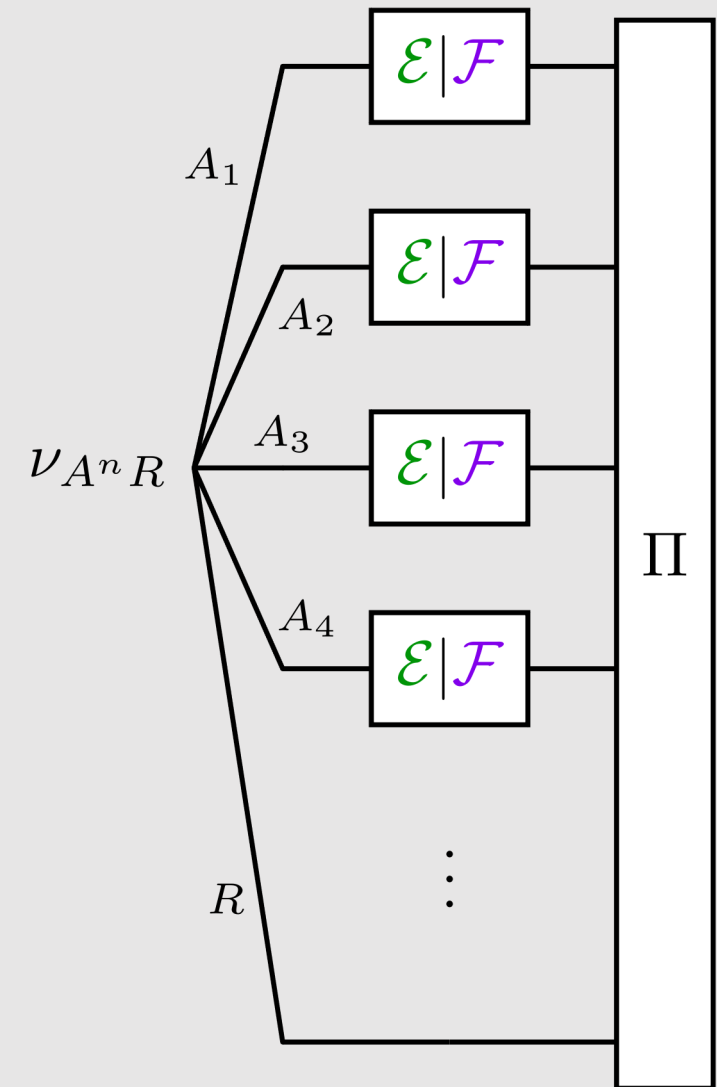
- Input states will in general depend on previous channel outputs
→ *adaptive strategy*

Way simpler – A *parallel strategy*

Joint input state chosen at the beginning.

Reference system can be chosen isomorphic to A^n

Every parallel strategy can be written as an adaptive strategy.



Interlude – Discriminating States

To discriminate between states ρ and σ , do a binary POVM $\{\Pi, \mathbb{1} - \Pi\}$ which will have errors:

$$\alpha(\rho, \Pi) = \text{Tr}(\rho(\mathbb{1} - \Pi))$$

$$\beta(\sigma, \Pi) = \text{Tr}(\sigma\Pi)$$

The asymmetric setting

Best type II error when type I error stays below ε :

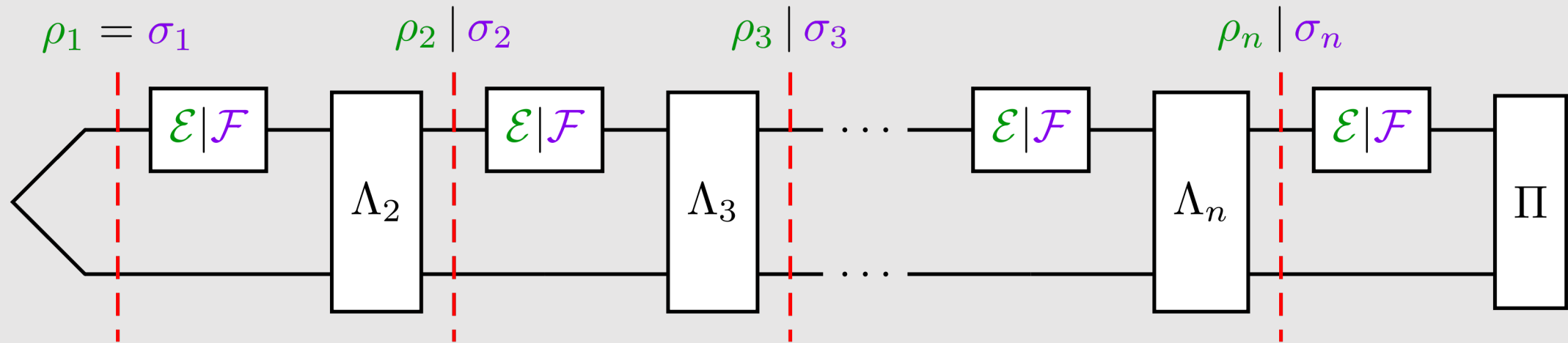
$$\beta(\varepsilon, \rho, \sigma) := \min \{ \text{Tr}(\Pi\sigma) \mid \Pi : 0 \leq \Pi \leq \mathbb{1}, \text{Tr}(\rho(\mathbb{1} - \Pi)) \leq \varepsilon \}$$

$$D_H^\varepsilon(\rho \parallel \sigma) := -\log \beta(\varepsilon, \rho, \sigma)$$

With n copies of the state: Type II error rate per used copy

$$\frac{1}{n} D_H^\varepsilon(\rho^{\otimes n} \parallel \sigma^{\otimes n}) = -\frac{1}{n} \log \beta(\varepsilon, \rho^{\otimes n}, \sigma^{\otimes n})$$

Error Exponents – Adaptive Strategy



Error rate per channel use of this strategy:

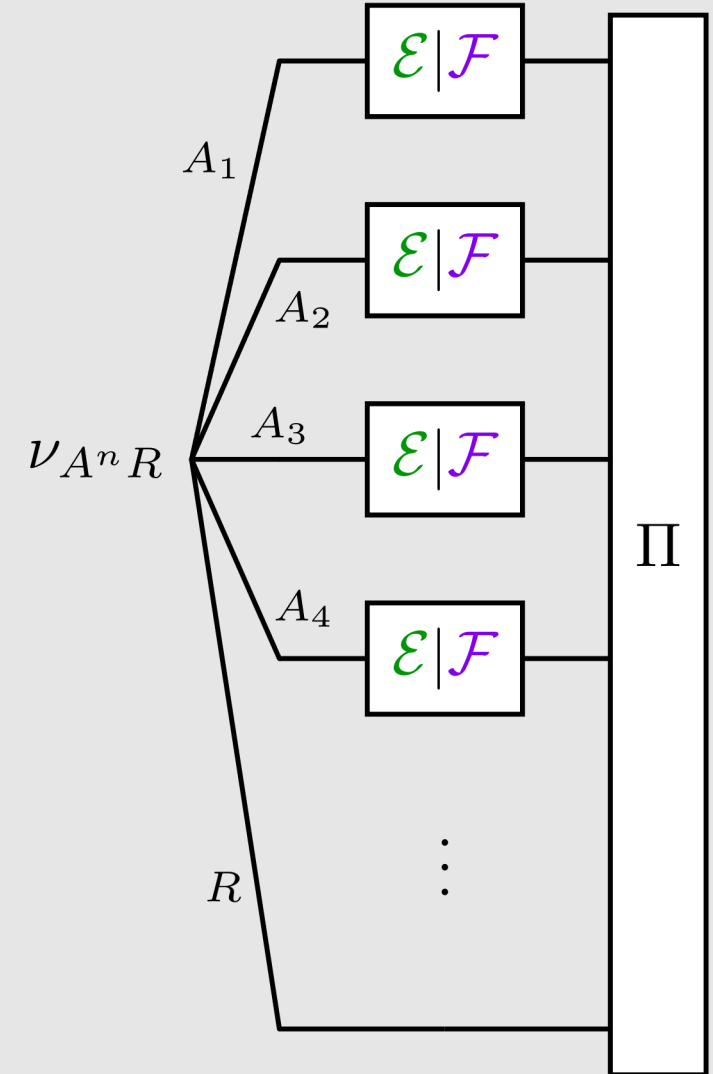
$$\frac{1}{n} D_H^\varepsilon((\text{id}_R \otimes \mathcal{E})(\rho_n) \| (\text{id}_R \otimes \mathcal{F})(\sigma_n))$$

Error Exponents – Parallel strategy

$$\frac{1}{n} D_H^\varepsilon \left((\text{id}_R \otimes \mathcal{E}^{\otimes n})(\nu) \parallel (\text{id}_R \otimes \mathcal{F}^{\otimes n})(\nu) \right)$$

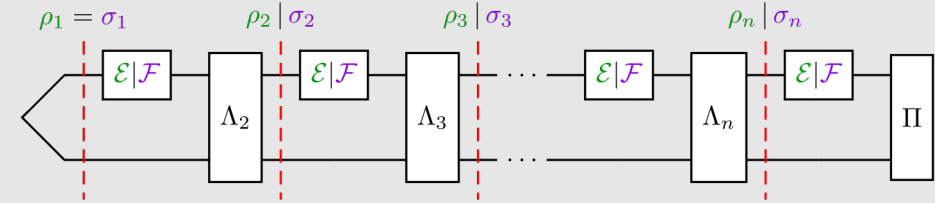
From now on, drop identities, i.e. write:

$$\frac{1}{n} D_H^\varepsilon \left(\mathcal{E}^{\otimes n}(\nu) \parallel \mathcal{F}^{\otimes n}(\nu) \right)$$



What was known already – Stein Exponents

Optimization over all sequential strategies



$$\lim_{\varepsilon \rightarrow 0} \lim_{n \rightarrow \infty} \sup_{\rho_1, \{\Lambda_i\}} \frac{1}{n} D_H^\varepsilon(\mathcal{E}(\rho_n) \| \mathcal{F}(\sigma_n)) = D_A(\mathcal{E} \| \mathcal{F})$$

$$= D_{\text{reg}}(\mathcal{E} \| \mathcal{F}) = \lim_{\varepsilon \rightarrow 0} \lim_{m \rightarrow \infty} \sup_{\nu} \frac{1}{m} D_H^\varepsilon(\mathcal{E}^{\otimes m}(\nu) \| \mathcal{F}^{\otimes m}(\nu))$$

Best achievable type II error rate such that the type I error goes to zero as $n \rightarrow \infty$.

Optimization over all parallel strategies

The question

- What happens at finite n ?
- How many channel uses does a parallel strategy need to achieve a similar rate as a (given) adaptive strategy?

Our result

Theorem. Let \mathcal{E}, \mathcal{F} s.t. $D_{\max}(\mathcal{E} \parallel \mathcal{F}) < \infty$. Given $\{\rho_i, \sigma_i\}_{i=1}^n$,
 $\forall \alpha_S, \alpha_P \in (0, 1), \forall m \in \mathbb{N} : \exists \nu \in \mathcal{D}(R^m A^m) :$

$$\frac{1}{m} D_H^{\alpha_P}(\mathcal{E}^{\otimes m}(\nu) \parallel \mathcal{F}^{\otimes m}(\nu)) \geq \frac{1 - \alpha_S}{n} D_H^{\alpha_S}(\mathcal{E}(\rho_n) \parallel \mathcal{F}(\sigma_n)) - \frac{Cn}{\sqrt{m}} \log\left(\frac{8}{\alpha_P}\right) - \frac{1}{n}$$

with:


$$C := 7 \log\left(2^{D_2(\mathcal{E} \parallel \mathcal{F})} + 2\right) \leq 7 \log\left(2^{D_{\max}(\mathcal{E} \parallel \mathcal{F})} + 2\right)$$

Prerequisites – A chain rule

$$D_{\max}(\rho\|\sigma) := \log \inf \{ \lambda \in \mathbb{R} \mid \rho \leq \lambda\sigma \}$$

$$D_{\max}(\mathcal{E}(\rho)\|\mathcal{F}(\sigma)) \leq D_{\max}(\rho\|\sigma) + D_{\max}(\mathcal{E}(\rho)\|\mathcal{F}(\rho))$$

$$D_{\max}^{\varepsilon}(\rho\|\sigma) := \inf_{\tilde{\rho} \in B_{\varepsilon}^{\circ}(\rho)} D_{\max}(\tilde{\rho}\|\sigma)$$

 ε -ball of *normalized* states in purified distance

$$D_{\max}^{\varepsilon+\varepsilon'}(\mathcal{E}(\rho)\|\mathcal{F}(\sigma)) \leq D_{\max}^{\varepsilon}(\rho\|\sigma) + D_{\max}^{\varepsilon'}(\mathcal{E}(\nu)\|\mathcal{F}(\nu))$$

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$$D_{\max}^{\varepsilon+\varepsilon'}(\mathcal{E}(\rho)\|\mathcal{F}(\sigma)) \leq D_{\max}^{\varepsilon}(\rho\|\sigma) + D_{\max}^{\varepsilon'}(\mathcal{E}(\nu)\|\mathcal{F}(\nu))$$

$$\nu \leq 2^{D_{\max}^{\varepsilon}(\rho\|\sigma)} \sigma \quad \Rightarrow \quad \mathcal{F}(\nu) \leq 2^{D_{\max}^{\varepsilon}(\rho\|\sigma)} \mathcal{F}(\sigma)$$

$$\tau \leq 2^{D_{\max}^{\varepsilon'}(\mathcal{E}(\nu)\|\mathcal{F}(\nu))} \mathcal{F}(\nu)$$

$$\tau \leq 2^{D_{\max}^{\varepsilon'}(\mathcal{E}(\nu)\|\mathcal{F}(\nu)) + D_{\max}^{\varepsilon}(\rho\|\sigma)} \mathcal{F}(\sigma)$$

$$P(\tau, \mathcal{E}(\rho)) \leq P(\tau, \mathcal{E}(\nu)) + P(\mathcal{E}(\nu), \mathcal{E}(\rho)) \leq \varepsilon' + P(\nu, \rho) \leq \varepsilon' + \varepsilon.$$

Prerequisites – Hypothesis Test. Rel. Entropy

$$D_{\max}^{\varepsilon}(\rho\|\sigma) \leq D_H^{1-\varepsilon^2}(\rho\|\sigma) - \log(1 - \varepsilon^2)$$

Anshu, Berta, Jain, Tomamichel, A Minimax Approach to One-Shot Entropy Inequalities. *J. Math. Phys.* **2019**, *60* (12), 122201.

$$D_H^{\varepsilon}(\rho\|\sigma) \leq \frac{1}{1-\varepsilon} (D(\rho\|\sigma) + h(\varepsilon))$$

Well-known, e.g. Hiai and Petz (1991), Hayashi (2006), Wang and Renner (2012)

Prerequisites – AEP

$$\frac{1}{n} D_{\max}^{\varepsilon}(\rho^{\otimes n} \| \sigma^{\otimes n}) \geq D(\rho \| \sigma) - \frac{4c_{\gamma}(\rho \| \sigma)}{\sqrt{n}} \log\left(\frac{2}{1-\varepsilon}\right)$$

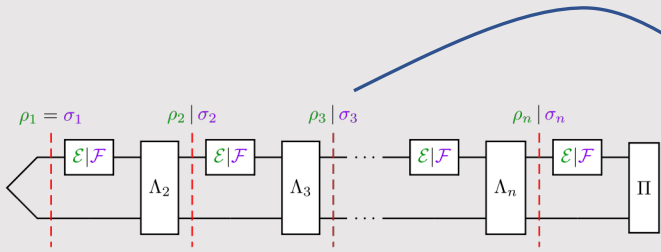
$$\frac{1}{n} D_{\max}^{\varepsilon}(\rho^{\otimes n} \| \sigma^{\otimes n}) \leq D(\rho \| \sigma) + \frac{4c_{\gamma}(\rho \| \sigma)}{\sqrt{n}} \log\left(\frac{2}{\varepsilon}\right) + \frac{1}{n} \log\left(\frac{1}{1-\varepsilon^2}\right)$$

$$c_{\gamma}(\rho \| \sigma) := \frac{1}{\gamma} \log\left(2^{\gamma D_{1+\gamma}(\rho \| \sigma)} + 2^{-\gamma D_{1-\gamma}(\rho \| \sigma)} + 1\right) \quad \gamma \in (0, 1]$$

Proof of our Theorem

$$\frac{1 - \alpha_S}{n} D_H^{\alpha_S}(\mathcal{E}(\rho_n) \| \mathcal{F}(\sigma_n)) \leq \frac{1}{n} D(\mathcal{E}(\rho_n) \| \mathcal{F}(\sigma_n)) + \dots$$

$D_H^\varepsilon(\rho \| \sigma) \leq \frac{1}{1 - \varepsilon} (D(\rho \| \sigma) + h(\varepsilon))$



$$\leq \frac{1}{n} \sum_{k=1}^n [D(\mathcal{E}(\rho_k) \| \mathcal{F}(\sigma_k)) - D(\rho_k \| \sigma_k)] + \dots$$

$$\leq D(\mathcal{E}(\rho_\ell) \| \mathcal{F}(\sigma_\ell)) - D(\rho_\ell \| \sigma_\ell) + \dots$$

$$\frac{1}{n} D_{\max}^\varepsilon(\rho^{\otimes n} \| \sigma^{\otimes n}) \geq D(\rho \| \sigma) - \dots$$

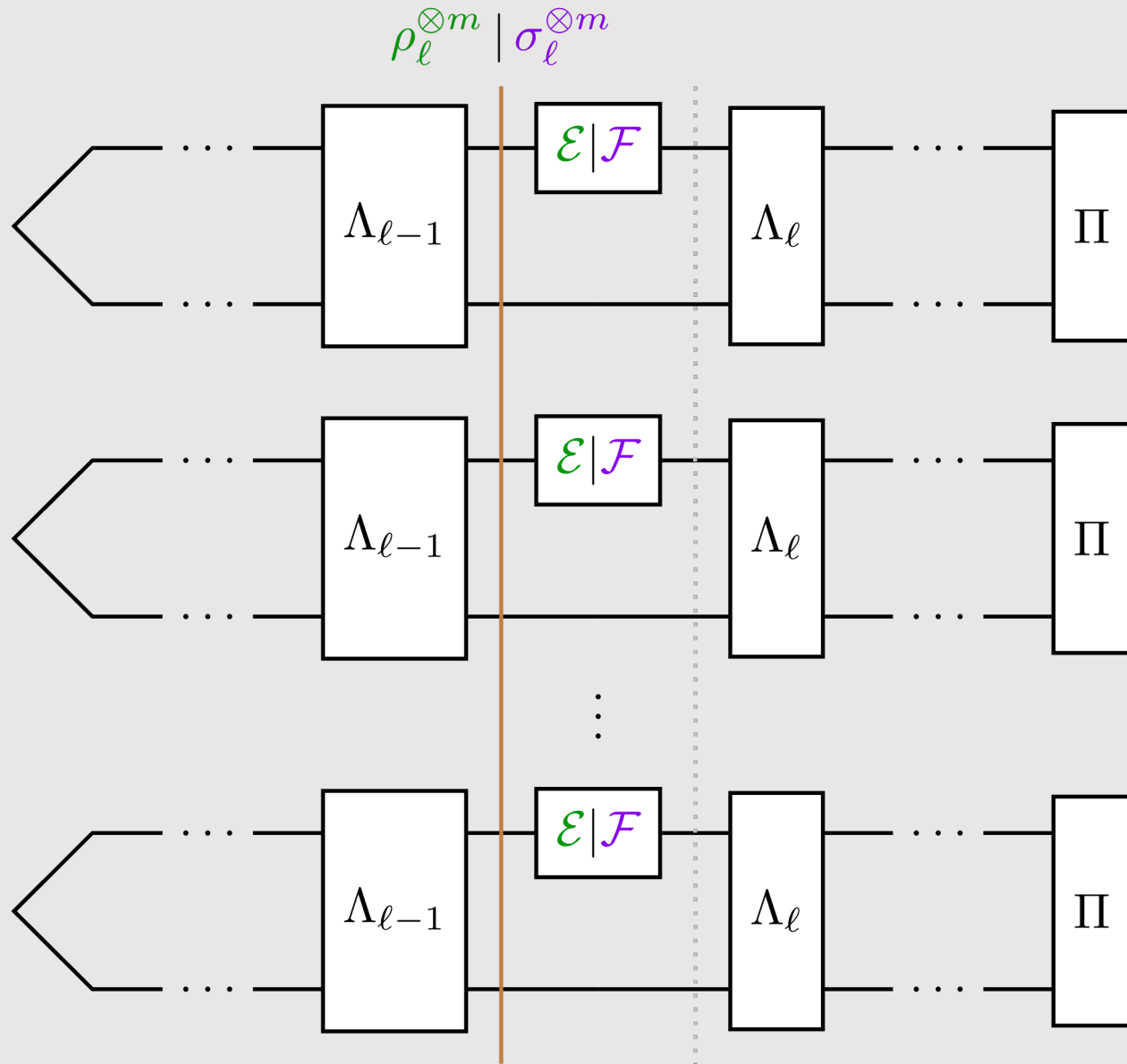
$$\frac{1}{n} D_{\max}^\varepsilon(\rho^{\otimes n} \| \sigma^{\otimes n}) \leq D(\rho \| \sigma) + \dots \rightarrow$$

$$\leq \frac{1}{m} \left[D_{\max}^{\varepsilon''}(\mathcal{E}^{\otimes m}(\rho_\ell^{\otimes m}) \| \mathcal{F}^{\otimes m}(\sigma_n^{\otimes m})) - D_{\max}^{\varepsilon'}(\rho_\ell^{\otimes m} \| \sigma_\ell^{\otimes m}) \right] + \dots$$

$D_{\max}^\varepsilon(\rho \| \sigma) \leq D_H^{1 - \varepsilon^2}(\rho \| \sigma) - \log(1 - \varepsilon^2)$

$$D_{\max}^{\varepsilon + \varepsilon'}(\mathcal{E}(\rho) \| \mathcal{F}(\sigma)) \leq D_{\max}^\varepsilon(\rho \| \sigma) + D_{\max}^{\varepsilon'}(\mathcal{E}(\nu) \| \mathcal{F}(\nu)) \rightarrow$$

$$\leq \frac{1}{m} D_{\max}^\varepsilon(\mathcal{E}^{\otimes m}(\nu) \| \mathcal{F}^{\otimes m}(\nu)) + \dots \leq \frac{1}{m} D_H^{\alpha_P}(\mathcal{E}^{\otimes m}(\nu) \| \mathcal{F}^{\otimes m}(\nu)) + \dots$$



$$\nu \approx^{\alpha_P} \rho_{\ell}^{\otimes m} \quad \text{s.t.} \quad D_{\max}(\nu \| \sigma_{\ell}^{\otimes m}) = \text{minimal}$$

Example

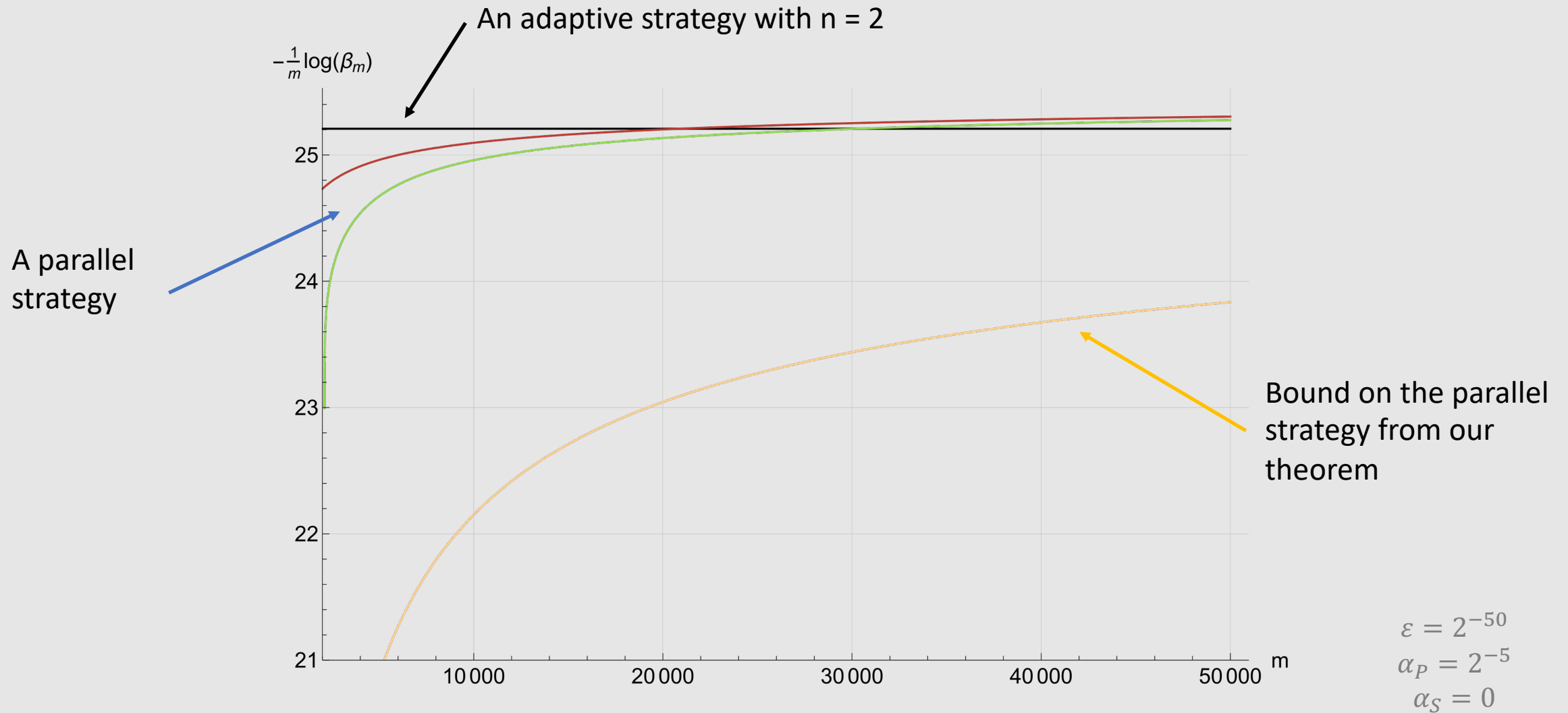
$$\mathcal{E}, \mathcal{F} : \mathcal{B}(\mathbb{C}^2 \otimes \mathbb{C}^2) \rightarrow \mathcal{B}(\mathbb{C}^2)$$

$$\mathcal{E}(\rho \otimes \omega) := |0\rangle\langle 0| \langle 0|\omega|0\rangle + |0\rangle\langle 0| \langle 0|\rho|0\rangle \langle 1|\omega|1\rangle + \frac{\mathbb{1}}{2} \langle 11|\rho \otimes \omega|11\rangle$$

$$\mathcal{F}(\rho \otimes \omega) := (1 - \varepsilon) \left[|+\rangle\langle +| \langle 0|\omega|0\rangle + |1\rangle\langle 1| \langle +|\rho|+\rangle \langle 1|\omega|1\rangle + \frac{\mathbb{1}}{2} \langle -1|\rho \otimes \omega| - 1\rangle \right] \\ + \varepsilon \frac{\mathbb{1}}{2} \text{Tr}[\rho \otimes \omega]$$

\mathcal{E} and \mathcal{F} almost perfectly distinguishable *adaptively* with just two uses

Example



Open questions

- Do we really need $\mathcal{O}(n^2)$ parallel channel uses to achieve the rate of a sequential strategy with n uses?
 - Helpful would be a convergence bound on D_{reg} or D_A
- Second order expansions for channel discrimination tasks
 - Requires controlling $V(\rho_n \parallel \sigma_n) = \mathcal{O}(n)$ or $V(\mathcal{E}^{\otimes n}(\nu) \parallel \mathcal{F}^{\otimes n}(\nu)) = \mathcal{O}(n)$
 - Would prove strong converse