# **Quantum Coupon Collector**

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## Collecting coupons

 Every year Albert Heijn hands out cards of soccer players ("voetbalplaatjes").
 You get a random one for each 2.50€ you spend on groceries



- ▶ There are 18 teams, 11 players per team: 198 different cards
- Your nephew really wants to have a complete set. How much money do you need to spend to get each card?
- ▶ Obviously, at least  $198 \times 2.5 = 495 \in$
- But it's worse: if you already have a copy of most of the cards, then your next 2.5€ spending will likely give you a card that you already have

### Analysis of coupon collector

Suppose you already have i out of k coupons, and you get another, uniformly random coupon.

$$\Pr[\text{see new coupon}] = \frac{k-i}{k}$$

Total expected number of samples:

$$\sum_{i=0}^{k-1} \frac{k}{k-i} = k \sum_{j=1}^{k} \frac{1}{j} = k \ln(k) + \Theta(k)$$

▶ With k = 198, that's 1162 samples. Need to spend 2905€!

You're unlikely to finish much earlier. Variance in number of samples is Θ(k<sup>2</sup>), so typically you need k ln(k) ± O(k) samples

### Can quantum help somehow?

Suppose get "quantum samples" instead of random samples. You want to learn unknown set S ⊆ [n] of size k from states

$$|S
angle = rac{1}{\sqrt{k}}\sum_{i\in S}|i
angle$$

- If you measure, you get a uniformly random sample from S, and we know k ln(k) random samples needed to learn S
- Maybe there's something smarter, using fewer copies of  $|S\rangle$ ?
  - ▶ yes if the number of "missing items" (m = n k) is small
  - no otherwise

Learning S by sampling the missing elements

• 
$$|S\rangle = \sqrt{\frac{k}{n}} |[n]\rangle + \sqrt{\frac{m}{n}} |\psi\rangle$$
  
where  $|\psi\rangle = \sqrt{\frac{m}{n}} |S\rangle - \sqrt{\frac{k}{n}} |\overline{S}\rangle \approx -|\overline{S}\rangle$  if  $m \ll n$ 

- If we measure one copy of  $|S\rangle$  with 2-outcome measurement  $|[n]\rangle\langle[n]|$  vs  $I |[n]\rangle\langle[n]|$ , then with prob  $\frac{m}{n}$  we obtain  $|\psi\rangle$
- ► Thus we can convert an expected number of n/m copies of |S⟩ into one copy of |S⟩ (up to small error)
- Now we can sample uniformly from the *complement* of S!
   *S* has *m* elements, so O(mlog(m + 1)) copies of |*S*⟩ suffice.
   Hence O(nlog(m + 1)) copies of |S⟩ suffice for a "quantum coupon collector". For m = O(1) and k = n O(1), this beats classical coupon collector by a log-factor

## Matching lower bound on number of copies of |S angle

- Claim: you need T = Ω(k log(m + 1)) copies of |S⟩ to learn the k-set S ⊆ [n] (assume m ≤ n/2)
- ► Approach: Use the adversary lower bound (without queries!) A learner should do state transformation:  $|S\rangle^{\otimes T} \mapsto S$ Consider Gram matrix  $M_{SS'} = (\langle S|S'\rangle)^T$ ; and  $F_{SS'} = 1 - \delta_{SS'}$ State transformation problem can be solved iff  $\gamma_2(M \circ F)$  is small
- Can witness  $\gamma_2(M \circ F) \ge 1/2$  via an adversary matrix:

$$\gamma_2(M \circ F) = \max_{\|\Gamma\| \le 1} \|\Gamma \circ M \circ F\|$$

How to construct such  $\Gamma$ ? Note that M, F-entries only depend on  $|S \cap S'|$ , so we need math that respects this symmetry.

### Using the Johnson association scheme

- ▶ Define Boolean matrices  $A_0, ..., A_m$  of dimension  $N = \binom{n}{k}$ , with  $(A_j)_{SS'} = 1$  iff  $|S \cap S'| = k j$
- ► ∃ pairwise-orthogonal projectors  $E_0, ..., E_m$  spanning the same space:  $A_i = \sum_{j=0}^{m} \underbrace{p_i(j)}_{\text{eigenvalues}} E_j = \frac{1}{N} \sum_{i=0}^{m} \underbrace{q_j(i)}_{\text{dual eigenvalues}} A_i,$  $E_i \circ E_j = \frac{1}{N} \sum_{\ell=0}^{m} \underbrace{q_{i,j}(\ell)}_{\text{Krein parameters}} E_\ell.$  These parameters are known.
- M<sub>SS'</sub> entries only depend on |S ∩ S'|, so we can write M as linear combination of A<sub>i</sub>s and hence of E<sub>j</sub>s
- Adversary matrix  $\Gamma = \sum_{j=0}^{m} \gamma_j E_j$ , with  $\gamma_0 = \cdots = \gamma_{m-1} = 1$ ,  $\gamma_m \in [-1, 0]$ . Ensures  $\|\Gamma\| \le 1$ , and diag $(\Gamma) = 0$  (so  $\Gamma \circ F = \Gamma$ )
- Complicated calculation involving Krein parameters (similar to classical coupon!): if T ≪ k log(m + 1) then ||Γ ∘ M ||≥ 1/2

### Relevance for proper vs improper PAC learning

PAC learner A for a concept class C = {f : [n] → {0,1}}: given samples (x, f(x)), x ~ D, for unknown target concept f ∈ C, find hypothesis h : [n] → {0,1} that is close to f:

 $\forall f \in \mathcal{C} \ \forall D : \Pr_{x \sim D}[f(x) \neq h(x)] \leq \varepsilon \text{ w.h.p.}$ 

- ► Fundamental Thm: Required # of samples is  $\Theta(VCdim(C)/\varepsilon)$
- $\mathcal{A}$  is called a proper learner if  $h \in \mathcal{C}$
- Requiring A to be proper can increase sample complexity:
   ∃C where proper learner needs Θ(VCdim(C) log(1/ε)/ε) examples.
   Related to coupon collector with m = 1, ε = 1/n:
   C = {f<sub>S</sub> : [n] → {0,1} is indicator of S}
- ►  $O(VCdim(C)/\varepsilon)$  quantum examples suffice for proper learner

## What if you can also **reflect** through $|S\rangle$ ?

- ▶ If you can get copies of  $|S\rangle$ , then maybe you actually have a quantum machine to produce such copies?  $U : |0\rangle \mapsto |S\rangle$
- ▶ Doing *U* and *U*<sup>-1</sup> would allow you to reflect through  $|S\rangle$ ! *R<sub>S</sub>* :  $|S\rangle \mapsto |S\rangle$ , *R<sub>S</sub>* :  $|\psi\rangle \mapsto -|\psi\rangle$  whenever  $\langle \psi|S\rangle = 0$
- ► Finding *S* more quickly, using amplitude amplification:
  - 1. Start from  $|[n]\rangle$ , rotate to  $|\overline{S}\rangle$ . Measure, get  $i_1 \in \overline{S}$
  - 2. Start from  $|[n]\rangle$ , rotate to  $|\overline{S} \setminus \{i_1\}\rangle$ . Measure, get  $i_2 \in \overline{S}$

Cost of finding all elements of  $\overline{S}$ :  $\sum_{j=0}^{m-1} \sqrt{\frac{n-j}{m-j}} = O(\sqrt{km})$ 

We show that this is tight for m ≤ n/2, and also get tight bound Θ(k) for case m ≥ n/2

## Summary

Classical coupon collector:

learn k-set  $S \subseteq [n]$  from  $\Theta(k \log k)$  uniform samples

Quantum coupon collector:

learn k-set  $S \subseteq [n]$  from  $\Theta(k \log(m+1))$  uniform superpositions (m = n - k is number of missing items)

- $\blacktriangleright$  We also gave tight bounds for learning S from copies of  $|S\rangle$  and reflections through  $|S\rangle$
- Open problem: are the quantum sample complexities of proper and improper learning the same for all C?