Strong Direct Sum for Randomized Query Complexity

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Direct Sum Theorems

Does computing f(x) on k copies scale with k?

Direct Sum Theorem: Computing k copies of f

requires ${\bf k}$ times the resources

Direct Product Theorem: Success prob. of



computing **k** copies of **f** with << k resources is $2^{-\Omega(k)}$

Strong Direct Sum: computing k copies of f(x) requires k*log(k) times the resources

Query Complexity

aka Decision Tree Complexity



Decision Tree for f: $\{0,1\}^n \rightarrow \{0,1\}$:

- · internal nodes labeled w/input bits \mathbf{x}_i
- leaves labeled w/output or ABORT
- cost(T): depth of T == worst-case #queries

Randomized DT:

- distribution A on decision trees
- · cost(A) = max_T cost(T)

Distributional QC D_{\delta,\epsilon}^{\mu}(f): min **cost(T)** s.t. **Pr[abort]** $\leq \delta$ and **Pr[error]** $\leq \epsilon$

Randomized QC $\mathbf{R}_{\delta,\epsilon}(\mathbf{f})$: minimum cost of randomized algorithm s.t.

Pr[abort] $\leq \delta$ and **Pr[error]** $\leq \varepsilon$

(q, δ , ϵ)-algorithm: q queries, abort prob. δ , error prob. ϵ

Query Complexity w/aborts

Minimax Lemma: $D_{2\delta,2\epsilon}^{\mu}(f) \leq R_{\delta,\epsilon}(f) \leq D_{\delta/2,\epsilon/2}^{\mu}(f)$

Error Reduction: $R_{0(1/t, 0(1/t)}(f) \leq O(\log(t)R_{1/2, 1/3}(f))$

Previous Work

[MWY13, MWY15]:

- strong direct sum for information complexity w/aborts + error
- applications for streaming/sketching algorithms

[Drucker12]:

direct product theorems for randomized query complexity

[GPW15, ABBLSS17]:

- query complexity separations based on pointer functions
- polynomial separation R₀(f) vs R_ε(f)

Theorem: Suppose any T-query algorithm computing f has success $\leq 1-\varepsilon$ under μ . Then, any $(\varepsilon Tk)/2$ -query algorithm for computing f^k has success $< (1-\varepsilon/2)^k$ under μ^k [Drucker 12]

Our Results

Strong Direct Sum Theorem: $D_{0,\epsilon}^{\mu^{\kappa}}(f^{k}) = \Omega(kD_{1/5,40\epsilon/k}^{\mu}(f))$

Scaling with ε : There is f: {0,1}^N \rightarrow {0,1} such that for all $\varepsilon > 2^{-\log(N)^2}$, we have $R_{\delta,\varepsilon}(f) = \Theta(N'\log(1/\varepsilon))$

Corollary: There is f such that $R_{\varepsilon}(f^{k}) = \Omega(klog(k)R_{\varepsilon}(f))$

Query-resistant codes: probabilistic encoding G: $\Sigma \rightarrow \{0,1\}^{N}$ such that N/2 bits of G(x) needed to learn anything about x

Query Resistant Codes

Definition: a δN -query resistant code of Σ is a set of distribs {G(x)}

- For each $x \in \Sigma$, G(x) is a distribution on $\{0,1\}^{N}$
- { *support*(G(x)) : x ∈ Σ} partition {0,1}^N
- For all $S \subseteq [N]$ with $|S| \le \delta N$, all $z \in \{0,1\}^{|S|}$ and all $x \ne x'$, distributions G(x), G(x') conditioned on S-bits = z are equal
- "decoding function" h(y) := x iff y e support(G(x))

Lemma: For any Σ , there is a **(N/2)**-query resistant code with $N = |\Sigma|$. Furthermore, conditional distributions $G(x)|_{S=z}$ are uniform.

Query Resistance

For $f: \Sigma^n \rightarrow \{0,1\}$, define $F: \{0,1\}^{nN} \rightarrow \{0,1\}$ as:

 $F(y_1,...,y_n) := f(h(y_1),..., h(y_n))$

Theorem: $\mathbf{R}^{cell}_{\delta,\epsilon}(\mathbf{f}) \leq (2/N)\mathbf{R}_{\delta,\epsilon}(\mathbf{F})$

Proof: Let **A** be a (q, δ , ε)-algorithm for **F**.

```
Algorithm B(x_1,...,x_n) {
emulate A(G(x_1),...,G(x_n))
when A queries G(x_i) for kth time:
if k < N/2, sample G(x_i) cond. on prev. queries
if k = N/2, sample x_i
if k \ge N/2, sample G(x_i) cond. on prev. history.
```

Functions



- BlueRed(y) = 0 if half colored entries Red, half Blue
- theorem: $\mathbf{R}_{\delta+0.1,\epsilon}$ (GapID) \leq (7/m)* $\mathbf{R}_{\delta,\epsilon}$ (BlueRed)

Theorem: $R_{0,\varepsilon}(EncFcn) = O(Nmlog(1/\varepsilon))$

[ABBLSS17]

EncFcn: query resistant code+PtrFcn

theorem: R^{cell}_{δ,ε}(PtrFcn) ≤ (2/N)R_{δ,ε}(EncFcn)

GapID Lower Bound

Theorem: $R_{\delta,\epsilon}(GapID) = \Omega(log(1/\epsilon))$

Hard Distribution $X \sim \mu$:

w/prob α := max(δ, 1.001ε), X=0ⁿ

w/prob **1**- α , **X** uniform on $|\mathbf{x}|=\mathbf{n}/2$

Fix $(log((1-\alpha)/\varepsilon), \delta, \varepsilon)$ -algorithm T for GapID wlog output NO if $X_i = 1$ queried

When all queries = 0:

- abort: **Pr[abort]** > $\alpha \ge \delta$
- output 0: $\Pr[error] = \alpha > \varepsilon$
- output 1: **Pr[error]** \approx (1- α)2^{-q} > ε

In all cases, abort prob. > δ or error prob. > ε .

BlueRed Lower Bound

Theorem: $R_{i+0.1,\epsilon}(GapID) \leq (7/m)^*R_{i,\epsilon}(BlueRed)$

Emulate (**q**,δ,ε)-algorithm **A** for **BlueRed**

- each colored entry in uniform row
- pick each i_j e [m] uniformly
- map $0 \rightarrow \text{Red}, 1 \rightarrow \text{Blue}$
- abort if A queries > 7q/m colored entries

Claim: $Pr[> 7q/m colored entries probed] \le 1/10.$

BlueRed Lower Bound

Claim: $Pr[> 7q/m \ colored \ entries \ probed] \le 1/10.$

- For any column, Pr[colored entry found on k-th query] = 1/m
- For any leaf w/t colored entries found, Pr[leaf] ≤ m^{-t}
- there are {q choose t} leaves w/t colored entries found

$$\begin{split} \text{Pr[> 7q/m colored entries]} &\leq \Sigma_{t > 7q/m} \{q \text{ choose }t\}m^{-t} \\ &\leq \Sigma_{t > 7q/m} (qe/mt)^t \\ &< \Sigma_{t > 7q/m} (e/7)^t \\ &< 1/10. \end{split}$$

PtrFcn Lower Bound

Theorem: R_{δ,2}ε(**BlueRed**) ≤ **R**^{cell}_{δ,ε}(**PtrFcn**)

Partially emulate (q, δ, ε) -algorithm **A** for **PtrFcn**:

- map **Black** \rightarrow [1, \perp , \perp ..., \perp]
- map Red \rightarrow [0, \perp , \perp ..., \perp]
- Blue: halt, output NO

Claim: Let **x & BlueRed**⁻¹(0). Then **Pr[no Blue entries queried] < 2***ɛ* **Proof:**

- Let z e PtrFcn⁻¹(1) be consistent with x.
- z' := z, w/value of special entry = 0.
- A(z)=A(z') unless special entry queried.
- **Pr[no blue entries queried]** \leq **Pr[special entry not queried]** \leq 2 ϵ

Strong Direct Sum Theorem: $D_{0,\epsilon}^{\mu^{k}}(f^{k}) = (kD_{1/5,40\epsilon/k}^{\mu}(f))$

Let **A** be an ε -error algorithm for **f**^k.

```
Let y = (y_1, ..., y_k).
```

Embed(y,i,x) := y, w/i-th coord replaced by x.

```
Algorithm B(x) {
   carefully select y,i
   emulate A(EMBED(y,i,x))
   abort if problems found
}
```

Strong Direct Sum Theorem: $D_{0,\epsilon}^{\mu^{k}}(f^{k}) = (kD_{1/5,40\epsilon/k}^{\mu}(f))$

k

$$1-\varepsilon \leq \Pr_{Y \sim \mu^{k}}[A(Y) = f^{k}(Y)] = \prod_{i=1}^{n} \Pr_{Y \sim \mu^{k}}[A(Y)_{i} = f^{k}(Y)_{i} \mid A(Y)_{$$

• at least 2k/3 i give $Pr[A(Y)_i = f^k(Y)_i | A(Y)_{<i} = f^k(Y)_{<i}] \le 10 \varepsilon/k$ (1)

(2)

- $q_i(Y)$: # queries of Y_i
- $q \ge \Sigma_i E_Y[q_i(Y)] \implies \ge 2k/3 i \text{ have } E[q_i(Y)] \le 3q/k$
- Fix **i*** to get **(1)** and **(2)**. **Y*** := **Embed(Y,i*,x)**. This **i*** satisfies:
- 1. $E_{Y \sim \mu k}$ [$Pr_{x \sim \mu}$ [$A(Y^*)_{<i} \neq f^k(Y^*)_{<i}$]] ≤ ε
- 2. $E_{Y}[Pr_{x \sim \mu}[A(Y^{*})_{i} \neq f^{k}(Y^{*})_{i} | A(Y)_{<i} = f^{k}(Y)_{<i}] \leq 10 \epsilon/k$
- 3. $E_{Y}[E_{X}[q_{i}(Y^{*})]] \le 3q/k$

Strong Direct Sum Theorem: $D_{0,\epsilon}^{\mu^{\kappa}}(f^{k}) = \Omega(kD_{1/5,40\epsilon/k}^{\mu}(f))$

This **i*** satisfies:

- **1.** $E_{Y \sim \mu k}[Pr_{x \sim \mu}[A(Y^*)_{<i} \neq f^k(Y^*)_{<i}]] \leq \varepsilon$
- 2. $E_{Y}[Pr_{x \sim \mu}[A(Y^{*})_{i} \neq f^{k}(Y^{*})_{i} | A(Y)_{<i} = f^{k}(Y)_{<i}] \leq 10 \epsilon/k$
- 3. $E_{Y}[E_{X}[q_{i}(Y^{*})]] \le 3q/k$

Markov Inequality: there is y* such that

- 1. $Pr_{x\sim\mu}[A(Y^*)_{<i} \neq f^k(Y^*)_{<i}] \leq 4\varepsilon$
- 2. $Pr_{x \sim \mu}[A(Y^*)_i \neq f^k(Y^*)_i | A(Y)_{<i} = f^k(Y)_{<i}] \le 40 \epsilon/k$
- 3. $E_X [q_i(Y^*)] \le 12q/k$

```
Algorithm B(x) {
  z := EMBED(y*,i*,x)
  emulate A(z)
  abort if qi*(z) > 120q/k
  abort if A(z)<i* ≠ f<sup>k</sup>(z)<i*</pre>
```

abort probability: $1/10 + 4\varepsilon < 1/5$

error probability: 40*c*/k

Open Problems

- 1. Give a more efficient query resistant code
- 2. *Characterize* functions robust to **aborts**
- 3. Strong Direct Sum for Composed Functions
- 4. How does $R_{\delta,\epsilon}(f)$ compare to other QC measures?

Thanks!

slide of common stuff



Fact: If $f \in ACC^{0}$ then f has NOF protocol with poly(log n) communication and k = poly(log n) players

R_{6,ε}(GapID)