

A Meta-Complexity Characterisation of Quantum Cryptography

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Joint work with
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The holy grail of cryptography?

Minimal complexity-theoretic assumption required for cryptography?

- $P \neq NP$: necessary; sufficient?
 - ▶ Specific constructions: Factoring, LWE, etc.
- Quantum complication: PRSs may exist even if $BQP = QMA$ (in particular, \nexists pqOWFs) [Kretschmer, TQC 2021]
 - ▶ The classical “holy grail” is no longer relevant
- Meta-complexity: classical characterisations!

This talk: an equivalence between **quantum** cryptography and complexity theory *via meta-complexity*.

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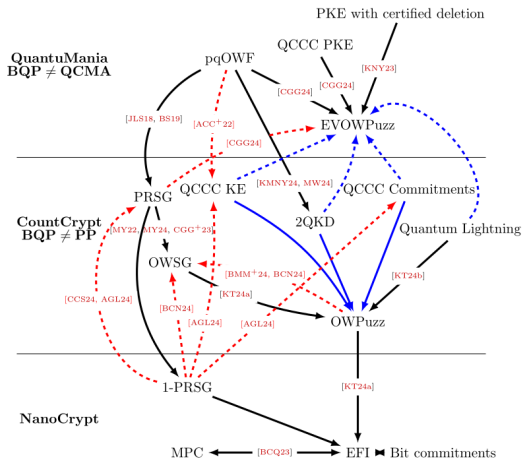
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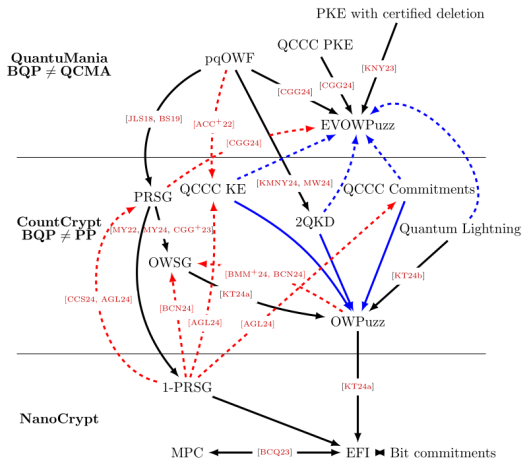
Minimal quantum cryptographic primitive?



Credits: Goldin, Morimae, Mutreja, and Yamakawa [2024]

- Primitives following from PRS are not known to be equivalent (in fact, plenty of oracle separations)
- OWPuzz's seem minimal in QCCC and CountCrypt
- EFI seems minimal for quantum communication
- This work: OWPuzz

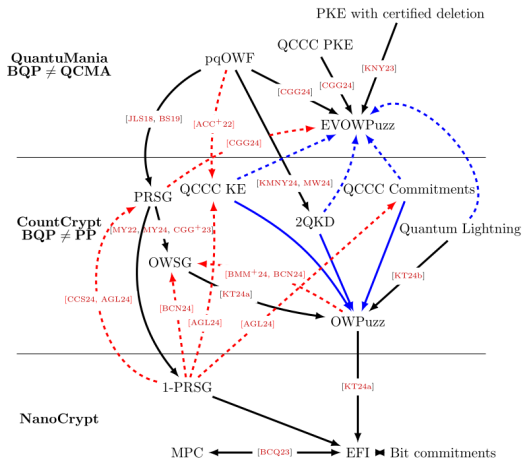
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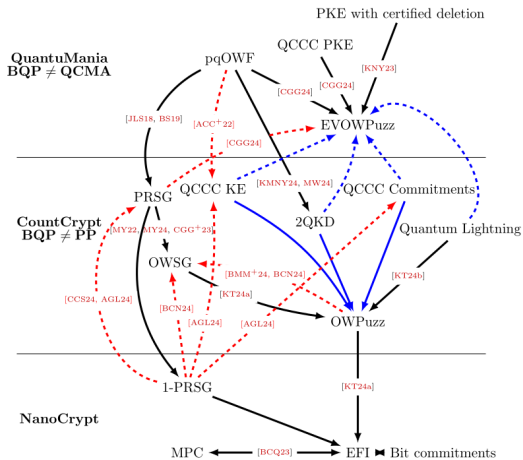
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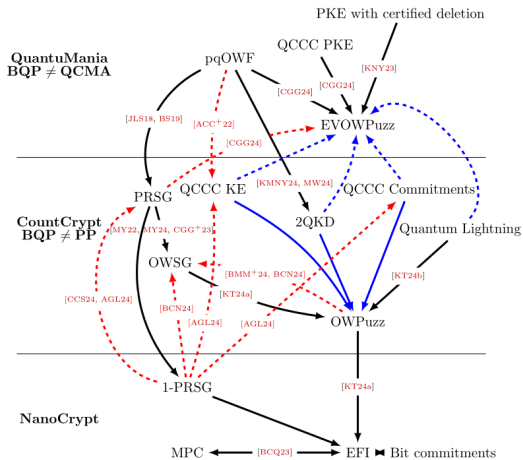
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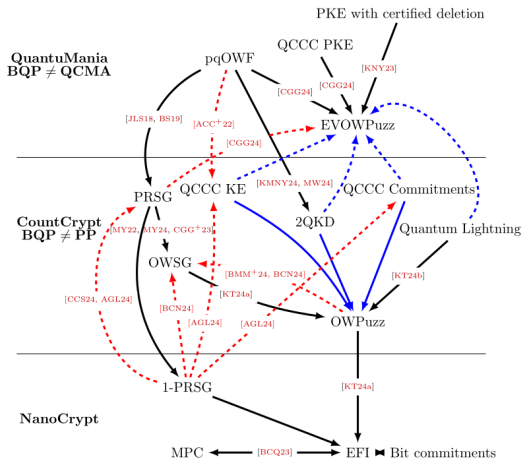
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One-way puzzles

- $\text{OWPuzz} = (\text{Samp}, \text{Ver})$. [Khurana-Tomer, STOC 2024]
- $\text{Samp}(1^\lambda) \rightarrow (k, s) \in \{0, 1\}^*$ in quantum polynomial-time (QPT).
 - ▶ k : key; s : puzzle.
- *Soundness*: $\mathbb{P}_{(k,s) \leftarrow \text{Samp}(1^\lambda)}[\text{Ver}(k, s) = \top] = 1 - \text{negl}(n)$
- *Correctness*: $\forall \text{QPT } \mathcal{A}, \mathbb{P}_{(k,s) \leftarrow \text{Samp}(1^\lambda)}[\text{Ver}(\mathcal{A}(s), s) = \top] = \text{negl}(n)$
- Notably: Ver can be inefficient!
- Arises naturally from *shadow tomography*.

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Kolmogorov Complexity

How much information is in a text?

- Example 1: "1010101010101010" (Print "10" 8 times)
- Example 2: "0001010100001011" (???)

Kolmogorov complexity $K(x)$ of a string $x \in \{0, 1\}^*$:
minimum length of a program that outputs x .

PROPERTIES

1. $K(x) \leq |x|$.
2. Random strings have near maximum Kolmogorov complexity.
3. Computing $K(x)$ is impossible!

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Approximating Kolmogorov Complexity (GapK)

Let $\text{GapK}[s, s + \Delta]$ be the “promise” problem of distinguishing between strings of K.c. at most s and those with K.c. at least $s + \Delta$.

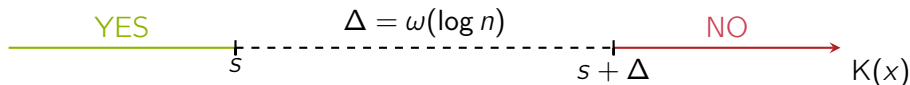


IRS'20: GapK can be solved in exponential-time *on average* (PP = PostBQP).

For every PPT-samplable distribution \mathcal{D} and $\Delta = \omega(\log n)$, it's possible to compute $\text{GapK}[s, s + \Delta]$ on \mathcal{D} with error at most $n^{-O(1)}$ in exponential time.
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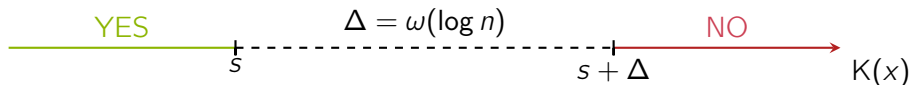


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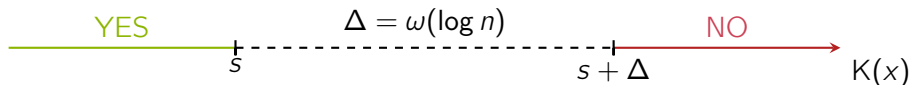


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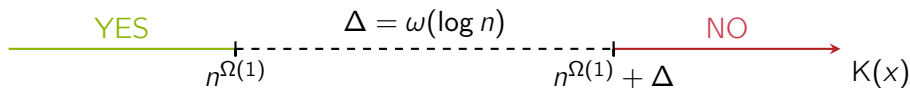
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Cryptography vs. Kolmogorov complexity

Theorem (Ilango-Ren-Santhanam [IRS], STOC 2020)

The following are equivalent:

- *One-way functions exist.*
- *For some $s = n^{\Omega(1)}$ and $\Delta = \omega(\log n)$, there exists a samplable distribution \mathcal{D} such that $\text{GapK}[s, s + \Delta]$ is average-case hard on \mathcal{D} .*



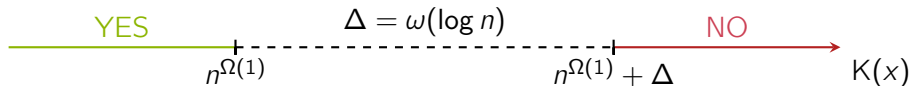
The OWF encodes a hard probability distribution, and vice-versa.

Quantum Cryptography vs. Kolmogorov complexity

Theorem (C.-Goldin-Gray-Hall [CGGH], EUROCRYPT 2025)

The following are equivalent:

- One-way **puzzles** exist.
- For some $s = n^{\Omega(1)}$ and $\Delta = \omega(\log n)$, there exists a **quantum** samplable distribution \mathcal{D} such that $\text{GapK}[s, s + \Delta]$ is average-case hard on \mathcal{D} **for quantum algorithms**.



The OWPuzz encodes a hard probability distribution, and vice-versa.

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- Further evidence of OWPuzz's centrality in QCCC!
 - ▶ Besides combiners, hardness amplification, etc.
 - ▶ Natural generalisation of OWFs
 - ▶ Embodies some fundamental hardness via meta-complexity

Probability estimation: crucial intermediate step

Let \mathcal{D} be a sampler. Let $p_x := \mathbb{P}[x \text{ is sampled by } \mathcal{D}]$.

δ -Prob. estimation: given $x \leftarrow \mathcal{D}$, output $\mathcal{A}(x) \in [\delta \cdot p_x, p_x]$ w.h.p..

Breaking OWPuzz \implies Prob. estimation

Classically: use the random bits of the sampler to construct a OWF candidate:

$$f_k(h, r) = \mathcal{D}(r), h, h(r),$$

where h is a hash function mapping to k bits.

Estimate p_x by sampling h, z and inverting (x, h, z) (many times).
(Intuition: select a random hash bucket and see if r “is there”)

Quantumly: randomness is inherent to the hard distribution!

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On Central Primitives for Quantum Cryptography with Classical Communication

Kai-Min Chung¹, Eli Goldin², and Matthew Gray³

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²New York University (eli.goldin@nyu.edu)

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- **Takeaway:** \nexists OWPuzz \implies sample from (k, s) *conditioned* on s .

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Amplitude estimation with a distributional inverter

One-way puzzle: sample $x \leftarrow \mathcal{D}$, h hash function mapping to k bits.

Key: x , Puzzle: $(h, h(x))$.

Distributionally invert [CGG]: as k increases, x is isolated by $(h, h(x))$.
(Intuition: fix the “right” random hash bucket, and see if anything else besides x is there.)

The threshold when x starts to become isolated is our estimate for p_x .

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Computing GapK with probability estimation

Just like the classical case [IRS].

Requires a “quantum coding theorem”:

$$K(x) \lesssim \log \left(\frac{1}{\mathbb{P}[\mathcal{D} \rightarrow x]} \right) + |\mathcal{D}|.$$

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(In other words, low probability \approx high Kolmogorov complexity.)

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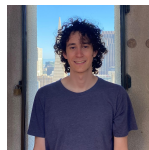
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CGG: OWPuzz \implies “non-uniform entropy-gap generator”.



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1. Khurana-Tomer (STOC 2025): Shows equivalence between probability estimation and one-way puzzles (with tighter approximation factor);
2. Hiroka-Morimae (2024): Shows how to construct one-way puzzles from average-case hardness of GapK (with a different proof).

Open questions

1. Characterising with a problem with worst-case complexity bound? (Classical: AM).
2. Characterise other primitives? E.g., EFIs, OWSGs.
3. Strong average-case hardness?
4. Oracle separation does not rule out: construct OWPuzzles from the hardness of a (meta-complexity?) problem in QMA.

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Thanks!