# Fundamental Finite Key Limits for Information Reconciliation in Quantum Key Distribution 

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## Outline

1 Quantum Key Distribution

2 Information Reconciliation

3 Motivation

4 Fundamental Limits for Information Reconciliation - Theoretical Results

■ Simulation Results

5 Conclusions / Open Questions

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## Quantum Key Distribution（QKD）

－Cryptographic primitive for key agreement
－Two honest parties：Alice and Bob；dishonest party（eavesdropper）：Eve．
－Achievement：Alice and Bob create an information－theoretic secure （composable）key．

## Quantum Key Distribution (QKD)

- Cryptographic primitive for key agreement
- Two honest parties: Alice and Bob; dishonest party (eavesdropper): Eve.
- Achievement: Alice and Bob create an information-theoretic secure (composable) key.

Information-theoretic security (informally)
The success probability of any (active or passive) attack is upper bounded by a (tiny) constant, regardless of the (quantum) computing resources used by the attacker.

## QKD protocol steps



## Prerequisites:

- Authentic classical channel (Eve can listen)
- Quantum channel (Eve introduces noise while listening)


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- One Way IR = Source Coding with Side Information
- Asymptotic limit it is sufficient to send $n H(X \mid Y)$ bits


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- However, this choice should depend on the distribution $P_{X Y}$, the frame length $n$, and the frame error rate $\varepsilon$.
- What are the fundamental / practical limits of $\log |\mathcal{M}|$ as a function of $P_{X Y}, n$, and $\varepsilon$ ?


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IR / Source coding with side information


Bounds on the asymptotic expansion up to second order (Hayashi 2008 and Tan and Kosut 2012)

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1 For an arbitrary $\left(P_{X Y}\right)^{\times n}$ we provide the asymptotic expansion up to third order for the converse bound
$\boxed{2}$ For a special case we provide a non-asymptotic converse bound
3 We compare these bounds to implementations of one-way IR using low-density parity-check codes.

## Fundamental Limits For Information Reconciliation

Definition
An IR protocol is $\varepsilon$-correct on $P_{X Y}$ if

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Theorem (Converse bound (Normal approximation))
Let $0<\varepsilon<1$. Then, for large $n$, any $\varepsilon$-correct IR protocol on $P_{X Y}$ satisfies

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where $H(X \mid Y):=\operatorname{Exp}\left[\log \frac{P_{Y}}{P_{X Y}}\right]$ is the conditional entropy,
$V(X \mid Y):=\operatorname{Var}\left[\log \frac{P_{Y}}{P_{X Y}}\right]$ is the conditional entropy variance, and $\Phi$ is the cumulative standard normal distribution.

## Special Case: Quantum Bit Error Rate Q

$P_{X Y}^{Q}$ results from measurements on a channel with (independent) qber $Q$ :

$$
\begin{aligned}
P_{X}^{Q}(0) & =P_{X}^{Q}(1)=P_{Y}^{Q}(0)=P_{Y}^{Q}(1)=1 / 2, \\
P_{X Y}^{Q}(0,0) & =P_{X Y}^{Q}(1,1)=(1-Q) / 2, \\
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An IR protocol is $(\varepsilon, Q)$-correct if it is $\varepsilon$-correct on $P_{X Y}^{Q}$.
Theorem (Non-asymptotic converse bound for ( $\varepsilon, Q$ )-correct prot.)

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\begin{aligned}
\log |\mathcal{M}| & \geq n h(Q)+\left(n(1-Q)-F^{-1}(\varepsilon(1+1 / \sqrt{n}) ; n, 1-Q)-1\right) \log \frac{1-Q}{Q} \\
& -\frac{1}{2} \log n-\log \frac{1}{\varepsilon} .
\end{aligned}
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where $F^{-1}(\cdot ; n, p)$ is the inverse of the CDF of the binomial distribution.

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Corollary (Converse bound for ( $\varepsilon, Q$ )-correct protocol)
Let $0<\varepsilon<1$ and let $0<Q<\frac{1}{2}$. Then, for large $n$, any $(\varepsilon, Q)$-correct IR protocol satisfies

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\log |\mathcal{M}| \geq \xi(n, \varepsilon ; Q) \cdot n h(Q)-\frac{1}{2} \log n-O(1), \quad \text { where }
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Numerically, this simple bound matches the non-asymptotic bound very well.

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$\xi$ as a function of the frame error rate $\varepsilon$


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## But what about realistic IR codes?

Theoretical Bound

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\frac{\log |\mathcal{M}|}{n h(Q)} \approx \xi(n, \varepsilon ; Q):=1+\frac{1}{\sqrt{n}} \frac{\sqrt{v(Q)}}{h(Q)} \Phi^{-1}(1-\varepsilon)
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Conjecture for LDPC codes

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\frac{\log |\mathcal{M}|}{n h(Q)}=: \hat{\xi}(n, \varepsilon ; Q) \approx \xi_{1}+\xi_{2} \cdot \frac{1}{\sqrt{n}} \frac{\sqrt{v(Q)}}{h(Q)} \Phi^{-1}(1-\varepsilon)
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Simulations of LDPC codes and fits


| $n$ | $\log \|\mathcal{M}\|$ | $\xi_{1}$ | $\xi_{2}$ |
| :---: | :---: | :---: | :---: |
| $10^{3}$ | $4 \cdot 10^{2}$ | 1.11 | 1.39 |
| $10^{3}$ | $3 \cdot 10^{2}$ | 1.12 | 1.45 |
| $10^{3}$ | $2 \cdot 10^{2}$ | 1.13 | 1.69 |
| $10^{4}$ | $4 \cdot 10^{3}$ | 1.07 | 1.41 |
| $10^{4}$ | $3 \cdot 10^{3}$ | 1.08 | 1.44 |
| $10^{4}$ | $2 \cdot 10^{3}$ | 1.11 | 1.89 |

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| $n$ | $Q$ | $\xi_{1}$ | $\xi_{2}$ |
| :---: | :---: | :---: | :---: |
| $10^{3}$ | 0.015 | 1.16 | 1.52 |
| $10^{3}$ | 0.030 | 1.16 | 1.31 |


| $n$ | $Q$ | $\xi_{1}$ | $\xi_{2}$ |
| :---: | :---: | :---: | :---: |
| $10^{4}$ | 0.025 | 1.14 | 1.26 |
| $10^{4}$ | 0.040 | 1.07 | 1.58 |

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- Commonly used approximation $\log |\mathcal{M}| \approx 1.1 n h(Q)$ is often too optimistic for one-way IR
- Numerical simulations for LDPC codes $\rightarrow$ approximation that can be used for the design of QKD systems


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- Behaviour for different code families
- Joint consideration of fundamental limits for finite-length reconciliation and privacy amplification


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## THANK YOU!


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