#### Non-Binary LDPC Codes for Quantum Key Distribution

Lara Dolecek

Based on joint work with: D. Mitra, L. Tauz, M. Sarihan, J. Shreekumar, Prof. C. W. Wong, and †Prof. E. Soljanin ECE Department, UCLA, †EE Department Rutgers University

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#### Outline

#### 1. Background and Motivation

2. System Model

#### 3. NB-LDPC codes for Quantum Key Distribution

- Preliminaries
- Channel code design
- Sequential Decoding and Interactive Communication
- Privacy Amplification-Aware LDPC Code Design

#### 4. Concluding Remarks and Future Outlook

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  - We focus on Discrete Variable (DV)-QKD





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Error Correction Coding is applied to the information reconciliation phase.



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- During each iteration, the raw key is shuffled and divided into top-level blocks
  - Size of the top-level block depends on the iteration number and the estimated quantum bit-error rate

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Can LDPC codes cognizant of the properties of the QKD channel improve key rates ?

• The time domain is divided into frames, each frame contains  $2^q$  bins: each arrival-time is represented by a q-bit symbol

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- Only the frames where both Alice and Bob have exactly one detection are kept
- Complex error sources: timing jitter errors, detector downtime, dark counts, photon loss



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• In the leftmost frame, both Alice and Bob map their result to '00' (the leftmost bin is occupied in both). Bit extraction is successful despite a slight arrival jitter.



• In the second frame, Alice detects two arrivals, with the spurious one due to dark counts. While Bob initially can map his result to '01' for his second frame, upon receiving information from Alice that this frame is invalid, he too discards it.



• In the third frame, both Alice and Bob observe a single arrival. However, due to timing jitter, the two arrivals fall into adjacent bins. Alice maps her result to '01' whereas Bob maps his to '10', resulting in a 2-bit discrepancy.



• In the rightmost frame, Alice and Bob each again detect a single arrival. In this example, two arrivals are due to dark counts and are uncorrelated (unbeknownst to Alice and Bob at this point). Alice maps her result to '01' and Bob his to '11'.

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Parity bits are visible to Eve.

Need to balance code rate and its error correction capability.



### Prior related work

Classical design approaches

- LDPC coding (Kasai '10, Mao '19, Milicevic '17, 18, Muller '23), Polar coding (Jouguet '14, Fang '22)
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- Multilevel binary LDPC codes (Zhou '13)
  - Implemented and tested in experimental systems (Zhong '15, Chang '23, Chang '24)
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Next: Information Reconciliation using LDPC codes designed based on the properties of the QKD channel

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  - Global errors: Uniform distribution - due to channel losses, and accidental
    - concurrent detections of stray photons
- The SNR in practice is very low resulting in high error rates


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- IR rate depends on both the coding rate  $\left(\frac{N-M}{N}\right)$  and FER of the code
- Maximum in the IR rate occurs for a relatively large value of FER ( $\sim 1-10\%)$

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Coding for QKD

Binary LDPC codes:

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- 2. Interleaved decoding among layers to improve IR rates
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### Irregular LDPC codes

- Improves the FER performance of the code
- $\bullet\,$  Can be characterized using their VN and CN degree distributions L(x) and P(x), respectively

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Our work: we optimize degree distribution L(x), P(x) for the QKD channel to result in large IR rates

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### Generally

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- L(x) is optimized for a fixed code rate R







$$\begin{array}{c} \label{eq:relation} \mathsf{IR} \ \mathsf{rate} = \sum_{i=1}^{T} (1 - FER_i) \ast Rate_i \ast \alpha_i \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & \\ & & \\ &$$

IR rate = 
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$$\mathbf{X}_i \longrightarrow \mathbf{Y}_i \mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_{i-1}$$



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- FER can be obtained with low complexity by Monte-Carlo simulation of a small number of codewords

L. Dolecek (UCLA)

Coding for QKD



- FER is the frame error rate encountered on channel  $\gamma_i$
- $\bullet\,$  Can jointly optimize the degree distribution L(x) and rate R to get high IR rates









Error propagation and current remedy:

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Sequential Decoding and Interactive Communication



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Decoding and interactive communication is sequential

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• Maximum number of decoding iterations:  $\Delta$ ,

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• Set-up: Wong Group @ UCLA. 2000 postprocessed frames.



q	best $a$
4	3
5	3
6	4
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- Small  $a \implies$  more layers  $\implies$  larger IR rate due to sum IR rate formula









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- Latency is non-monotonic in *a*: since as *a* increases, the increase in decoding complexity is offset by the decrease in the number of layers
- Best trade-off between IR rate and latency is obtained for a small value of a (3 or 4)



• 32 bins per frame  $\rightarrow q = 5$ .



• Improvement due to  $a = 1 \rightarrow a = 3$ 



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- $\bullet~$  Improvement due to  $a=1 \rightarrow a=3$
- Improvement due to using JRDO-LDPC codes
- Overall, JRDO-LDPC codes combined with NB-MLC with a small a result in the largest IR rates (effectively uses one code over GF(8) and one code over GF(4))

#### Performance Comparison: Benefits of Interleaved Decoding



- IDC: Interleaved Decoding and Communication
- SDC: Sequential Decoding and Communication

#### Performance Comparison: Benefits of Interleaved Decoding



• IDC protocol results in higher IR rates compared to the conventional SDC protocol

#### The Overall Comparison


## The Overall Comparison



- Overall, our techniques result in around 40 60% improvement in IR rates compared to the MLC scheme (prior state of the art).
- Asymptotic limit of a simplified (better) channel is around 2.1 for q = 5 and 1.8 for q = 4.

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• Assume that  $\mathbf{H}$  can be partitioned into two sub-matrices  $\mathbf{H} = [\mathbf{H}_1, \mathbf{H}_2]$  such that  $\mathbf{H}_2$  is  $m \times m$  and is full rank.



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- There may be multiple subsets that satisfy this property.

L. Dolecek (UCLA)

Coding for QKD

• MDS codes:

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- Our Solution Block-MDS (BMDS) QC-LDPC codes:
  - Structured code construction that guarantees the required subsets on a block matrix level while still having the benefits of LDPC codes

$$\mathbf{B} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \quad \mathbf{H} = \begin{bmatrix} s_{1,1}\mathbf{C}^{p_{1,1}} & s_{1,2}\mathbf{C}^{p_{1,2}} & s_{1,3}\mathbf{C}^{p_{1,3}} \\ s_{2,1}\mathbf{C}^{p_{2,1}} & s_{2,2}\mathbf{C}^{p_{2,2}} & s_{2,3}\mathbf{C}^{p_{2,3}} \end{bmatrix}$$

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#### Main Result

A binary regular QC-LDPC code with column weight  $\gamma$ , row weight  $\kappa$ , and girth  $\gamma + 2$  can be transformed into a non-binary BMDS code with finite field size at least  $\kappa$ , for a particular prime choice of the lifting factor.

# Full Codeword Decoding vs Subset Decoding Simulations





- All codes have girth 10, length of approximately 2000, and perform operations in GF(8)
- Channel is a q-ary symmetric channel with transition probability *p*

## Outline

#### 1. Background and Motivation

2. System Model

### 3. NB-LDPC codes for Quantum Key Distribution

- Preliminaries
- Channel code design
- Sequential Decoding and Interactive Communication
- Privacy Amplification-Aware LDPC Code Design

### 4. Concluding Remarks and Future Outlook

#### Quantum Communications



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  - Code designs that operate in a low SNR regime; refined channel model
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- Key Result: Novel non-binary codes (Mitra '24, Tauz '24)
  - Research supported by NSF-QuIC-TAQS no. 2137984 and NSF-FET no. 2008728.

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  - Relevant attack models and security proofs for them.


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Each of these technologies will benefit from new studies in channel coding and related disciplines.

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Coding for QKD

• The verification protocol is based on the following hash function [Federov '18]:

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 $F = \{0, 1, 2, 3, 4, 5, 6\}$ 

k = 3

$X = 10 \ 10$	01	11	00	10	11	p = 7
<i>N</i> = 10 10	01		00	10		$\lfloor \log_2 p \rfloor = 2$

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Example:

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- k is a random element of  $\mathbb F$

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Example:

$$\begin{array}{c|c} x = 10 & 10 & 01 & 11 & 00 & 10 & 11 \\ \hline X = 10 & 10 & 01 & 11 & 00 & 10 & 11 \\ \hline y = 7 \\ [log_2 p] = 2 \\ F = \{0, 1, 2, 3, 4, 5, 6\} \\ k = 3 \end{array}$$

• Partition X into substrings of length  $\lfloor \log_2 p \rfloor$ 

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Example:

 $2k^0+2k^1+k^2+3k^3+0k^4+2k^5+3k^6 \bmod p=1$ 

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Example:

• Verification is achieved by matching the hash of the string X at Alice and decoded (reconciled) string  $\widehat{X}$  at Bob

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