

# An Accurate Analysis of the BINARY Information Reconciliation Protocol by Generating Functions



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# Information Reconciliation (IR)



- Errors in raw key shared by Alice and Bob in quantum key exchange because of:
  - Eve's eavesdropping
  - Quantum channel noise
- Need information reconciliation to correct shared key by public exchange of parity bits.
- Want to:
  - Minimize leakage of parity bits
  - Maximize Decoding Success Probability (probability of having no errors left after decoding)

# BINARY and CASCADE IR Protocol



- **BINARY IR Protocol:**
  1. At pass  $i$ , divide shared secret into blocks of length  $k_i$
  2. Binary search error correction performed on each odd-parity block. Correct one error by exchanging  $\log_2(k_i)$  parity bits.
  3. Not all errors corrected, so partially corrected secret is permuted and chopped up into blocks of size  $k_{i+1} = 2k_i$ .
  4. Repeat steps 2 to 3 for a number of passes, e.g. four or five passes.
  
- **CASCADE IR Protocol:**
  - In step 2, when an error bit is identified in a current pass, backtrack to correct more error bits which are paired with this bit in previous passes.

# Objective



- To find the exact error probability distribution for the BINARY IR protocol at each pass.
- This will help us determine:
  - Leakage of parity bits
  - Decoding Success Probabilityfor the BINARY protocol.
- Will help us in analysis of CASCADE, one of the popular IR protocols in use today.

# Notations



Symbol	Meaning
$n$	Length of secret
$k_i$	Size of blocks at pass $i$
$p$	Bit error rate (BER)
$\Delta_i(j-y j)$	Probability that at $i^{\text{th}}$ pass, $j-y$ errors are corrected conditioned on there being $j$ errors
$P_i(y)$	Probability that there are $y$ errors at the $i^{\text{th}}$ pass

# Probability Distribution of BINARY



- **Theorem 1:** Let the initial probability distribution before BINARY IR is given by:

$$P_0(y) = \binom{n}{y} p^y (1-p)^{n-y}$$

The probability that there are  $y$  errors left in the raw key after pass  $i$  of the BINARY IR protocol is:

$$P_i(y) = \sum_{j=y}^{y+n/k_i} P_{i-1}(j) \Delta_i(j-y | y)$$

**Proof Idea:** Supposed there are  $j$  errors before executing pass  $i$ , we need to correct  $j-y$  errors to be left with  $y$  errors after pass  $i$ . Then we sum over all possible number of errors  $j$  before pass  $i$ .

# Probability Distribution of BINARY



- **Theorem 2:** The quantity  $\Delta_i(j-y|j)$  needed in Theorem 1 is computed by: 
$$\frac{\binom{n/k_i}{j-y}}{\binom{n}{j}} \times C_{i,j,y}$$

where  $C_{i,j,y}$  is the coefficient of  $x^j$  in:

$$\left( \frac{(1+x)^{k_i} - (1-x)^{k_i}}{2} \right)^{j-y} \left( \frac{(1+x)^{k_i} + (1-x)^{k_i}}{2} \right)^{n-(j-y)}$$

**Proof Idea:** To correct  $j-y$  out of  $j$  errors, we need to distribute  $j$  error bits among  $n/k_i$  blocks such that  $j-y$  of them has odd parity.

# Example



An example computation of Theorem 1 and comparison with simulation for  $n=2048$ -bit,  $\text{BER} = 3\%$ .

	Pass $i = 1$	Pass $i = 2$	Pass $i = 3$	Pass $i = 4$
$P_i(0)$	0.00002	0.08808	0.63320	0.92559
$P_i(2)$	0.00020	0.17874	0.23883	0.06221
$P_i(4)$	0.00114	0.21261	0.08548	0.00974
$P_i(6)$	0.00421	0.19004	0.02891	0.00192
$P_i(8)$	0.01166	0.14029	0.00937	0.00042
$P_i(0)$	0.00000	0.08700	0.63740	0.92750
$P_i(2)$	0.00010	0.17690	0.23460	0.05950
$P_i(4)$	0.00110	0.21610	0.08540	0.01030
$P_i(6)$	0.00360	0.19370	0.02840	0.00230
$P_i(8)$	0.01070	0.13500	0.01020	0.00020

TABLE II  
BINARY SIMULATION (TOP) AND CALCULATION (BOTTOM):  $k_1 = 16$ ,  
 $n = 2048$ ,  $p = 0.03$



# Brassard-Savail's Bound



- Brassard and Savail derived a bound to track the number of errors remaining in (an initial) block after pass  $i$ .
- They use  $\delta_i(j)$  to denote the probability that there are  $2j$  error bits remaining after pass  $i$ .
- They proved that  $\delta_i(j) \leq \delta_{i-1}(j)/2$  under suitable conditions. From this, we get lower bound for decoding success probability:

$$P_i(o) \geq (1 - \sum_{i \neq 0} \delta_1(j)/2^{i-1})^{n/k_1}$$

# Comparison with Brassard-Savail's Bound



- Here we compare decoding success probability using our formula for BINARY and Brassard's lower bound for CASCADE:

Value	Our Calculation for BINARY	Brassard's Lower Bound for CASCADE
$P_1(0)$	0.00002	0.00002
$P_2(0)$	0.08808	0.00476
$P_3(0)$	0.63320	0.07106
$P_4(0)$	0.92559	0.26839
$P_5(0)$	0.98420	0.51894
$P_6(0)$	0.99492	0.72068
$P_7(0)$	0.99722	0.84902

TABLE VI  
COMPARISON,  $k_1 = 16$ ,  $n = 2048$ ,  $p = 0.03$

# Comparison with Brassard-Savail's Bound



- CASCADE has better error correction performance than BINARY because of the backtracking step.
- So decoding success probability for CASCADE should be higher than BINARY.
- From the comparison table in previous slide, we see that the lower bound of decoding success from Brassard-Savail's formula may not be tight enough.

# Conclusion



- Brassard-Savail's formula
  - Focuses on bound for error probability distribution within a block for CASCADE IR.
  - Is very easy to compute.
- But when extrapolated to deduce behavior of error distribution across the whole string, it may not be tight enough.
- Our formula
  - Calculates exact probability for BINARY IR. Hope to extend this formulation to compute probability for CASCADE.
  - Can be complex to compute for large  $n$ . Hope to find an efficient way to compute the generating functions coefficients  $C_{i,j,y}$