

# Zero error Function computation in Sensor Networks

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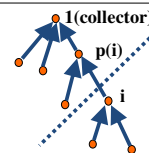
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## Introduction

- In wireless sensor network applications, nodes are not typically interested in raw data
- Instead, nodes are interested in a relevant *function* of the measurements
- Examples
  - MAXIMUM in a fire-alarm system
  - MINIMUM
  - MEAN for weather monitoring
  - MEDIAN, MODE, Boolean functions

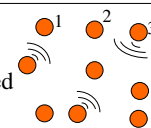
## Functions on trees: Worst-case

- Directed tree (V,E) with root node 1
- Each edge is a cut-edge
- Codebook on  $i \rightarrow p(i)$  is
 
$$\{g: X_{V_T(i)} \rightarrow D \mid \text{there exists } \hat{x}_{T(i)} \text{ s.t. } f(\hat{x}_{T(i)}, x_{V_T(i)}) = g(x_{V_T(i)}) \forall x_{V_T(i)}\}$$
- Decoder:** Assign nominal values to variables in  $X_{T(i)}$  based on codewords received on incoming edges
- Encoder:** Substitute the nominal values of variables in  $X_{T(i)}$  to obtain a function
- Theorem:** This coding scheme achieves zero-error function computation at root, and is optimal for every edge



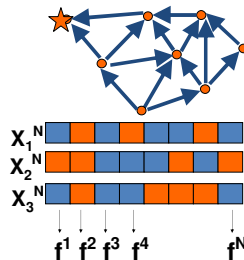
## Collocated Networks

- Broadcast scenario: every node hears transmissions from every other node
- In earlier work, Giridhar-Kumar showed that the complexity of computing *type-threshold* functions is  $O(\log n)$
- Suppose  $X_i \sim \text{Bern}(p)$ , i.i.d measurements
- Theorem:** For type-threshold functions, the average case complexity is  $O(1)$
- Example: Consider the function  $\text{AND}(X_1, X_2, \dots, X_n)$ 
  - Expected no. of bits =  $H(p) + pH(p) + \dots + p^{(n-1)}H(p) \leq H(p)/1-p$



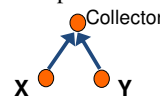
## Problem formulation

- Nodes are connected in a given directed graph topology
- Suppose a single collector node needs to compute the function block with **zero error**
- Nodes accumulate blocks of measurements and can achieve savings by **block computation**
- Objective:** To find optimal communication and in-network computation strategies



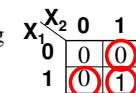
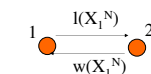
## Functions on trees: Average-case

- If  $p(x_1, x_2, \dots, x_n) > 0$  for all  $(x_1, x_2, \dots, x_n)$ , then can find optimal codes on each edge as before
- However if there are zeros in the probability matrix, we can tradeoff between links
- Example: Distributed compression will result in error
- $$\begin{matrix} X & Y \\ \begin{matrix} x_1 \\ x_2 \end{matrix} & \begin{matrix} y_1 & y_2 \\ a & \dots \\ \dots & b \end{matrix} \end{matrix}$$
- (0,1) and (1,0) are feasible
- Theorem:** If  $p(x_1, x_2, \dots, x_n) > 0$  for all  $(x_1, x_2, \dots, x_n)$ , then the optimal code on each edge is the Huffman code for the two node problem corresponding to that edge



## Interactive computation

- Nodes 1 and 2, both want to compute  $\text{AND}(X_1, X_2)$
- Trivial protocol has complexity 2 bits
- Alternate protocol:
  - Suppose node 1 releases prefix-free codeword  $l(X_1^N) = N \log_2 3 - w(X_1^N)$ . This satisfies Kraft inequality
  - Node 2 replies only for the instances where  $X_1 = 1$
  - Worst-case number of bits =  $\log_2 3!$
- Lower bound can be shown by observing that any protocol must have at least three rectangular partitions



## The Case of Two nodes

- Two nodes with measurements X, Y
- X communicates to Y only
- Y wants to compute  $f(X, Y)$  with zero-error
- Minimum worst-case/average case number of bits?



### Worst-case

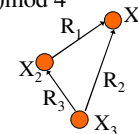
Node X separates  $x_1$  and  $x_2$  if  $\exists y$  such that  $f(x_1, y) \neq f(x_2, y)$ . Separation rule creates equivalence classes. Min. number of bits =  $\log_2(\text{no. of equiv. classes})$

### Average-case

Node X separates  $x_1$  and  $x_2$  if  $\exists y$  s.t.  $f(x_1, y) \neq f(x_2, y)$  and  $p(x_1, y)p(x_2, y) > 0$ . Separation rule creates equivalence classes if  $p(x, y) > 0$ . Huffman code is optimal

## Functions on DAGs: Issues

- Rate = Number of bits per computation
- No unique rate point that is jointly optimal for all edges  $X_1, X_2, X_3 \in \{0, 1, 2, 3\}$ ;  $f = (X_1 + X_2 + X_3) \bmod 4$
- $(R_1, R_2, R_3) = (2, 0, 2), (2, 2, 0)$  and  $(2, 1, 1)$  are all Pareto optimal
- Rate points which satisfy cut capacity constraints give outer bound to rate region.
- Theorem:** Rates achieved by only activating some tree of edges in the DAG are extreme points of the rate region
- Are these the only extreme points?
- For the parity, MAX/MIN functions, cut set bound is tight



## Boolean symmetric functions

- Extend the above approach to larger class of functions
- For a Boolean symmetric function of  $X_1, X_2, \dots, X_n$ , it is enough to know number of 1s
- Threshold type functions: Are there  $< k$  ones
- Exact complexity is  $\log_2 \binom{n+1}{k}$  bits
- Delta functions: Are there exactly  $k$  ones
- Exact complexity is  $\log_2 \binom{n+1}{k} + \binom{n}{k+1}$  bits
- Lower bound is shown by fooling set lower bounds from communication complexity theory