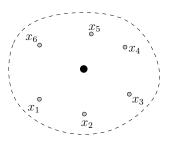
# On Analog Computation of Vector-Valued Functions in Clustered Wireless Sensor Networks

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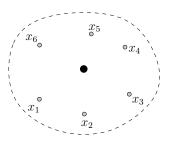
March 23, 2012

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- To combat interference, traditional schemes like TDMA were used.
- The fusion center reconstructs each sensor signal separately and subsequently computes f.
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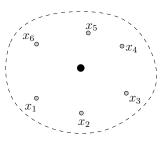


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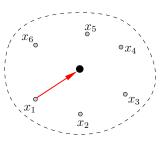
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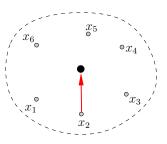
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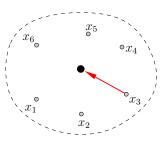
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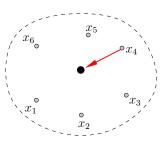


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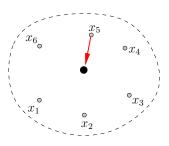




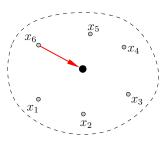
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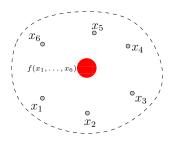
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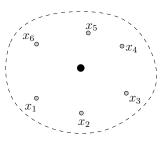
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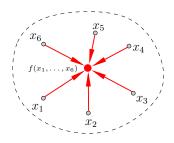
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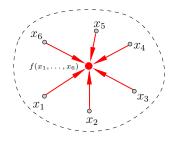
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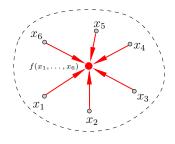
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#### Question:

• What happens in more general networks?



# Outline

- System Model & Problem Statement
- Preliminaries
- 3 Analog Computation of Vector-Valued Functions
- Performance & Properties
- Summary



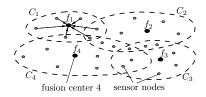
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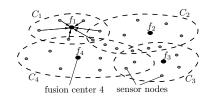
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- N sensor nodes organized into K clusters  $C_k \ (C_k \cap C_\ell \neq \emptyset)$
- $\bullet$  Sensor readings  $x_n \in [0,1]$  ,  $n=1,\dots,N$



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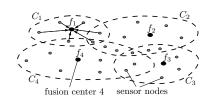


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$$y_k = \sum_{n \in C_k} h_{kn} x_n + v_k \quad , \quad k = 1, \dots, K$$

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$$\boldsymbol{f}(x_1, \dots, x_N) = \begin{pmatrix} f_1(x_{1_1}, \dots, x_{1_{|C_1|}}) \\ \vdots \\ f_K(x_{K_1}, \dots, x_{K_{|C_K|}}) \end{pmatrix}$$

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# Receive Signal

$$y = \sum_{i=1}^{N} h_i x_i + v \tag{*}$$

- ullet  $\Rightarrow$  The ideal MAC can be used to compute every function  $f:[0,1]^N \to \mathbb{R}$  that has a representation (\*).
- Examples
  - Arithmetic Mean:  $f(x_1,\ldots,x_N)=rac{1}{N}\sum_i x_i,\, \varphi_i(x)=x,\, \psi(y)=rac{1}{N}y$
  - Geometric Mean:  $f(x_1,\ldots,x_N)=\left(\prod_i x_i\right)^{1/N},\ \varphi_i(x)=\log(x),\ \psi(y)=\exp(y/N)$
  - ullet Euclidean Norm:  $f(x_1,\ldots,x_N)=\sqrt{x_1^2+\cdots+x_N^2}$ ,  $arphi_i(x)=x^2$ ,  $\psi(y)=\sqrt{y_i}$

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#### Ideal Wireless MAC

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### Observation

The space of functions (\*) is exactly the space of nomographic functions.





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Every function is universally computable via an ideal MAC, since every  $f:[0,1]^N\to\mathbb{R}$  is nomographic.

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### Theorem [Arnol'd 57], [Buck 82]

The space of nomographic functions with continuous pre- and post-processing functions is nowhere dense in the Banach-space of continuous functions.



• Let  $\ell \in \mathbb{N}$  and  $\{\varphi_{ij}\}_{1 \leq i \leq N, 1 \leq j \leq \ell}$  be a collection of  $\ell N$  pre-processing functions and consider the sequence of receive signals

$$\left\{y_1 = \sum_{i=1}^N \varphi_{i1}(x_i), \dots, y_\ell = \sum_{i=1}^N \varphi_{i\ell}(x_i)\right\}$$

- Question: Which functions are computable with such sequence of length  $\ell$ ?
- The question is closely related to the 13<sup>th</sup> Hilbert problem formulated in 1900.
- Hilbert's conjecture: The computation of continuous multivariate functions is in general not possible with finite \( \ell \).
- The conjecture was disproven by Kolmogorov in 1957. He has shown that every continuous
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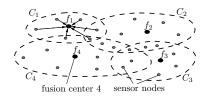


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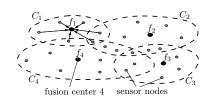
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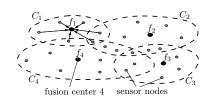
### Geometric interpretation (single cluster case):

a

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- ullet  $(x_1,\ldots,x_N)\mapsto (y_0,\ldots,y_{2N})\in \Gamma$  is a homeomorphic embedding of  $[0,1]^N$  in  $\Gamma\subset \mathbb{R}^{2N+2N}$
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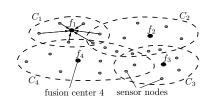
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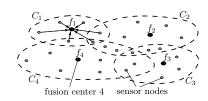
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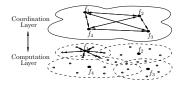
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Standard TDMA

### **Analog Computation via Channels**

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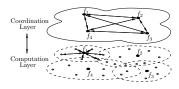


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$$\mathcal{O}(KL \ge K \max_{k} |C_k|)$$

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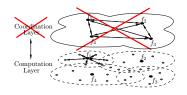
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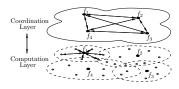
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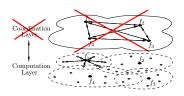
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Due to the universality property

- The hard- and software of sensor nodes is independent of  $oldsymbol{f}$
- Only dumb (cheap) nodes are required
- Feedback about functions is not necessary



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- By a geometric interpretation of Kolmogorov's theorem, we proposed an efficient analog computation scheme for clustered wireless networks
- Analog systems are gaining more attention for sensor networks (digitial signal processing has some fundamental limits [Boche, Mönich 11])
- The scheme exploits the superposition property of wireless channels
- With and without restrictions on pre- and post-processing functions, huge performance gains are possible compared with orthogonal approaches
- Coordination of nodes and clusters is not necessary
- No information about the functions at nodes is necessary (i.e., no feedback, dumb sensor nodes).
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## Thanks for your attention



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