# Let's get Physical!

ETH Zurich – Distributed Computing – www.disco.ethz.ch

ICALP 2010 – Roger Wattenhofer

### Spot the Differences



Too Many!

### Spot the Differences



Still Many!

### Spot the Differences





Better Screen Bigger Disk More RAM Cooler Design

Better Screen Bigger Disk More RAM Cooler Design

Same CPU Clock Speed



# The Future of Computing



### Why Should I Care?





# 

## Algorithms









### The Future of Computing?









### Talk Overview

### Introduction & Motivation

### Some Examples for Physical Algorithms

### What are Physical Algorithms?

### Well-Known Examples

#### Small World Phenomenon







## Natural Algorithms

+

++

[Bernard Chazelle, 2009]

T

AR



# game theory



### **Clock Synchronization**

#### **Clock Synchronization in Networks**



#### **Clock Synchronization in Networks**



#### **Problem: Physical Reality**







#### message delay



#### Clock Synchronization in Theory?

Given a communication network

- 1. Each node equipped with hardware clock with drift
- 2. Message delays with jitter



worst-case (but constant)

Goal: Synchronize Clocks ("Logical Clocks")

• Both global and local synchronization!

#### Time Must Behave!

• Time (logical clocks) should **not** be allowed to **stand still** or **jump** 



#### Time Must Behave!

• Time (logical clocks) should not be allowed to stand still or jump



- Let's be more careful (and ambitious):
- Logical clocks should always move forward
  - Sometimes faster, sometimes slower is OK.
  - But there should be a minimum and a maximum speed.
  - As close to correct time as possible!

#### Local Skew

Tree-based Algorithms e.g. FTSP Neighborhood Algorithms e.g. GTSP



#### Synchronization Algorithms: An Example ("A<sup>max</sup>")

- Question: How to update the logical clock based on the messages from the neighbors?
- Idea: Minimizing the skew to the fastest neighbor
  - Set clock to maximum clock value you know, forward new values immediately
- First all messages are slow (1), then suddenly all messages are fast (0)!


#### Local Skew: Overview of Results



#### **Experimental Results for Global Skew**



[Lenzen, Sommer, W, SenSys 2009]

#### **Experimental Results for Global Skew**



#### Clock Synchronization vs. Car Coordination

• In the future cars may travel at high speed despite a tiny safety distance, thanks to advanced sensors and communication



#### Clock Synchronization vs. Car Coordination

• In the future cars may travel at high speed despite a tiny safety distance, thanks to advanced sensors and communication



- How fast & close can you drive?
- Answer possibly related to clock synchronization
  - clock drift  $\leftrightarrow$  cars cannot control speed perfectly
  - message jitter ↔ sensors or communication between cars not perfect

## Wireless Communication

### Wireless Communication

EE, Physics Maxwell Equations Simulation, Testing 'Scaling Laws' Network Algorithms

CS, Applied Math [Geometric] Graphs Worst-Case Analysis Any-Case Analysis

#### CS Models: e.g. Disk Model (Protocol Model)









#### Signal-To-Interference-Plus-Noise Ratio (SINR) Formula



#### Example: Protocol vs. Physical Model



Assume a single frequency (and no fancy decoding techniques!)



Let  $\alpha$ =3,  $\beta$ =3, and N=10nW Transmission powers: P<sub>B</sub>= -15 dBm and P<sub>A</sub>= 1 dBm

SINR of A at D: 
$$\frac{1.26mW/(7m)^3}{0.01\mu W + 31.6\mu W/(3m)^3} \approx 3.11 \ge \beta$$
  
SINR of B at C: 
$$\frac{31.6\mu W/(1m)^3}{0.01\mu W + 1.26mW/(5m)^3} \approx 3.13 \ge \beta$$

#### This works in practice!

... even with very simple hardware (sensor nodes)



Time for transmitting 20'000 packets:

	Time required	
	standard MAC	"SINR-MAC"
Node $u_1$	721s	267s
Node $u_2$	778s	268s
Node $u_3$	780s	270s

	Messages received	
	standard MAC	"SINR-MAC"
Node $u_4$	19999	19773
Node $u_5$	18784	18488
Node $u_6$	16519	19498

Speed-up is almost a factor 3

[Moscibroda, W, Weber, Hotnets 2006]

# The Capacity of a Network

(How many concurrent wireless transmissions can you have)

#### ... is a well-studied problem in Wireless Communication



#### Network Topology?

- All these capacity studies make very strong assumptions on node deployment, topologies
  - randomly, uniformly distributed nodes
  - nodes placed on a grid
  - etc.





#### **Physical Algorithms**



#### "Convergecast Capacity" in Wireless Networks



#### Wireless Communication

EE, Physics Maxwell Equations Simulation, Testing 'Scaling Laws'

#### Network Algorithms

CS, Applied Math [Geometric] Graphs Worst-Case Analysis Any-Case Analysis

#### Possible Application – Hotspots in WLAN



#### Possible Application – Hotspots in WLAN



## Physical Algorithms?

#### **Physical Algorithms**





### no seq. input/output

### beyond laws of physics



Network



Network

Agents

# Some Unifying Theory?

#### Example: Maximal Independent Set (MIS)

- Given a mobile network, nodes with unique IDs.
- Maintain a Maximal Independent Set (MIS)
  - a non-extendable set of pair-wise non-adjacent nodes



• A simple algorithm:

- Can be implemented by constantly sending (ID, in MIS or not in MIS)
- Algorithm is simple, and it will eventually stabilize!





IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS





IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS





IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS





IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS





IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS




#### Example

IF no higher ID neighbor is in MIS  $\rightarrow$  join MIS IF higher ID neighbor is in MIS  $\rightarrow$  do not join MIS

• What if we have minor changes?



- Proof by animation: Stabilization time is linear in the diameter of the network
  - We need an algorithm that does not have linear causality chain ("butterfly effect")

#### Local Algorithms

- Given a graph, each node must determine its decision as a function of the information available within radius *t* of the node.
- Or: Each node can exchange a message with all neighbors, for t communication rounds, and must then decide.
- Or: Change can only affect nodes in distance *t*.



V

\*3

• Or: ...

#### Locality is Way to Understand Physical Algorithms



**Results: MIS** 

Join MIS with prob 1/degree, repeat



...similarly connected dominating sets, coloring, matching, covering, packing, max-min LPs, etc.

#### Lower Bound Example: Minimum Dominating Set (MDS)

- Input: Given a graph (network), nodes with unique IDs.
- Output: Find a Minimum Dominating Set (MDS)
  - Set of nodes, each node is either in the set itself, or has neighbor in set



- Differences between MIS and MDS
  - Central (non-local) algorithms: MIS is trivial, whereas MDS is NP-hard
  - Instead: Find an MDS that is "close" to minimum (approximation)
  - Trade-off between time complexity and approximation ratio

#### Lower Bound for MDS: Intuition

• Two graphs (m << n). Optimal dominating sets are marked red.



#### Lower Bound for MDS: Intuition (2)

- In local algorithms, nodes must decide only using local knowledge.
- In the example green nodes see exactly the same neighborhood.



• So these green nodes must decide the same way!

#### Lower Bound for MDS: Intuition (3)

• But however they decide, one way will be devastating (with n = m<sup>2</sup>)!



#### Graph Used in the Lower Bound

- The example is for t = 3.
- All edges are in fact special bipartite graphs with large enough girth.



 $\delta_2 \delta_1 \delta_0 \delta_3 \delta_2 \delta_0$ 

#### Lower Bounds

- Results: Many "local looking" problems need non-trivial t.
- E.g., a polylogarithmic dominating set approximation (or a maximal independent set, etc.) needs at least Ω(log Δ) and Ω(log<sup>½</sup> n) time.



[Kuhn, Moscibroda, W, 2004, 2006, 2010]

Local Algorithms ("Tight" Lower & Upper Bounds)



#### Summary & Open Problems



# Thank You!

**Questions & Comments?** 

Thanks to my co-authors Fabian Kuhn Christoph Lenzen Thomas Moscibroda Thomas Locher Johannes Schneider Philipp Sommer

www.disco.ethz.ch

## Let's get **Physical!**

### Let's get Physical!

Roger Wattenhofer

ETH Zurich – Distributed Computing – www.disco.ethz.ch