

# Energy-Efficient Algorithms

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Chair for Efficient Algorithms

## Motivation

Energy is scarce and/or expensive resource.

- **Limited availability:** Portable, battery-operated devices; sensor networks.



## Motivation

- Electricity cost: substantial strain for computing and data centers  
Google: 1 billion \$ per year

“What matters most [...] at Google is not speed, but power, low power because data centers can consume as much electricity as a city.” Eric Schmidt, *NYT* 2002

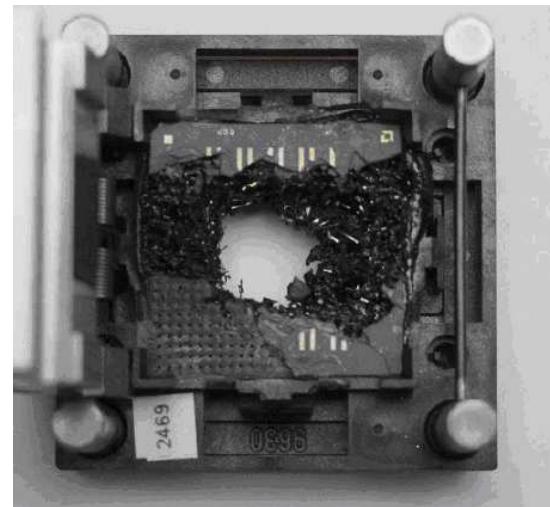
Google uses enough electricity to power 200.000 homes. *NYT* 2011

Low-cost power major criterion where to build data centers. Apple, Facebook, Google



## Motivation

- Thermal problems: Most of the energy is converted into heat.



## Energy-efficient algorithms

### Topics

- **Power-down mechanisms:** Transition system into sleep state when idle
- **Dynamic speed scaling:** Microprocessors can run at variable speed  
Intel XScale, Intel Speed Step, AMD PowerNow
- **Networks:** Optimize transmission energy

## Power-down strategies

2-state system

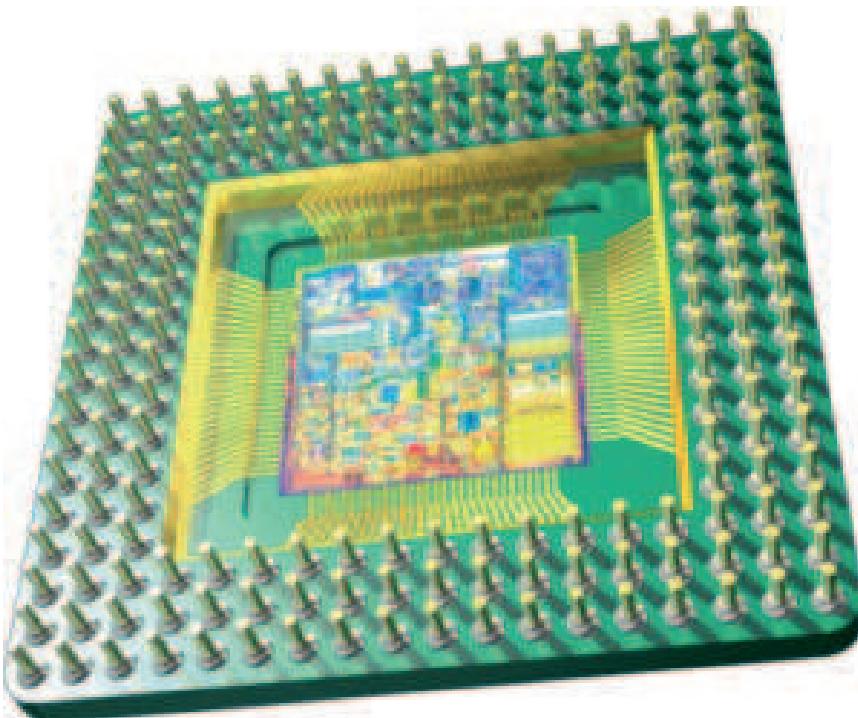
- Active state:  $r$  energy units per time unit.
- Sleep state: 0 energy units per time unit.
- Wake-up operation:  $W$  energy units.
- When active period starts, system must be in / moved to active state.



## Dynamic speed scaling

Variable-speed microprocessors

Intel XScale, Intel Speed Step, AMD PowerNow



The higher the speed, the higher  
the energy consumption

Speed  $s$

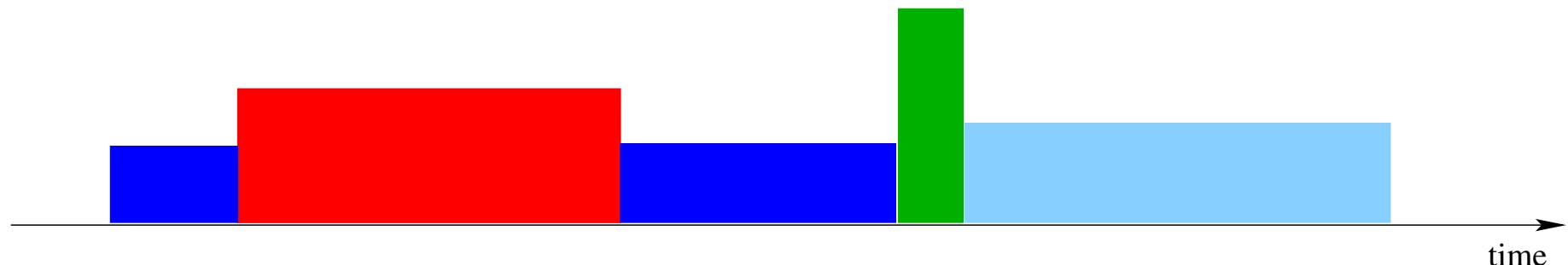
Power required

$$P(s) = s^\alpha \quad \alpha > 1$$

$P(s)$  = general convex function

## Scheduling with deadlines

1 processor



- Speed  $s$                               Power consumption  $P(s) = s^\alpha$        $\alpha > 1$
- $\sigma = J_1, \dots, J_n$
- $J_i$ :  $a_i$  = arrival time  
         $b_i$  = deadline  
         $v_i$  = processing volume       $t = v_i/s$
- Preemption allowed
- Construct feasible schedule minimizing total energy consumption.

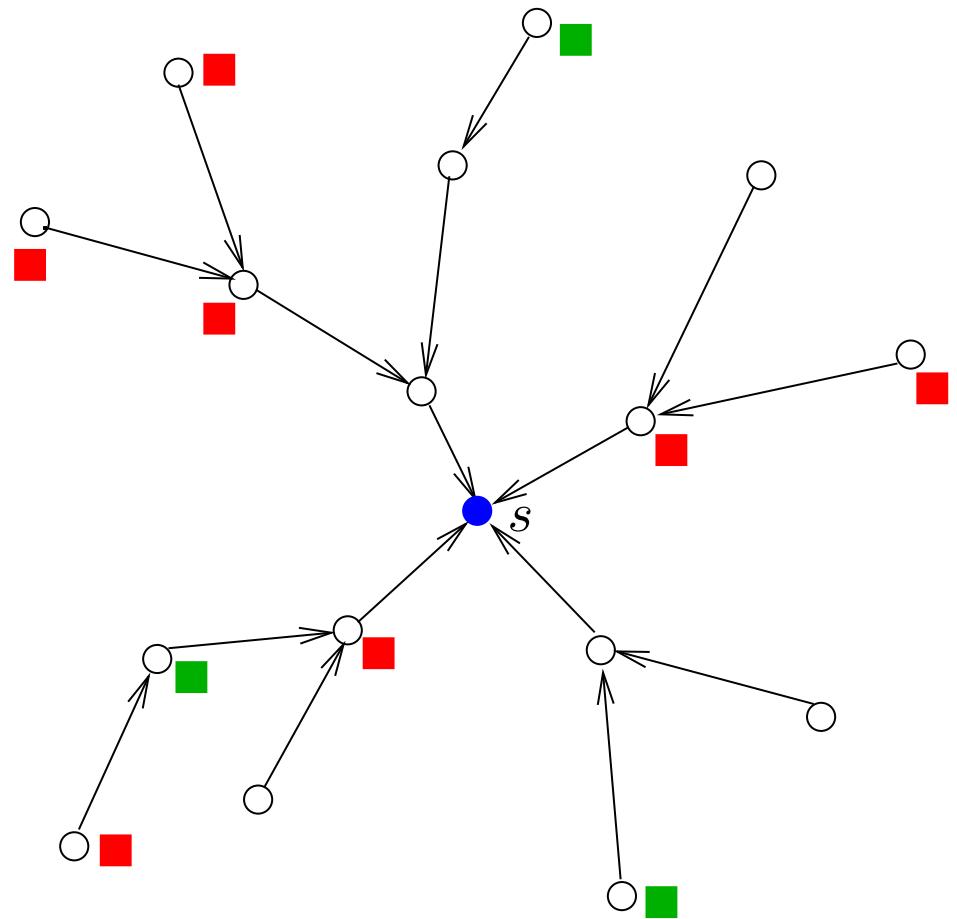
## Data aggregation

Tree topology

Data (packets) to be sent to sink

Packets may be aggregated

Energy: 1 per sending operation



## Routing

$G = (V, E)$

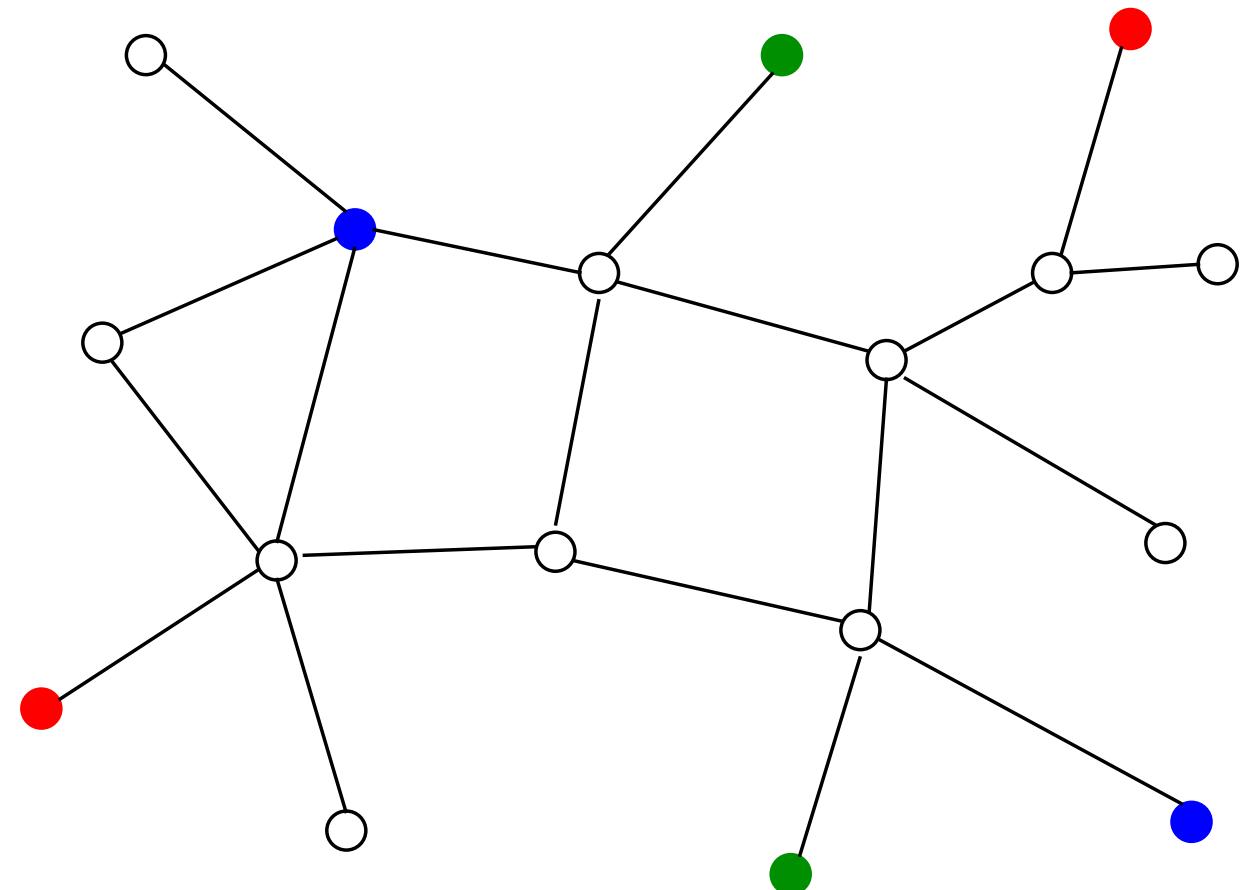
$D$  demands  $(s_i, t_i)$

$l_e$  = load on  $e$

$f(l) = c + l^\alpha$

$\min \sum_e f(l_e)$

Andrews, Antonakopoulos,  
Zhang FOCS'10



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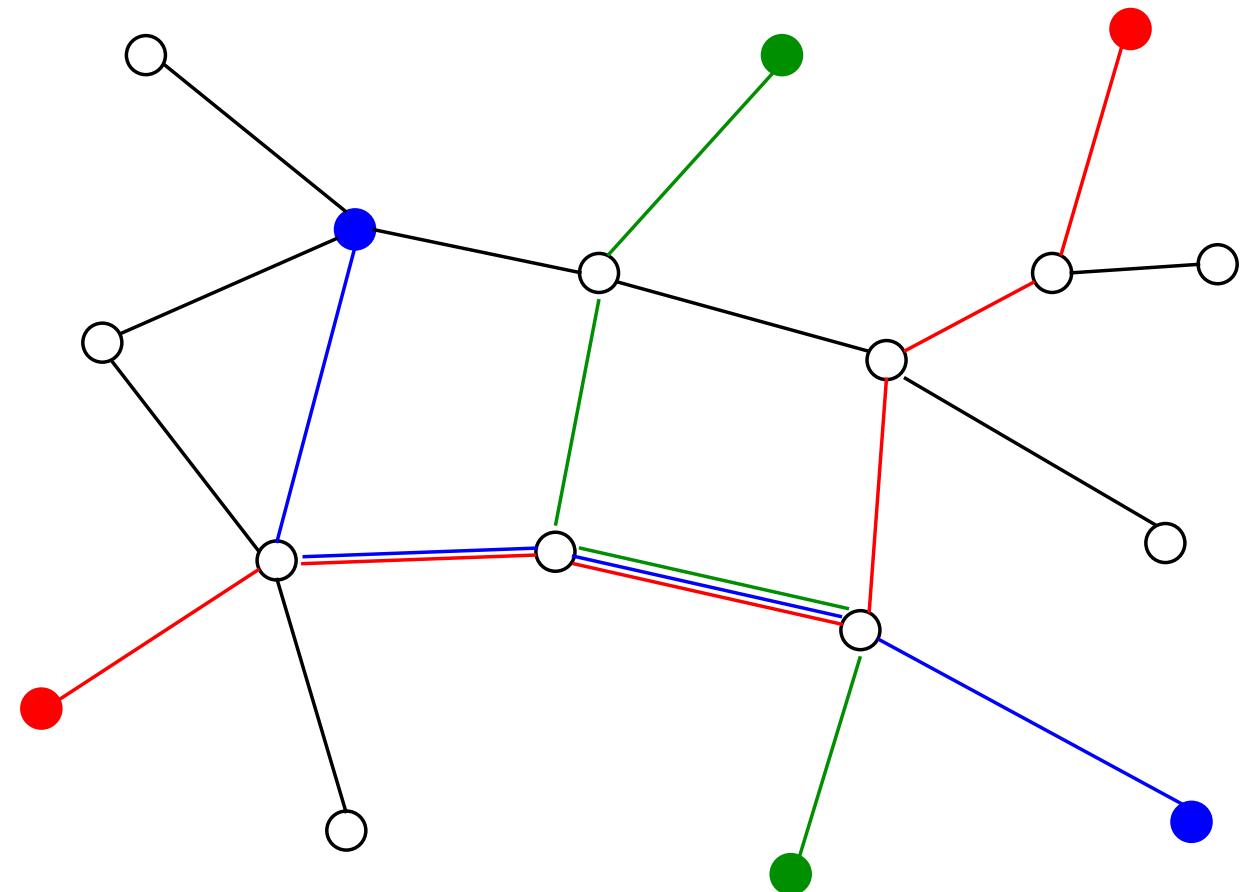
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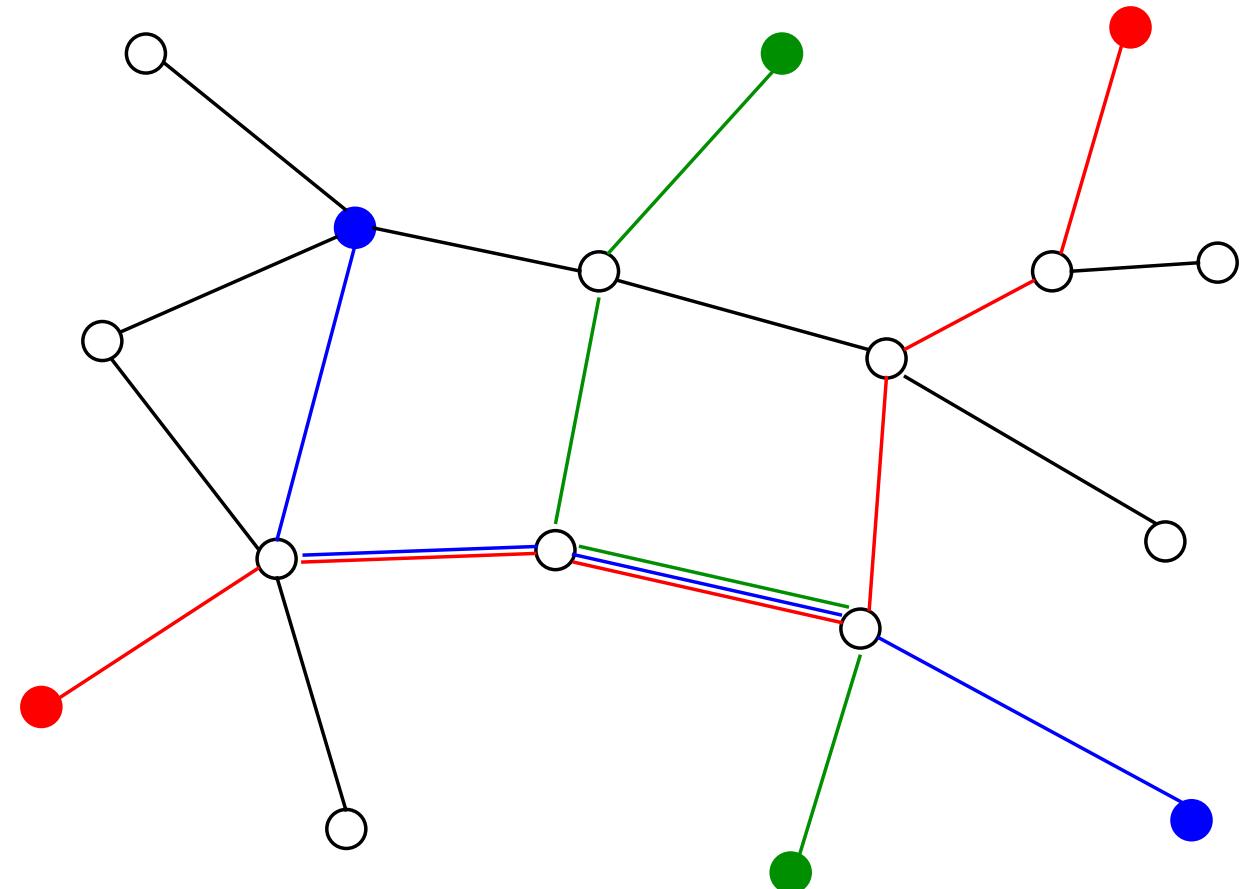
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polylog( $n, D$ )-approximation

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## Topics

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## Topics

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## Topics

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