Distributed Computing: From Principles to Pervasive, Grid and Cloud Computing
Agenda

• Principles of Distributed Computing
  – Frédéric Le Mouël, (3*2h)

• Pervasive Computing
  – Frédéric Le Mouël, (4*2h)

• Cloud Computing
  – Julien Ponge, (4*2h)

• Grid Computing
  – Yves Caniou, (4*2h)
Successful lecture?

• No mystery to succeed in research
  – Read books -> knowledge
  – Ask questions -> interactions
  – Criticize -> ideas

• Research evaluation
  – Exam, 2h
  – Ideas, method to answer a new problem
Who am I

• Frédéric Le Mouël
  – Associate Professor, INSA de Lyon, CITI Lab
  – Researcher, INRIA, Amazones Team
  – Topics: Middleware, OS, Java/OSGi, Mobile/Pervasive Computing, Ambient Intelligence, Autonomic Computing

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DC Principles : Outline

• Introduction
  – What is Distributed Computing ?
  – Why distributing ?
  – What are the problems in distributing ?

• From theory …
  – Distributed Algorithms: Byzantine problem
    • Synchronizers, Logical clocks, etc.
    • Graph colouring, Mutual exclusion, Consensus, Self-stabilization, etc.
  – Complexity
DC Principles : Outline

• ... to practice
  – Architectures
  – Message Passing
  – Message-Oriented Middleware (MOM)
What is Distributed Computing?

« Distributed computing is a field of computer science that studies distributed systems. A distributed system consists of multiple autonomous computers that communicate through a computer network. The computers interact with each other in order to achieve a common goal. A computer program that runs in a distributed system is called a distributed program, and distributed programming is the process of writing such programs. »

[wikipedia/DC] [Andrews 2000] [Ghosh 2007]
What is Distributed Computing?

- Autonomous Computers
- Networks
- Programs
What is Distributed Computing?

One Computer
Several Processors

Communication
Buses

Threads
Parallel Computing vs. Distributed Computing

Parallel computing ≈ tightly-coupled form of distributed computing

Distributed computing ≈ loosely-coupled form of parallel computing
What is Distributed Computing?

- Autonomous Computational Entities / Nodes (with their own memory)
- Communication through Message Passing
- Process
What is Distributed Computing?

« Distributed computing is a field of computer science that studies distributed systems. A distributed system consists of multiple autonomous computers that communicate through a computer network. The computers interact with each other in order to achieve a common goal. A computer program that runs in a distributed system is called a distributed program, and distributed programming is the process of writing such programs »

[wikipedia/DC] [Andrews 2000] [Ghosh 2007]
What is Distributed Computing?

Common Goal?

Process = Memory zone split in
• Data segment
• Code segment
• Execution flow (stack, heap, etc.)
What is Distributed Computing?

Goal = Physical Data Sharing

Process = Memory zone split in

• **Data segment**
• Code segment
• Execution flow (stack, heap, etc.)
What is Distributed Computing?

Goal = Physical Data Sharing

Process = Memory zone split in
• **Data segment**
• Code segment
• Execution flow (stack, heap, etc.)

k replicas upon n nodes
What is Distributed Computing?

Goal = Physical Code Sharing

Process = Memory zone split in
- Data segment
- **Code segment**
- Execution flow (stack, heap, etc.)
What is Distributed Computing?

Goal = Temporal Synchronization of Computation

Process = Memory zone split in
- Data segment
- Code segment
- Execution flow (stack, heap, etc.)

\[(A \parallel B) < (F \parallel G) < (H \parallel K \parallel L \parallel M \parallel N) < I < J\]
What is Distributed Computing?

« Distributed computing is a field of computer science that studies distributed systems. A distributed system consists of multiple autonomous computers that communicate through a computer network. The computers interact with each other in order to achieve a common goal. A computer program that runs in a distributed system is called a distributed program, and distributed programming is the process of writing such programs. »

[wikipedia/DC] [Andrews 2000] [Ghosh 2007]
What is Distributed Programming?

« Computer programming (often shortened to programming or coding) is the process of designing, writing, testing, debugging, and maintaining the source of computer programs. This source code is written in one or more programming languages. The purpose of programming is to create a program that performs specific operations or exhibits a certain desired behaviour. »

[wikipedia/CP]

Distributed Programming:
Designing = Distributed Algorithms
Writing = Programming Languages/Frameworks for Distribution
Testing, debugging, maintaining = Distributed OS/Middleware

Section 2: From theory …

Section 3: … to practice
DC Principles : Outline

• Introduction
  – What is Distributed Computing ?
  – Why distributing ?
  – What are the problems in distributing ?
• From theory …
• … to practice
Why distributing?

Application truly geographically distributed
Why distributing?

Performance gain in
- computation
- storage

Speedup
$S_n = \frac{T_1}{T_n}$

Efficiency
$E_n = \frac{T_1}{n \times T_n}$

$T_n$: execution time on $n$ nodes

[Eager 1989]
Why distributing?

Robustness: to guaranty fault-tolerance in
• computation
• storage

No Single Point Of Failure (SPOF): other nodes can execute/host the same task/data executed/hosted on the failed node
What are problems in distributing?

Dynamic Topology: the structure of the system is not known in advance and can change during execution.

- Number of nodes
- Number of links
- Links characteristics
What are problems in distributing?

Byzantine Faults:
- Omission Failures
- Commission Failures
What are problems in distributing?

Byzantine Faults:
- **Omission Failures**
- **Commission Failures**
- **Crash Failures**
- Failing to receive a request
- Failing to send a response
What are problems in distributing?

Byzantine Faults:
• Omission Failures
• Commission Failures

• Processing a request incorrectly
• Corrupting local state
• Sending an incorrect or inconsistent response to a request
What are problems in distributing?

Decision-making: each autonomous node has an incomplete view of the system that leads to uncertain decisions.

For scalability reasons, trade-off required between optimal decision and performance [Hayashibara 2002]
What are problems in distributing?

Problems typically depend on the application domain.

Section 4: Pervasive Computing

Section 5: Cloud Computing

Section 6: Grid Computing
DC Principles : Outline

• Introduction
  – What is Distributed Computing ?
  – Why distributing ?
  – What are the problems in distributing ?

• From theory …
  – Distributed Algorithms:
    • Graph colouring, Mutual exclusion,
    • Consensus, Byzantine fault-tolerance, Self-stabilization
  – Complexity
DC : Theory

• Theoretical Computer Science
  – Computational Problem:
    • An entity (human-being or other) asks a question (Input)
    • A computer performs some computation
    • The computer produces an answer (Output)
DC : Theory

• Theoretical Computer Science
  – Computability Theory:
    ⇔ Which problems are computational?
    ⇔ Which problems can be solved using a computer?
    ⇔ Which problems can be designed by an algorithm and executed by Random Access Machine or a Universal Turing Machine?
DC : Theory

- Theoretical Computer Science
  - Computational Complexity Theory: How efficient is the algorithm that solves the problem?
DC : Theory

• and Distributed Computing ?
  – Computational Problem:
    • An entity (human-being or other) asks a question (Input)
    • Several computers perform some computations and interactions
    • These computers produce an answer (Output)
DC : Theory

• and Distributed Computing ?
  – Computability Theory:
    ⇔ Which problems are computational ?
    ⇔ Which problems can be solved using a network of computers ?
    ⇔ What is the concurrent or distributed equivalent of a sequential general-purpose computer ?
DC : Theory

• and Distributed Computing ?
  – Computational Complexity Theory: How efficient is the concurrent or distributed algorithm that solves the problem ?

  Speedup
  \[ S_n = \frac{T_1}{T_n} \]

  Efficiency
  \[ E_n = \frac{T_1}{(n \times T_n)} \]

  \( T_n \): execution time on n nodes

  [Eager 1989]
An algorithm is an effective method expressed as a **finite** list of **well-defined instructions** for **calculating a function**. [...] In simple words an algorithm is a step-by-step procedure for calculations. » [wikipedia/AL]
DC : Distributed Algorithm ?

« A distributed algorithm is an algorithm designed to run on computer hardware constructed from interconnected processors » [wikipedia/DA]

Paxos Consensus Protocol [Lamport 1998]
DC : Byzantine Generals Problem

All problems comes from ... interconnected processors, truly ?

« Two armies, each led by a general, are preparing to attack a fortified city. The armies are encamped near the city, each on its own hill. A valley separates the two hills, and the only way for the two generals to communicate is by sending messengers through the valley. Unfortunately, the valley is occupied by the city's defenders and there's a chance that any given messenger sent through the valley will be captured. Note that while the two generals have agreed that they will attack, they haven't agreed upon a time for attack before taking up their positions on their respective hills. »
DC: Byzantine Generals Problem

OK

Attack!

A1

B

A2

Attack!
DC : Byzantine Generals Problem

FAIL!

Attack!
DC : Byzantine Generals Problem

Let’s attack at 3pm ?

FAIL!
DC : Byzantine Generals Problem

Let's attack at 3pm?

FAIL!

ok

A1

A2

No solution to the problem [Akkoyunlu 1975] and can even be more complex: n generals, traitors, etc.

Solution to generate BFT protocols [Guerraoui 2010]
DC : Classic Distributed Algorithms

- Data validation \rightarrow Order
  - Synchronizers
  - Logical Clock
  - Clock Synchronization
  - Atomic commit
  - Replication

Time is also a data!
DC : Classic Distributed Algorithms

• Resource allocation
  – Mutual exclusion
  – Graph colouring

• Decision-making
  – Leader election
  – Consensus
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Synchronizers

  – « A Synchronizer is an algorithm that can be used to run a synchronous algorithm on top of asynchronous processor network » [Awerbuch 1985]
  • Alpha/Beta/Gamma synchronizer
DC : Classic
Distributed Algorithms

• Data validation ➔ Order
  – Synchronizer : Alpha Synchronizer

  – In round r, the synchronizer at p sends p's message (tagged with the round number) to each neighbor p' or no-msg(r) if it has no messages. When it collects a message or no-msg from each neighbor for round r, it delivers all the messages.

  – Allow to model simplified "ideal network"
DC : Classic Distributed Algorithms

- Data validation ➔ Order
  - Logical Clock is a mechanism for capturing chronological and causal relationships in a distributed system
    - Lamport timestamps
    - Vector clocks
    - Version vectors
    - Matrix clocks
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Logical Clock: Lamport timestamps [Lamport 1978]
    
    – A process increments its counter before each event in that process;
    – When a process sends a message, it includes its counter value with the message;
    – On receiving a message, the receiver process sets its counter to be greater than the maximum of its own value and the received value before it considers the message received.
DC : Classic
Distributed Algorithms

• Data validation ➔ Order
  – Logical Clock: Lamport timestamps [Lamport 1978]
DC : Classic
Distributed Algorithms

• Data validation ➔ Probability
  – Clock Synchronization: Internal clocks of different computers differs because of the clock drift

• Trivial algorithms in a centralized approach [Cristian 1989] [Gusella 1989]:
  – P requests the time from S
  – After receiving the request from P, S prepares a response and appends the time T from its own clock
  – P then sets its time to be T + RTT/2
DC : Classic
Distributed Algorithms

• Data validation ➔ Probability
  – Clock Synchronisation: Internal clocks of different computers differs because of the clock drift

• In a distributed network, Marzullo's algorithm [Marzullo 1984] is an agreement algorithm used to select sources for estimating accurate time from a number of noisy time sources
  – A refined version, “the intersection algorithm” is used in NTP [Mills 2006]
DC : Classic Distributed Algorithms

• Data validation $\Rightarrow$ Probability
  – Clock Synchronization: intersection algorithm

[Diagram of intervals and intersections with notes about consistency and largest number of sources]
DC : Classic Distributed Algorithms

• Data validation ➔ Probability
  – Clock Synchronization: intersection algorithm

[diagram showing intervals and their intersections, indicating data consistency]

[wikipedia/MA]
DC : Classic Distributed Algorithms

• Data validation ➔ Probability
  – Clock Synchronization: intersection algorithm
DC : Classic Distributed Algorithms

• Data validation ➔ Probability
  – Clock Synchronization: depending on network topology
    • CS-MNS: Clock Sampling Mutual Network Synchronization is suitable for distributed and mobile applications [Rentel 2005]
    • Reference broadcast synchronization is often used in wireless networks and sensor networks. An initiator broadcasts a reference message to urge the receivers to adjust their clocks [Ganeriwal 2003] [Maroti 2004]
    • PTP: Precision Time Protocol is used in measurement and control systems in local area networks and can achieve sub-microsecond accuracy [IEEE 1588]
DC : Classic
Distributed Algorithms

• Data validation ➔ Order
  – Atomic commit

  – An atomic commit is an operation where a set of distinct changes is applied as a single operation. If the atomic commit succeeds, it means that all the changes have been applied. If there is a failure before the atomic commit can be completed, the "commit" is aborted and no changes will be applied.
DC : Classic
Distributed Algorithms

• Data validation ➔ Order
  – Atomic commit : two-phase algorithm

• The coordinator sends a query to commit message to all cohorts and waits until it has received a reply from all cohorts.
• The cohorts execute the transaction up to the point where they will be asked to commit. They each write an entry to their undo log and an entry to their redo log
• Each cohort replies with an agreement message (cohort votes Yes to commit), if the cohort's actions succeeded, or an abort message (cohort votes No, not to commit), if the cohort experiences a failure that will make it impossible to commit.
DC : Classic
Distributed Algorithms

- Data validation ➔ Order
  - Two-phase algorithm
  - Pb: blocking algo

A single node will continue to wait even if all other sites have failed.
If the coordinator fails permanently, some cohorts will never resolve their transactions.

Figure 1 • The two-phase commit protocol
DC : Classic Distributed Algorithms

- Three-phase algorithm [Skeen 1983]
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Replication :

« Replication is the process of sharing information so as to ensure consistency between redundant resources, such as software or hardware components, to improve reliability, fault-tolerance, or accessibility. It could be data replication if the same data is stored on multiple storage devices, or computation replication if the same computing task is executed many times. A computational task is typically replicated in space, i.e. executed on separate devices, or it could be replicated in time, if it is executed repeatedly on a single device. » [wikipedia/R]
DC : Classic
Distributed Algorithms

• Data validation ➔ Order
  – Replication : two kinds

  – Active replication : each request is executed on each replica and each of these modifications are transferred to all replica
  – Passive replication : each request is executed on a single replica and this modification is transferred to all replica (master-slave)
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Replication : three algorithms models

  – Transactional replication (active/passive) : transactions on replicated data in accordance with ACID properties (atomicity, consistency, isolation, durability)

  – (slow, on copy-guaranty)
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Replication: three algorithms models
  – State machine replication (active): deterministic finite state machine, atomic broadcast, distributed consensus
  – (mid-fast)
DC: Classic Distributed Algorithms

- Data validation ➔ Order
  - Replication: three algorithms models
  - State machine replication (active)

Paxos Consensus Protocol [Lamport 1998]
DC : Classic Distributed Algorithms

• Data validation ➔ Order
  – Replication : three algorithms models
  – Virtual synchrony (active/passive) : process group, checkpoint, multicast
    – (fastest, less rigorous)
DC : Classic Distributed Algorithms

- Data validation ➔ Order
  - Replication: three algorithms models
  - Virtual synchrony (active/passive)

[Défago 2004]
DC : Classic
Distributed Algorithms

• Resource allocation
  – Mutual exclusion
  – Graph colouring

• Decision-making
  – Leader election
  – Consensus
DC : Classic Distributed Algorithms

• Resource allocation
  – Mutual exclusion (Mutex) : algorithm to avoid simultaneous use of a common resource. This piece of code in which a process access a common resource is called a critical section.

  – Multiple use : locks, reentrant mutexes, semaphores, monitors, message passing, tuple space
DC : Classic
Distributed Algorithms

• Resource allocation
  – Mutual exclusion (Mutex) : different algorithms [Raynal 1986]
    – Dekker's algorithm
    – Peterson's algorithm
    – Lamport's bakery algorithm
    – Szymanski's algorithm
DC : Classic Distributed Algorithms

- Resource allocation
  - Mutual exclusion (Mutex) : different algorithms
  - Dekker's algorithm in 1965 [E.W. Dijkstra 2009]

```plaintext
flag[0] := false
flag[1] := false
turn := 0   // or 1

flag[0] := true
while flag[1] = true {
  if turn ≠ 0 {
    flag[0] := false
    while turn ≠ 0 {}
    flag[0] := true
  }
}

// critical section
...

turn := 1
flag[0] := false
```
DC : Classic Distributed Algorithms

• Resource allocation
  – Graph colouring :

« Graph coloring is a special case of graph labeling; it is an assignment of labels traditionally called "colors" to elements of a graph subject to certain constraints. In its simplest form, it is a way of coloring the vertices of a graph such that no two adjacent vertices share the same color; this is called a vertex coloring. » [wikipedia/GC]
DC : Classic Distributed Algorithms

- Resource allocation
  - Graph colouring :
    - Multiple use : scheduling, bandwidth allocation, compilation register allocation, pattern matching, Sudoku 😊 (9-coloring on given specific graph with 81 vertices)
DC : Classic Distributed Algorithms

• Resource allocation
  – Graph colouring : Multi-trials technique allows to break symmetry efficiently [Schneider 2010] : randomized fastest algorithm that employs more attempts with every message exchange
  – Idea : In classical algorithms, every uncolored node randomly picks an available color and keeps it if no neighbor (concurrently) chooses the same color. For the multi-trials technique, a node gradually increases the number of chosen colors in every communication rounds.
DC : Classic
Distributed Algorithms

• Decision-making
  – Leader election

« Leader election is the process of designating a single process as the organizer of some task distributed among several computers (nodes). » [wikipedia/LE]
DC : Classic
Distributed Algorithms

• Decision-making
  – Leader election : different algorithms [Korach 1990]
    – Bully algorithm : higher process ID
    – Chang and Roberts algorithm : UID, token ring
    – Gallager, Humblet, and Spira algorithm
    – Voting System
A minimum spanning tree (MST) or minimum weight spanning tree is then a spanning tree with weight less than or equal to the weight of every other spanning tree.
DC : Classic
Distributed Algorithms

• Decision-making
  – Consensus

“Consensus is the problem of task of group agreement in the presence of faults.”

Consensus has been shown to be impossible to solve in several models of distributed computing [Fischer 1985] [Loui 1987]
DC : Classic Distributed Algorithms

• Decision-making
  – Consensus : different algorithms
    – Randomized consensus algorithm [Aspnes 1990]
    – Chandra–Toueg consensus algorithm [Chandra 1996]
    – Paxos consensus algorithms family [wikipedia/Paxos]
DC : Classic Distributed Algorithms

• Decision-making
  – Consensus: Paxos Consensus Protocol [Lamport 1998]
DC : Classic
Distributed Algorithms

• Decision-making
  – Consensus: Paxos Consensus Protocol [Lamport 1998]

  – Quorums: Quorums express the safety properties of Paxos by ensuring at least some surviving processor retains knowledge of the results.
DC : Classic Distributed Algorithms

• Decision-making
  – Consensus: Paxos Consensus Protocol [Lamport 1998]

  – Quorums: Typically, a Quorum is any majority of participating Acceptors. For example, given the set of Acceptors \{A,B,C,D\}, a majority Quorum would be any three Acceptors: \{A,B,C\}, \{A,C,D\}, \{A,B,D\}, \{B,C,D\}. More generally, arbitrary positive weights can be assigned to Acceptors and a Quorum defined as any subset of Acceptors with the summary weight greater than half of the total weight of all Acceptors.
DC : Classic Distributed Algorithms

• Decision-making
  – Consensus: Paxos algorithms family

  – … according to conditions: different kind of failures of (participants: acceptors, etc.), multi-Paxos, cheap Paxos, fast-Paxos (conflicting or not), BFT-Paxos, virtual-Paxos, etc.
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  – Why distributing ?
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• From theory …
  – Distributed Algorithms: Byzantine problem
    • Synchronizers, Logical clocks, etc.
    • Graph colouring, Mutual exclusion, Consensus, Self-stabilization, etc.
  – Complexity
DC Complexity

• 5 aspects of distributed computing system complexity [Ranganathan 2007] :
  – Task-Structure Complexity,
  – Unpredictability,
  – Size Complexity,
  – Chaotic Complexity
  – and Algorithmic Complexity
DC Complexity

• Task-Structure Complexity
  – Cyclomatic complexity (CC) is:

\[ CC = E - N + p \]

where \( E \) = the number of edges of the task graph
\( N \) = the number of nodes of the task graph
\( p \) = the number of connected components
DC Complexity

- Task-Structure Complexity
  - Cyclomatic complexity (CC) measures the decision points

CC=9-9+1=1

CC=24-20+1=5
DC Complexity

• Unpredictability
  – Entropy $H$ of the system

  – Higher the entropy of the system, the more difficult to predict

  – $k$ different states with probabilities $p_1$, $p_2$, etc
  – $\log_2(1/pi)$ the surprisal factor

\[
H = \sum_{i=1}^{k} p_i \log_2 \left( \frac{1}{p_i} \right)
\]
DC Complexity

• Size Complexity
  – Traditionally, the size of a distributed system is measured by the number of nodes, devices, services, applications or other components.
  – In addition, a distributed system may have high cognitive complexity if users need to be aware of a large number of concepts in order to use the system. A concept is any logical item of knowledge defined or used by the system. A concept includes abstract notions like file-types, security policies, context information, device characteristics and QoS parameters. A large number of concepts contributes to greater difficulty in understanding the system as a whole.
DC Complexity

• Chaotic Complexity
  • Chaotic Complexity refers to the property of systems by which small variations in a certain part of the system can have large effects on overall system behaviour.
  • Chaotic complexity makes it difficult to understand systems. It often results from a lack of modular design and from a number of inter-dependencies between different parts of the distributed system.
DC Complexity

• Chaotic Complexity
  • An important factor contributing to chaotic complexity is coupling between different components.
    – Components are data coupled if they pass data through scalar or array parameters.
    – Components are control coupled if one passes a value that is used to control the internal logic of the other.
    – Components are common coupled if they refer to the same global data.
    – Components are content coupled if they access and change each other's internal data state or procedural state.
DC Complexity

• Chaotic Complexity: measuring coupling between components is fan in - fan out complexity
  – Count of the number of data flows into and out of a component plus the number of global data structures that the component updates.

Complexity = Length * (Fan-in * Fan-out)^2
Length is any measure of length such as lines of code.
DC Complexity

• Cognitive Algorithmic Complexity: Halstead’s measures [Halstead 1977]

\[ n_1 = \text{the number of distinct operators} \]
\[ n_2 = \text{the number of distinct operands} \]
\[ N_1 = \text{the total number of operators} \]
\[ N_2 = \text{the total number of operands} \]
DC Complexity

- Cognitive Algorithmic Complexity: Halstead’s measures [Halstead 1977]

Program length \( (N) = N_1 + N_2 \)
Program vocabulary \( (n) = n_1 + n_2 \)
Program Volume \( (V) = N \times (\log_2 n) \). The program volume measures the information content of a program or the size of implementation of an algorithm.
DC Complexity

• Cognitive Algorithmic Complexity: Halstead’s measures [Halstead 1977]

Difficulty ($D$) = \( \frac{n_1}{2} \times \frac{N_2}{n_2} \). The difficulty of a program is also related to the error-proneness of the program.

Effort ($E$) = $D \times V$. The effort to implement or understand a program is proportional to the volume and to the difficulty level of the program.
DC Complexity

• Algorithmic Complexity
  – The traditional definition of the complexity of an algorithm is in terms of its time and space requirements or its relation to Turing machines or universal computers

  – A simple example is the use of an $O(n^2)$ algorithm for sorting (like insertion-sort or bubble-sort) as opposed to an $O(n \log n)$ algorithm (like quick-sort).
DC Complexity

• Algorithmic Complexity
  – The traditional definition of the complexity of an algorithm is in terms of its time and space requirements or its relation to Turing machines or universal computers
  
    – A simple example is the use of an $O(n^2)$ algorithm for sorting (like insertion-sort or bubble-sort) as opposed to an $O(n \log n)$ algorithm (like quick-sort).
DC Complexity

- Algorithmic Complexity
  - In distributed algorithms, another resource in addition to time and space is the number of computers
  - More attention is usually paid on communication operations than computational steps
DC Complexity

- Algorithmic Complexity
  - The complexity measure is closely related to the diameter D of the network
  - « longest shortest path »
DC Complexity

• Algorithmic Complexity
  – Diameter $D$ of a graph: a graph's diameter is the largest number of vertices which must be traversed in order to travel from one vertex to another when paths which backtrack, detour, or loop are excluded from consideration

\[
\begin{align*}
D=3 & & D=4 & & D=5 & & D=7 \\
\end{align*}
\]
DC Complexity

• Algorithmic Complexity
  – The complexity measure is closely related to the diameter $D$ of the network

  – Any computable problem can be solved trivially in a synchronous distributed system in approximately $2D$ communication rounds: simply gather all information in one location ($D$ rounds), solve the problem, and inform each node about the solution ($D$ rounds).
DC Complexity

• Algorithmic Complexity
  – The complexity measure is closely related to the diameter $D$ of the network
  – Goal: if the running time of the algorithm is much smaller than $D$ communication rounds, then the nodes in the network must produce their output without having the possibility to obtain information about distant parts of the network. In other words, the nodes must make globally consistent decisions based on information that is available in their local neighbourhood.
DC Principles : Outline

• ... to practice
  – Architectures
  – Message Passing
  – Message-Oriented Middleware (MOM)
DC Principles : Outline

• ... to practice
  – Architectures
  – Message Passing
  – Message-Oriented Middleware

• .. in the field of
  – Pervasive Computing ➔ Frédéric Le Mouël
  – Cloud Computing ➔ Julien Ponge
  – Grid Computing ➔ Yves Caniou
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