



PA160: Net-Centric Computing II.

Distributed Systems

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Lecture overview

Distributed Systems

- Key characteristics
- Challenges and Issues
- Distributed System Architectures
- Inter-process Communication

Middleware

- Remote Procedure Calls (RPC)
- Remote Method Invocation (RMI)
- Common Object Request Broker Architecture (CORBA)

Service Oriented Architecture (SAO)

Web Services

Issues Examples

- Scheduling/Load-balancing in Distributed Systems
- Fault Tolerance in Distributed Systems

Conclusion

Scheduling/Load-balancing in Distributed Systems

For concurrent execution of interacting processes:

- **communication** and **synchronization between processes** are the two essential system components

Before the processes can execute, they need to be:

- **scheduled** and
- allocated with resources

Why scheduling in distributed systems is of special interest?

- because of the issues that are different from those in traditional multiprocessor systems:
 - *the communication overhead is significant*
 - *the effect of underlying architecture cannot be ignored*
 - *the dynamic behaviour of the system must be addressed*
- local scheduling (on each node) + global scheduling



Scheduling/Load-balancing in Distributed Systems

- let's have a pool of jobs
 - there are some inter-dependencies among them
- and a set of nodes (processors) able to reciprocally communicate

Load-balancing

The term **load-balancing** means assigning the jobs to the processors in the way, which minimizes the time/communication overhead necessary to compute them.

- load-balancing – divides the jobs among the processors
- scheduling – defines execution order of the jobs (on each processor)
 - load-balancing and planning are tightly-coupled (synonyms in DSs)
- **objectives:**
 - enhance overall system performance metric
 - process completion time and processor utilization
 - location and performance transparency

Scheduling/Load-balancing in Distributed Systems

- the scheduling/load-balancing task can be represented using graph theory:
 - the pool of N jobs with dependencies can be described as a graph $G(V, U)$, where
 - the nodes represent the jobs (processes)
 - the edges represent the dependencies among the jobs/processes (e.g., an edge from i to j requires that the process i has to complete before j can start executing)
 - the graph G has to be split into p parts, so that:
 - $N = N_1 \cup N_2 \cup \dots \cup N_p$
 - which satisfy the condition, that $|N_i| \approx \frac{|N|}{p}$, where
 - $|N_i|$ is the number of jobs assigned to the processor i , and
 - p is the number of processors, and
 - the number/cost of the edges connecting the parts is minimal
 - the objectives:
 - uniform jobs' load-balancing
 - minimizing the communication (the minimal number of edges among the parts)
 - the splitting problem is NP-complete
 - the heuristic approaches have to be used

Scheduling/Load-balancing in Distributed Systems

An illustration

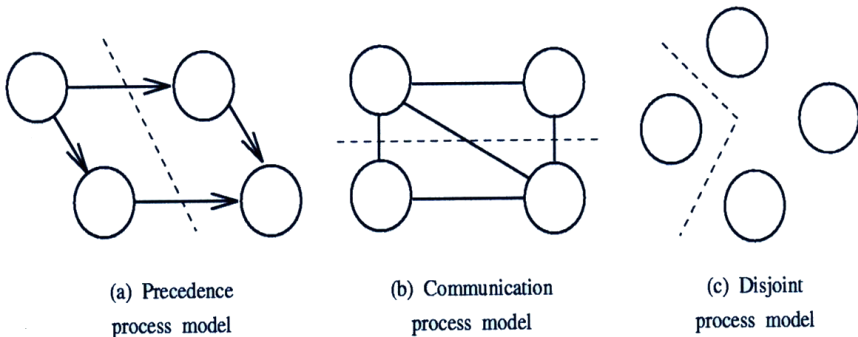


Figure: An illustration of splitting 4 jobs onto 2 processors.



Scheduling/Load-balancing in Distributed Systems

- the “proper” approach to the scheduling/load-balancing problem depends on the following criteria:
 - *jobs' cost*
 - *dependencies among the jobs*
 - *jobs' locality*

Scheduling/Load-balancing in Distributed Systems

Jobs' Cost

- the job's cost may be known:
 - before the whole problem set's execution
 - during problem's execution, but before the particular job's execution
 - just after the particular job finishes
- *cost's variability* – all the jobs may have (more or less) the same cost or the costs may differ
- the problem classes based on jobs' cost:
 - all the jobs have the same cost: *easy*
 - the costs are variable, but, known: *more complex*
 - the costs are unknown in advance: *the most complex*

Scheduling/Load-balancing in Distributed Systems

Dependencies Among the Jobs

- is the order of jobs' execution important?
- the dependencies among the jobs may be known:
 - before the whole problem set's execution
 - during problem's execution, but before the particular job's execution
 - are fully dynamic
- the problem classes based on jobs' dependencies:
 - the jobs are fully independent on each other: *easy*
 - the dependencies are known or predictable: *more complex*
 - flooding
 - in-trees, out-trees (balanced or unbalanced)
 - generic oriented trees (DAG)
 - the dependencies dynamically change: *the most complex*
 - e.g., searching/lookup problems



Scheduling/Load-balancing in Distributed Systems

Locality

- communicate all the jobs in the same/similar way?
- is it suitable/necessary to execute some jobs “close” to each other?
- when the job’s communication dependencies are known?
- the problem classes based on jobs’ locality:
 - the jobs do not communicate (at most during initialization): *easy*
 - the communications are known/predictable: *more complex*
 - regular (e.g., a grid) or irregular
 - the communications are unknown in advance: *the most complex*
 - e.g., a discrete events’ simulation



Scheduling/Load-balancing in DSs – Solving Methods

- in general, the “proper” solving method depends on the time, when the particular information is known
- basic solving algorithms' classes:
 - *static* – offline algorithms
 - *semi-static* – hybrid approaches
 - *dynamic* – online algorithms
- some (but not all) variants:
 - static load-balancing
 - semi-static load-balancing
 - self-scheduling
 - distributed queues
 - DAG planning

Scheduling/Load-balancing in DSs – Solving Methods

Semi-static load-balancing

- suitable for problem sets with slow changes in parameters, and with locality importance
- iterative approach
 - uses static algorithm
 - the result (from the static algorithm) is used for several steps (slight unbalance is accepted)
 - after the steps, the problem set is recalculated with the static algorithm again
- often used for:
 - particle simulation
 - calculations of slowly-changing grids (but in a different sense than in the previous lectures)

Scheduling/Load-balancing in DSs – Solving Methods

Self-scheduling I.

- a centralized pool of jobs
- idle processors pick the jobs from the pool
- new (sub)jobs are added to the pool
- + ease of implementation
- suitable for:
 - a set of independent jobs
 - jobs with unknown costs
 - jobs where locality does not matter
- unsuitable for too small jobs – due to the communication overhead
 - \Rightarrow coupling jobs into bulks
 - fixed size
 - controlled coupling
 - tapering
 - weighted distribution

Scheduling/Load-balancing in DSs – Solving Methods

Self-scheduling II. – Fixed size & Controlled coupling

Fixed size

- typical offline algorithm
- requires much information (number and cost of each job, ...)
- it is possible to find the optimal solution
- theoretically important, not suitable for practical solutions

Controlled coupling

- uses bigger bulks in the beginning of the execution, smaller bulks in the end of the execution
 - lower overhead in the beginning, finer coupling in the end
- the bulk's size is computed as: $K_i = \lceil \frac{R_i}{p} \rceil$
where:
 - R_i ... the number of remaining jobs
 - p ... the number of processors

Scheduling/Load-balancing in DSs – Solving Methods

Self-scheduling II. – Tapering & Weighted distribution

Tapering

- analogical to the Controlled coupling, but the bulks' size is further a function of jobs' variation
- uses historical information
 - low variance \Rightarrow bigger bulks
 - high variance \Rightarrow smaller bulks

Weighted distribution

- considers the nodes' computational power
- suitable for heterogenous systems
- uses historical information as well

Scheduling/Load-balancing in DSs – Solving Methods

Distributed Queues

- \approx self-scheduling for distributed memory
- instead of a centralized pool, a queue on each node is used (per-processor queues)
- suitable for:
 - distributed systems, where the locality does not matter
 - for both static and dynamic dependencies
 - for unknown costs
- an example: diffuse approach
 - in every step, the cost of jobs remaining on each processor is computed
 - processors exchange this information and perform the balancing
 - locality must not be important

Scheduling/Load-balancing in DSs – Solving Methods

Centralised Pool vs. Distributed Queues

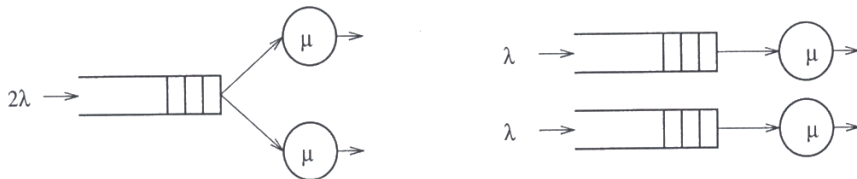


Figure: Centralised Pool (left) vs. Distributed Queues (right).

Scheduling/Load-balancing in DSs – Solving Methods

DAG Planning

DAG Planning

- another graph model
 - the nodes represent the jobs (possibly weighted)
 - the edges represent the dependencies and/or the communication (may be also weighted)
- e.g., suitable for digital signal processing
- basic strategy – divide the DAG so that the communication and the processors' occupation (time) is minimized
 - NP-complete problem
 - takes the dependencies among the jobs into account



Scheduling/Load-balancing in DSs – Design Issues I.

When the scheduling/load-balancing is necessary?

- for middle-loaded systems
 - lowly-loaded systems – rarely job waiting (there's always an idle processor)
 - highly-loaded systems – little benefit (the load-balancing cannot help)

What is the performance metric?

- mean response time

What is the measure of load?

- must be easy to measure
- must reflect performance improvement
- example: queue lengths at CPU, CPU utilization

Scheduling/Load-balancing in DSs – Design Issues I.

Types of policies:

- *static* (decisions hardwired into system), *dynamic* (uses load information), *adaptive* (policy varies according to load)

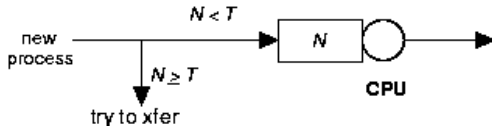
Policies:

- *Transfer policy*: when to transfer a process?
 - threshold-based policies are common and easy
- *Selection policy*: which process to transfer?
 - prefer new processes
 - transfer cost should be small compared to execution cost
 - \Rightarrow select processes with long execution times
- *Location policy*: where to transfer the process?
 - polling, random, nearest neighbor, etc.
- *Information policy*: when and from where?
 - demand driven (only a sender/receiver may ask for), time-driven (periodic), state-change-driven (send update if load changes)

Scheduling/Load-balancing in DSs – Design Issues II.

Sender-initiated Policy

- Transfer policy



- Selection policy: newly arrived process

- Location policy: three variations

- Random – may generate lots of transfers
 - ⇒ necessary to limit max transfers
- Threshold – probe n nodes sequentially
 - transfer to the first node below the threshold, if none, keep job
- Shortest – poll N_p nodes in parallel
 - choose least loaded node below T
 - if none, keep the job



Scheduling/Load-balancing in DSs – Design Issues II.

Receiver-initiated Policy

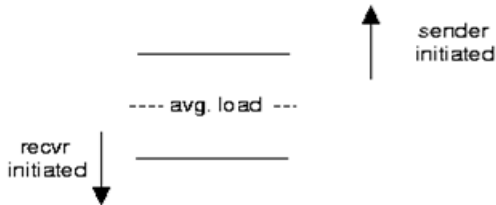
- *Transfer policy*: if departing process causes load $< T$, find a process from elsewhere
- *Selection policy*: newly arrived or partially executed process
- *Location policy*:
 - Threshold – probe up to N_p other nodes sequentially
 - transfer from first one above the threshold; if none, do nothing
 - Shortest – poll n nodes in parallel
 - choose the node with heaviest load above T



Scheduling/Load-balancing in DSs – Design Issues II.

Symmetric Policy

- combines previous two policies without change
 - nodes act as both senders and receivers
- uses average load as the threshold





Scheduling/Load-balancing in DSs – Case study

V-System (Stanford)

- state-change driven information policy
 - significant change in CPU/memory utilization is broadcast to all other nodes
- M least loaded nodes are receivers, others are senders
- sender-initiated with new job selection policy
- *Location policy*:
 - probe random receiver
 - if still receiver (below the threshold), transfer the job
 - otherwise try another



Scheduling/Load-balancing in DSs – Case study Sprite (Berkeley) I.

- *Centralized information policy*: coordinator keeps info
 - state-change driven information policy
 - Receiver: workstation with no keyboard/mouse activity for the defined time period (30 seconds) and below the limit (active processes < number of processors)
- *Selection policy*: manually done by user \Rightarrow workstation becomes sender
- *Location policy*: sender queries coordinator
- the workstation with the foreign process becomes sender if user becomes active



Scheduling/Load-balancing in DSs – Case study

Sprite (Berkeley) II.

- *Sprite process migration:*
 - facilitated by the Sprite file system
 - state transfer:
 - swap everything out
 - send page tables and file descriptors to the receiver
 - create/establish the process on the receiver and load the necessary pages
 - pass the control
 - the only problem: communication-dependencies
 - solution: redirect the communication from the workstation to the receiver

Scheduling/Load-balancing in DSs

Code and Process Migration

- key reasons: *performance* and *flexibility*
- flexibility:
 - dynamic configuration of distributed system
 - clients don't need preinstalled software (download on demand)
- process migration (*strong mobility*)
 - process = code + data + stack
 - examples: Condor, DQS
- code migration (*weak mobility*)
 - transferred program always starts from its initial state
- *migration in heterogeneous systems*:
 - only weak mobility is supported in common systems (recompile code, no run time information)
 - the virtual machines may be used: interpreters (scripts) or intermediate code (Java)

Fault Tolerance in Distributed Systems I.

- single machine systems
 - failures are all or nothing
 - OS crash, disk failures, etc.
- distributed systems: multiple independent nodes
 - partial failures are also possible (some nodes fail)
 - probability of failure grows with number of independent components (nodes) in the system
- **fault tolerance:** system should provide services despite faults
 - *transient faults*
 - *intermittent faults*
 - *permanent faults*

Fault Tolerance in Distributed Systems I.

Failure Types

Type of failure	Description
Crash failure	A server halts, but is working correctly until it halts
Omission failure <i>Receive omission</i> <i>Send omission</i>	A server fails to respond to incoming requests A server fails to receive incoming messages A server fails to send messages
Timing failure	A server's response lies outside the specified time interval
Response failure <i>Value failure</i> <i>State transition failure</i>	The server's response is incorrect The value of the response is wrong The server deviates from the correct flow of control
Arbitrary failure	A server may produce arbitrary responses at arbitrary times



Fault Tolerance in Distributed Systems II.

- *handling faulty processes*: through redundancy
- organize several processes into a group
 - all processes perform the same computation
 - all messages are sent to all the members of the particular group
 - majority needs to agree on results of a computation
 - ideally, multiple independent implementations of the application are desirable (to prevent identical bugs)
- use process groups to organize such processes

Fault Tolerance in Distributed Systems III.

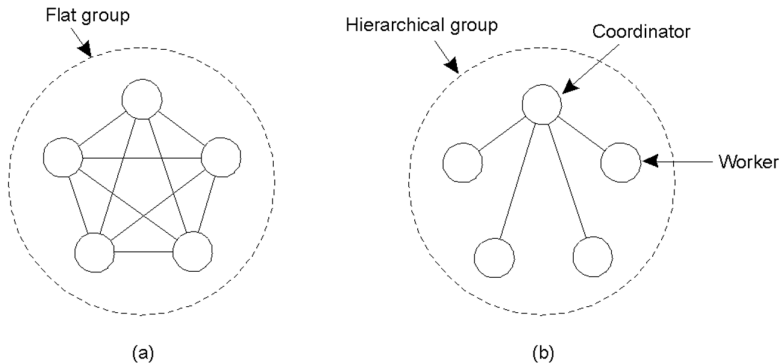


Figure: Flat Groups vs. Hierarchical Groups.



Fault Tolerance in DSs – Agreement in Faulty Systems

- How should processes agree on results of a computation?
- *K-fault tolerant*: system can survive k faults and yet function
 - assume processes fail silently
 - \Rightarrow need $(k + 1)$ redundancy to tolerant k faults
- *Byzantine failures*: processes run even if sick
 - produce erroneous, random or malicious replies
 - byzantine failures are most difficult to deal with

Fault Tolerance in DSs – Agreement in Faulty Systems

Byzantine Generals Problem:

- four generals lead their divisions of an army
 - the divisions camp on the mountains on the four sides of an enemy-occupied valley
- the divisions can only communicate via messengers
 - messengers are totally reliable, but may need an arbitrary amount of time to cross the valley
 - they may even be captured and never arrive
- if the actions taken by each division is not consistent with that of the others, the army will be defeated
- we need a scheme for the generals to agree on a common plan of action (attack or retreat)
 - even if some of the generals are traitors who will do anything to prevent loyal generals from reaching the agreement
- the problem is nontrivial even if messengers are totally reliable
 - with unreliable messengers, the problem is very complex
 - Fischer, Lynch, Paterson: in asynchronous systems, it is impossible to reach a consensus in a finite amount of time

Fault Tolerance in DSs – Agreement in Faulty Systems

Formal definition of the agreement problem in DSs:

- let's have a set of distributed processes with initial states $\in \{0, 1\}$
- *the goal*: all the processes have to agree on the same value
 - additional requirement: it must be possible to agree on both 0 or 1 states
- basic assumptions:
 - *system is asynchronous*
 - no bounds on processes' execution delays exist
 - no bounds on messages' delivery delay exist
 - there are no synchronized clocks
 - *no communication failures* – every process can communicate with its neighbors
 - *processes fail by crashing* – we do not consider byzantine failures

Fault Tolerance in DSs – Agreement in Faulty Systems

Formal definition of the agreement problem in DSs: cont'd.

■ *implications:*

- ⇒ there is no deterministic algorithm which resolves the consensus problem in an asynchronous system with processes, which may fail
 - because it is impossible to distinguish the cases:
 - a process does not react, because it has failed
 - a process does not react, because it is slow
 - practically overcome by establishing timeouts and by ignoring/killing too slow processes
 - timeouts used in so-called Failure Detectors (see later)



Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast

- *if there was a proper type of fault-tolerant broadcast, the agreement problem would be solvable*
- various types of broadcasts:
 - *reliable broadcast*
 - *FIFO broadcast*
 - *causal broadcast*
 - *atomic broadcast* – the broadcast, which would solve the agreement problem in asynchronous systems

Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast – Reliable Broadcast

- basic features:
 - Validity – if a correct process broadcasts m , then it eventually delivers m
 - Agreement – if a correct process delivers m , then all correct processes eventually deliver m
 - (Uniform) Integrity – m is delivered by a process at most once, and only if it was previously broadcasted
- possible to implement using send/receive primitives:
 - the process p sending the broadcast message marks the message by its identifier and sequence number
 - and sends it to all its neighbors
 - once a message is received:
 - if the message has not been previously received (based in sender's ID and sequence number), the message is delivered
 - if the particular process is not message's sender, it delivers it to all its neighbors

Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast – FIFO Broadcast

- the reliable broadcast cannot assure the messages' ordering
 - it is possible to receive a subsequent message (from the sender's view) before the previous one is received
- *FIFO broadcast*: the messages from a single sender have to be delivered in the same order as they were sent
- FIFO broadcast = Reliable broadcast + FIFO ordering
 - if a process p broadcasts a message m before it broadcasts a message m' , then no correct process delivers m' unless it has previously delivered m
 - $broadcast_p(m) \rightarrow broadcast_p(m') \Rightarrow deliver_q(m) \rightarrow deliver_q(m')$
- a simple extension of the reliable broadcast

Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast – Causal Broadcast

- the FIFO broadcast is still not sufficient: it is possible to receive a message from a third party, which is a reaction to a particular message before receiving that particular message
 - \Rightarrow *Causal broadcast*
- Causal broadcast = Reliable broadcast + causal ordering
 - if the broadcast of a message m happens before the broadcast of a message m' , then no correct process delivers m' unless it has previously delivered m
 - $broadcast_p(m) \rightarrow broadcast_q(m') \Rightarrow deliver_r(m) \rightarrow deliver_r(m')$
- can be implemented as an extension of the FIFO broadcast



Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast – Atomic Broadcast

- even the causal broadcast is still not sufficient: sometimes, it is necessary to guarantee the proper in-order delivery of all the replicas
 - two bank offices: one of them receives the information about adding an interest before adding a particular amount of money to the account, the second one receives these messages contrariwise
 - \Rightarrow inconsistency
 - \Rightarrow Atomic broadcast
- Atomic broadcast = Reliable broadcast + total ordering
 - if correct processes p and q both deliver messages m, m' , then p delivers m before m' if and only if q delivers m before m'
 - $deliver_p(m) \rightarrow deliver_p(m') \Rightarrow deliver_q(m) \rightarrow deliver_q(m')$
- does not exist in asynchronous systems

Fault Tolerance in DSs – Agreement in Faulty Systems

Fault-tolerant Broadcast – Timed Reliable Broadcast

- a way to practical solution
- introduces an upper limit (time), before which every message has to be delivered
- Timed Reliable broadcast = Reliable broadcast + timeliness
 - there is a known constant Δ such that if a message is broadcasted at real-time t , then no correct (any) process delivers m after real-time $t + \Delta$
- feasible in asynchronous systems
- A kind of “approximation” of atomic broadcast

Fault Tolerance in DSs – Agreement in Faulty Systems – Failure Detectors I.

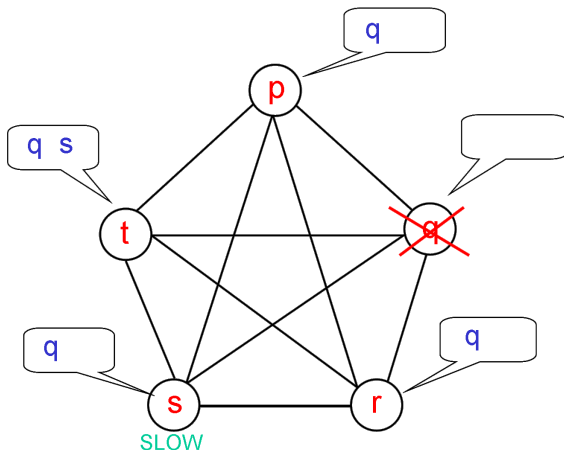
- impossibility of consensus caused by inability to detect slow process and a failed process
 - synchronous systems: let's use timeouts to determine whether a process has crashed
 - \Rightarrow Failure Detectors

Failure Detectors (FDs):

- a distributed oracle that provides hints about the operational status of processes (which processes had failed)
 - FDs communicate via atomic/time reliable broadcast
- every process maintains its own FD
 - and asks just it to determine, whether a process had failed
- however:
 - hints may be incorrect
 - FD may give different hints to different processes
 - FD may change its mind (over & over) about the operational status of a process



Fault Tolerance in DSs – Agreement in Faulty Systems – Failure Detectors II.



Fault Tolerance in DSs – Agreement in Faulty Systems

Perfect Failure Detector:

- properties:
 - Eventual Strong Completeness – eventually every process that has crashed is permanently suspected by all non-crashed processes
 - Eventual Strong Accuracy – no correct process is ever suspected
- hard to implement
- is perfect failure detection necessary for consensus? **No.**
 - \Rightarrow weaker Failure Detector

weaker Failure Detector:

- properties:
 - Strong Completeness – there is a time after which every faulty process is suspected by every correct process
 - Eventual Strong Accuracy – there is a time after which no correct process is suspected
- can be used to solve the consensus
 - this is the weakest FD that can be used to solve the consensus



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Service Oriented Architecture (SAO)

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- Fault Tolerance in Distributed Systems

Conclusion

Distributed Systems – Further Information

■ FI courses:

- PA150: Advanced Operating Systems Concepts (doc. Staudek)
- PA053: Distributed Systems and Middleware (doc. Tůma)
- PA039: Supercomputer Architecture and Intensive Computations (prof. Matyska)
- PA177: High Performance Computing (LSU, prof. Sterling)
- IV100: Parallel and distributed computations (doc. Královič)
- IB109: Design and Implementation of Parallel Systems (dr. Barnat)
- etc.

■ (Used) Literature:

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- Z. Tari and O. Bukhres. Fundamentals of Distributed Object Systems: The CORBA perspective. John Wiley & Sons, 2001.
- etc.