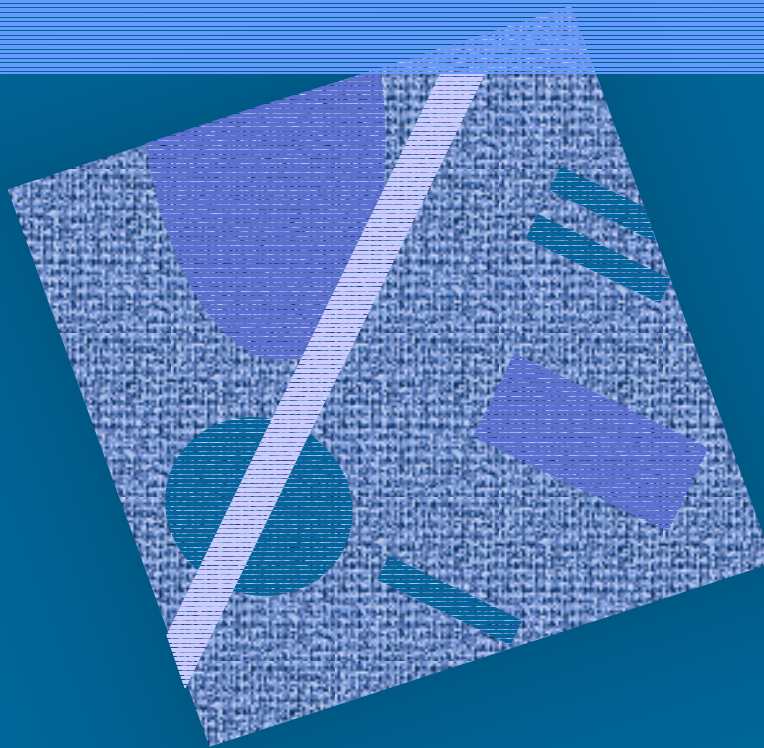


# Lecture 12

## Parameter Estimation Summary



HW/SW Monitors  
Events/Sampling  
Overhead  
Example  
Summary

# Order Processing Example

- Create model, Fig 9.1
- How to get parameter values?
  - OS
  - TP monitor
  - DBMS
  - Application

Customers	disk A
Inventory	disk A?
Orders	disk B
Paging	disk C

Job classes:	Order entry	init R=3.1 sec
	Order query	

# Measurement Process

- See Fig. 9.2
- Which data is needed?
- Where to place instrumentation?
- When to collect data?
- How long to collect data?
- How to account for measurement error?
  - load caused by the instrumentation?
  - does instrumentation have side effects?

# HW Monitors

- Collect raw data
- Wire monitors to physical units

Fig. 2.7 & Tbl 2.3 [Ferrari 78]

- HW tool characteristics (1978)

Tbl. 2.2 [Ferrari 78]

- Can be made to work with no (or very little) side effects or load
  - why?

# SW Monitors

- Modify SW to collect data  
Application? OS? DBMS?
- Use off-the-shelf monitoring system  
OS? DBMS? Other?
- Problem: runs in the same system that is being monitored
  - biased data. uses system resources
- Compare to HW monitors

Tbl 7.1 [Jain 91]



# Use Accounting Data for Parameter Estimation

- Good: it is there already (or is it?)
- Bad: it is built for accounting purposes
- Who used resources?
  - user id, program, project, account nr or class
- What resources were used?
  - CPU time, I/O op counts, NW use, mem use
- When were resources used?
  - user time only?
  - Granularity may be too coarse?

# Program Analyzers

- Runs with some programs
- Designed for one program

IBM DB2, IMS, CICS

- Generic analyzers
  - gprof
  - get nr. of transactions, CPU time, mpl

# Hybrid Monitors

- HW + SW

Special  
HW monitor

events

SW  
monitor

- Data filtering
  - Post run analysis
  - Can be very large monitoring system
    - as large as actual monitored system
- same system?  
different system?

[Wybraniec & Haban]





# Events or Sampling

- Event driven or time driven data gathering?
- Volume?
- Data collection method?
- Data storage? Where?
- Overhead?

# Events

- Predefined system events
  - “I/O request completed”
  - “process terminates”
- Create events
  - how much overhead?
- Log them into log files, or
- Send them to filtering and analysis

# Sampling

- Good: smaller volume
- Bad: may miss something
- Sampling rate? too high? too low?
- Collect state information
  - “CPU busy”, “disk busy”
  - “amount of memory at use”
- Assign data to threads/processes/programs?
  - “CPU busy 5 ms due to process P” ???

# Measurement Unit

- Module
- Subroutine
- Statement
  - language dependent?
- Machine instruction?

# Time Unit

- Real time?
- System time?
- External time?
- CPU time?
- Elapsed time?



# Instrumentation

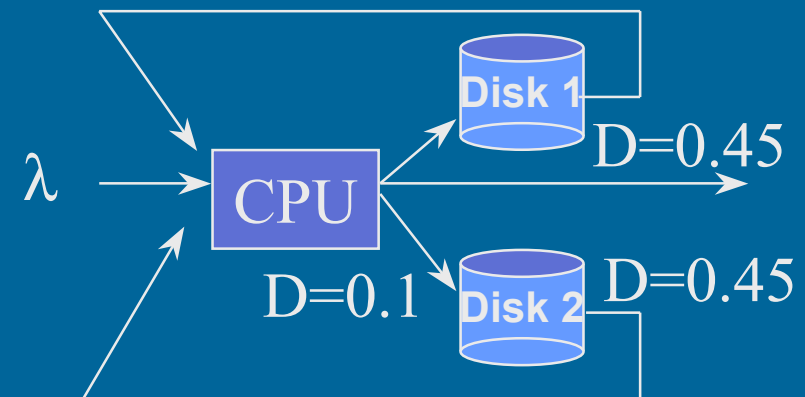
- When?
  - source code
  - compiler generated code
  - before linking, or at load time
  - run time
- How?
  - source code? Obj. code?
  - run time env.? OS? HW?

# Report

- Hierarchical
- Graphics
- What time is used?
  - inherited time
  - own time

# Parameter Values

- Job classes
  - clustering? factor analysis?
- $N_r$   $\lambda_r$   $Z_r$   $M_r$ 
  - external to system: plain measurements
- $D_{ir}$ 
  - internal: need internal data - e.g.,  $U_{ir}$
  - $D_{ir} = U_{ir} * T / C_{0r}$
  - sometimes get only  $U_i$



# Parameter Example

Fig. 9.3 [Men 94]

SW monitor

Accounting data

CPU demand



# Transaction Processing System Data

components not known

- Have  $U_{\text{total}} = U_{\text{cpu, OS}} + U_{\text{cpu, TP}} + U_{\text{cpu, prog}}$
- Need  $U_{\text{cpu, r}}$  for each transaction class  $r$

$$U_{\text{cpu, r}} = U_{\text{cpu}} * f_{\text{cpu, tp}} * f_{\text{tp, r}}$$

$$\frac{U_{\text{CPU, TP}}}{\sum_s U_{\text{CPU, s}}}$$

sampling  
system monitor

$$\frac{T_{\text{CPU, r}}^{\text{TP}}}{\sum_{s \in \text{TP}} T_{\text{CPU, s}}^{\text{TP}}}$$

analyzer for  
TP monitor



# TP Example <sup>(1)</sup>

- System with 3 classes:
  - Batch (B), Interactive (I)
  - TP (queries & updates)

for our TP model we need demands for these!

- System monitor: T=1800,  $U_{cpu}^{total} = 72\%$

- Accounting:  $U_{cpu,B}^{no OS} = 32\%$      $U_{cpu,I}^{no OS} = 10\%$      $U_{cpu,TP}^{no OS} = 28\%$

- TP analyzer:

- 1200 queries in 120 sec
- 400 updates in 140 sec

0.35 sec per update

- Now,

$$U_{cpu,U} = 0.72 * \frac{28}{70} * \frac{140}{260} = 15.5\%$$

$$D_{cpu,U} = U_{cpu,U} X_{0,U} = 15.5\% * \frac{1800}{400} = 0.70\text{sec}$$

$$D_{cpu,Q} = \left[ 0.72 * \frac{28}{70} * \frac{140}{260} \right] * \frac{1800}{1200} = 0.20\text{sec}$$

# Arrival Rate & MPL

- Assume flow balance
  - arrival count  $\cong$  completion count
  - $\lambda_r = C_{0r} / T$

- Average mpl:

$$\bar{N} = \lambda R = \frac{1}{T/n} * \frac{\sum_i e_i}{n} = \frac{\sum_i e_i}{T}$$

elapsed time  
for job i

nr of jobs

Fig 9.4 [Men 94]

# M and Z <sub>(2)</sub>

- M - Number of terminals session length  
(from accounting logs)

$$\overline{M} = \text{"session rate"} * \text{"session time"} = \frac{n}{T} * \frac{\sum S_i}{n} = \frac{\sum S_i}{T}$$

- Z:  $Z = \frac{M}{X} - R = \frac{M}{C_0 / T} - R$  average measured  
elapsed time  
in system

Example: T = 1 hour = 3600 sec

M=40    4900 interactive commands

2.5 sec aver resp time

$$Z = 40 / (4900/3600) - 2.5 = 29.4 - 2.5 = 26.9 \text{ sec.}$$

# CPU Demand: $D_{cpu,r}$

$$D_{cpu,r} = U_{cpu,r} * \frac{T}{C_{0,r}} = U_{cpu}^{Total} * f_{cpu,r} * \frac{T}{C_{0,r}}$$

where  $f_{cpu,r} = \frac{C_{0,r}}{\sum_s C_{0,s}} = f(jobs)$

or  $f_{cpu,r} = \frac{U_{cpu,r}^{mon}}{\sum_s U_{cpu,s}^{mon}} = f(time)$

or  $f_{cpu,r} = \frac{nrIO(r)}{\sum_s nrIO(s)} = f(nr\ IO's)$

or ...

# Capture Ratio <sup>(1)</sup>

- Capture ratio
  - class dependent!
- Example:

$$C_r = \frac{\text{measured } T_{cpu,r}}{\text{actual } T_{cpu,r}}$$

measure  
estimate  
calibrate  
guess

$$C_{batch} = 0.80$$

$$T = 7200 \text{ sec}$$

$$C_{term} = 0.60$$

$$U_{cpu}^{Total} = 64\%$$

$$T_{cpu,batch}^{Meas} = T_{cpu,term}^{Meas} = 1500 \text{ sec}$$

$$T_{cpu,batch} = \frac{1500}{0.80} = 1875 \text{ sec (actual!)}$$

$$T_{cpu,term} = \frac{1500}{0.60} = 2500 \text{ sec}$$

$$T_{cpu} = 4375 \text{ sec}$$

actual,  
not measured

$$U_{cpu,batch} = U_{cpu}^{Total} * f_{cpu,batch} = 0.64 * \frac{1875}{4375} = 0.27$$



# $D_{ir}$ for Disks

IO count

Mean service time

$$U_i^{Total} = \frac{IOC_i * MST_i}{T} \text{ (or from HW monitor)}$$

$$D_{ir} = \frac{U_i^{Total} * T}{C_{0,r}} * f_{ir} = \frac{IOC_i * MST_i}{C_{0,r}} * f_{ir}$$

where  $f_{ir} = \frac{IOC_{ir}}{IOC_{0,r}}$  (user data)

or  $f_{ir} = \frac{SWAPC_{ir}}{SWAPC_{0,r}}$  (swap data)

or  $f_{ir} = \frac{PageInC_{ir}}{PageInC_{0,r}}$  (paging data)

# Delay Node

- High disk utilization: include all disks in model
- Low (< 5%?) disk utilization: may use one aggregate **delay** node in model
  - no queueing likely ....

$$D_{ir} = \sum_{\text{disks } j} \frac{U_j^{Total} * T}{C_{0,r}} * f_{jr} = \frac{T}{C_{0,r}} \sum_{\text{disks } j} U_j^{Total} * f_{jr}$$

# Order Processing Example

Tbl 9.1 [Men 94]

- Want: Arrival rates, demands
- CPU demands
- File ops per application class
- File ops per disk
- Disk util. & demands per class
- $Z_r$ 's when  $R_r$ 's are know: use Little's Law

Tbl 9.3

Tbl 9.4

Tbl 9.5

Tbl 9.6

# Estimate $Z_r$ 's if $R_r$ unknown

- Use subsystem as open model
- Use  $\lambda_r = X_{0r}$
- Solve model, get  $R_r = (3.00, 0.98, 2.23)$
- Use those estimates to compute

$$Z_r = M_r / X_{0r} - R_r$$

$$Z_{OE} = 10/1.43 - 3.0 = 3.99 \quad (\text{vs } 3.75)$$

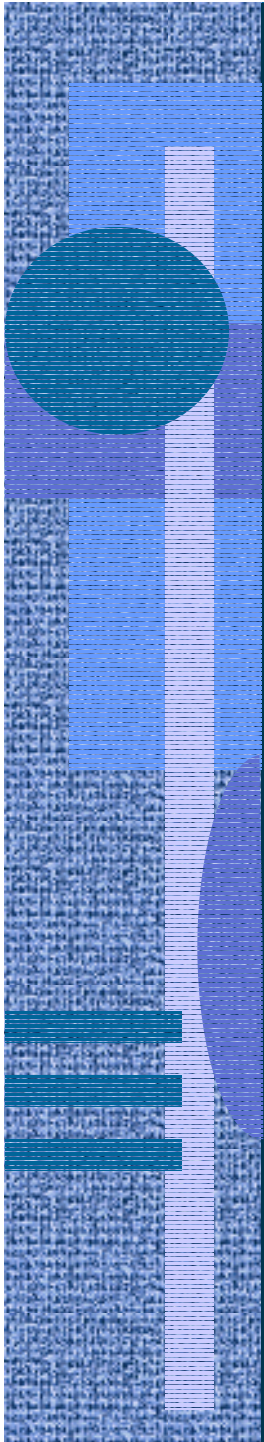
$$Z_{OI} = 5/0.83 - 0.98 = 5.04 \quad (\text{vs } 5.04)$$

$$Z_{Ot} = 5/0.31 - 2.23 = 13.9 \quad (\text{vs. } 13.9)$$

# Calibration

- Use parameters to build baseline model
- Use solver
  - e.g., PMVA fig.9.1.out
  - get  $R_{OE} = 2.5$  (not 3.25 as in Tbl 9.1)
- What now? Calibrate model!
  - scale (or add to) output to match data
  - adjust MPL or its distribution
  - add new class fig.9.1a.out
  - modify demands (at bottleneck) fig.9.1b.out
  - add ghost server for unaccounted for load



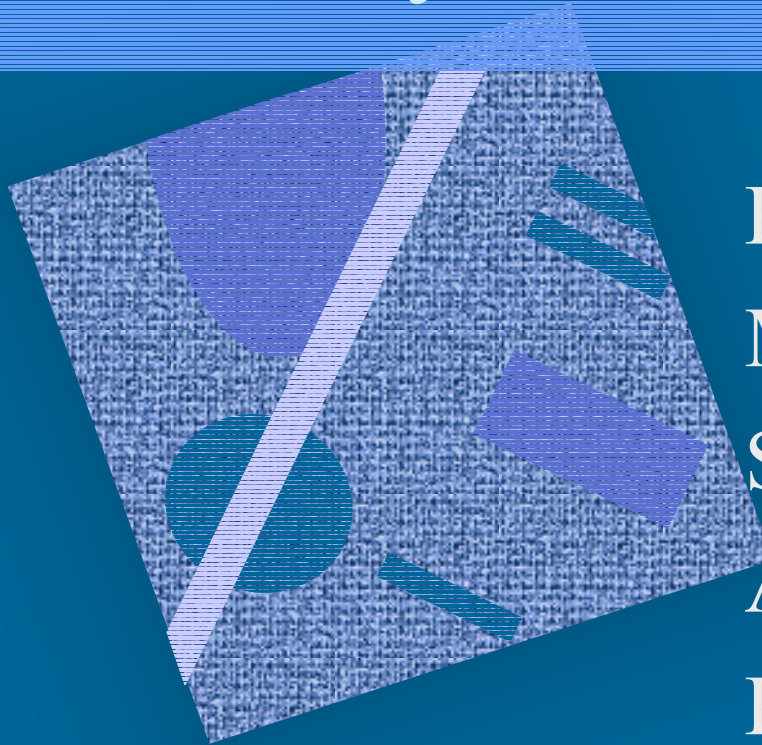


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# Summary



Probability Theory  
Modelling  
Solution Methods  
Approximations  
Parameter Estimation

# Probability Theory

- Distributions
- Sample
- Variance
- Confidence intervals



# Modeling

- Capacity planning process
- System model
- Workload model
- Capacity
- Performance

# Solution Methods

- M/M/1, Markov Chain
- Operational analysis
- Closed models
  - Convolution
  - MVA
  - Approximate MVA
- Open models

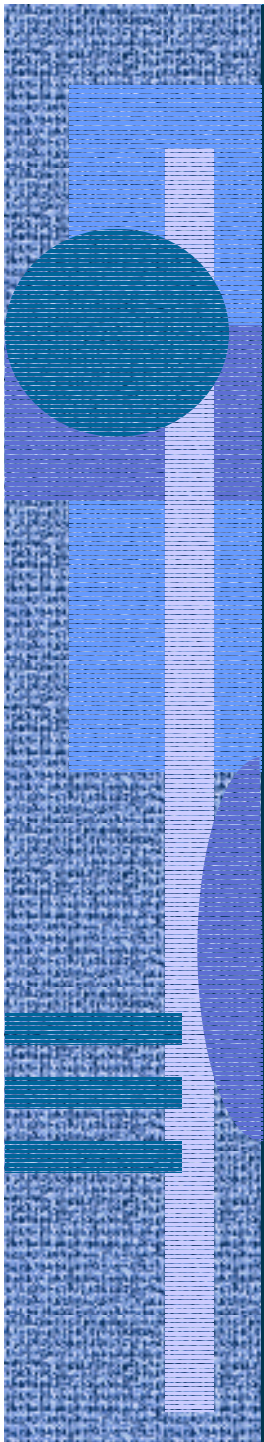


# Approximations

- Flow equivalent server
- Aggregate model
- Iterative solutions
  - multiple class memory
- Load concealment
- Shadow server

# Parameter Estimation

- Measurement process
- HW/SW monitors
  - hybrid monitors
- Accounting data
- Program analyzers
- Parameter values
  - combine data from various sources
  - fractions
  - overhead



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