



SPE Systems – Common Mistakes – PE tools

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http://disi.unitn.it/locigno/index.php/teaching-duties/spe





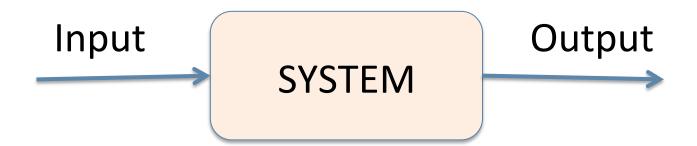
- What is a system?
- Common mistakes in PE
- PE tools
 - Measures
 - Simulations
 - Analysis



Systems



- A "system" is any physical or logical ensemble of which we want to measure some metrics
- Normally we whish to predict some "output" given some "input"







- We are given a "web server"
- Evaluate the service "responsiveness"
- Inputs = queries q_i
- Output = answer delays d_i







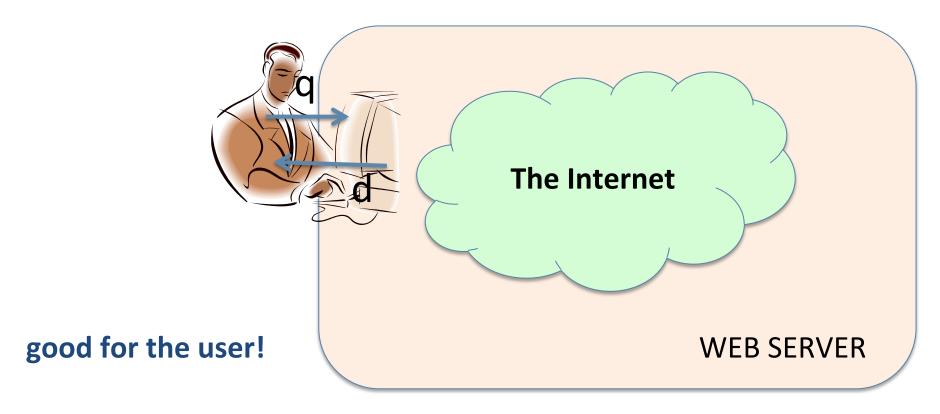
- How do we define a "web server", what is the goal of our evaluation
 - Hardware?
 - Software?
 - What are the "measurement points" of a web server?
- Is a delay a good measure of responsiveness?
- What are the queries q_i that are meaningful for the evaluation?
- What is the "correct" sequence of queries
 - In PE this is normally called the "workload" of the system under evaluation and its choice is of the utmost importance







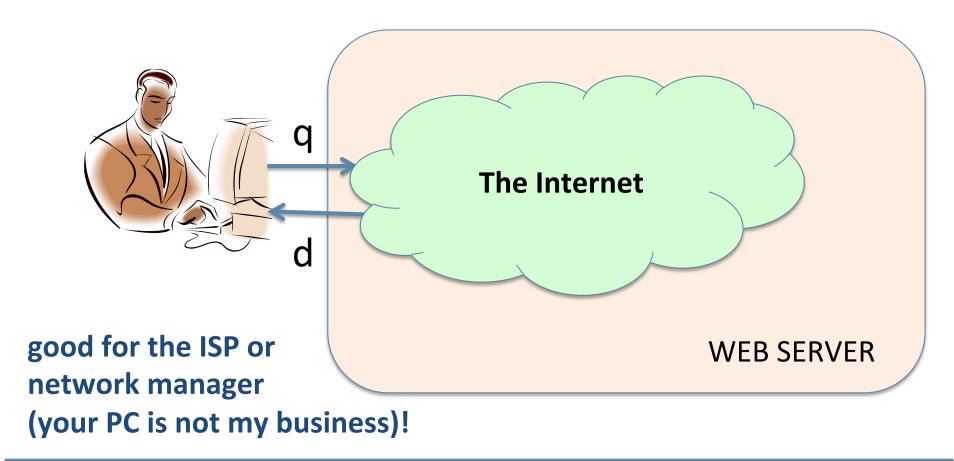
- What is a "web server"?
 - (Answer) or Model 1







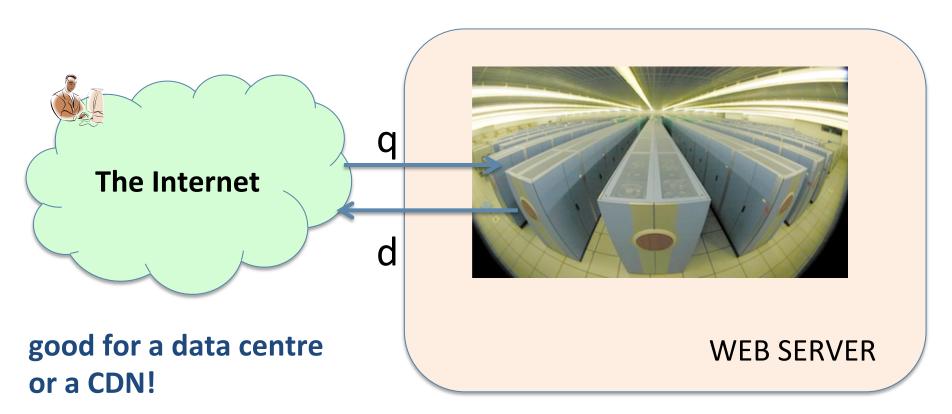
- What is a "web server"?
 - Model 1 bis







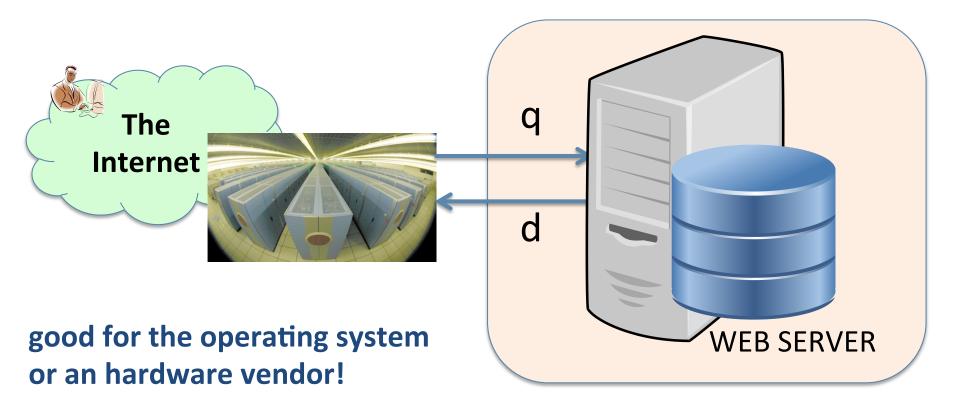
- What is a "web server"?
 - Model 2







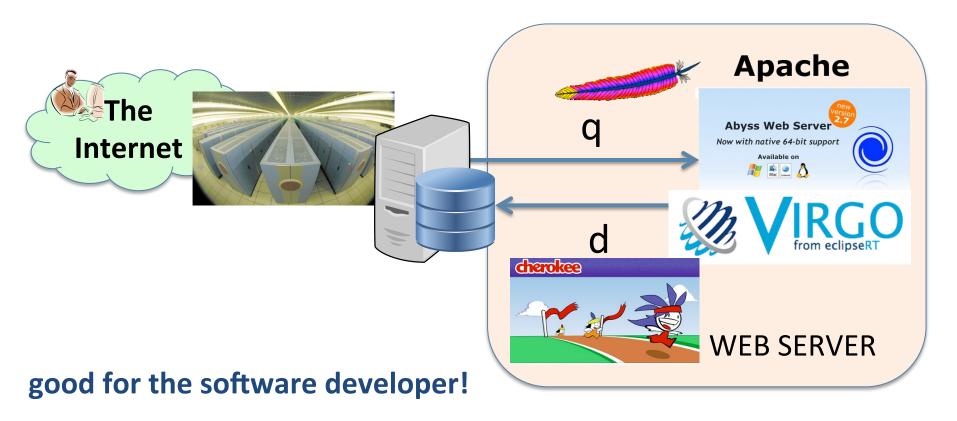
- What is a "web server"?
 - Model 3







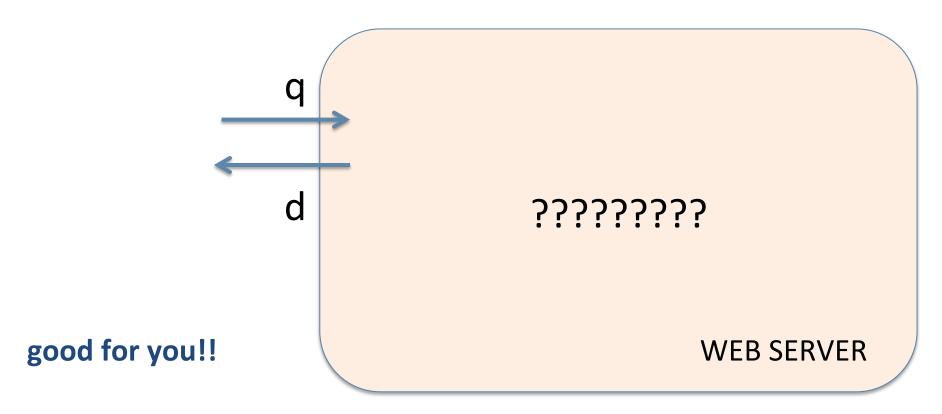
- What is a "web server"?
 - Model 4







- What is a "web server"?
 - Model 5







- Different models may require different evaluation tools or techniques
 - Model 1 (specially 1bis) can be easily measured if you seek a "local" answer, but measuring "on average" can simply be impossible
 - Models 2 and 3 can be also easily measured
 - but measures do not give "what-if" answers, so they are not good for design
 - Model 4 is very difficult to measure
- All models can be simulated, but we need to design the simulator correctly
- All models can are prone to analytical solutions, which are fast and easy, but can be very "rough"
 - Very useful to "dimension" systems





- The workload
 - $-q_1 \rightarrow d_1$; $q_2 \rightarrow d_2$; $q_3 \rightarrow d_3$; $q_4 \rightarrow d_4$; ...
 - Is it meaningful?
 - Is it representative of all real situations?
- Queries are not isolated and arrive with temporal relationships
 - Q = { $q_1, t_1; q_2, t_2; q_3, t_3; q_4, t_4; ...$ }; q_i in \mathbf{q} ; t_i in \mathbf{R}^+ or \mathbf{Z}
 - where q is the set of all possible queries
 - R⁺ and Z are real or natural numbers (including zero)
 depending on time being continuous or discrete





- The workload
- Does the sequence and temporal distribution of the workload influence results?
 - Q = { $q_1, t_1; q_2, t_2; q_3, t_3; q_4, t_4; ...$ } \rightarrow { $d_1; d_2; d_3; d_4; ...$ }
 - imply that
 - Q = { $q_3, t_1; q_2, t_2; q_1, t_3; q_4, t_4; ...$ } \rightarrow { $d_3; d_2; d_1; d_4; ...$ }
 - or not??
- The workload is often called also the "arrival process" to the system
- Notice that the PE output is normally NOT the output/ service of the system
 - A web server does not yield "delays" it returns web pages!!





- The output: responsiveness
 - What delay are we measuring?
 - Model 1: is the time needed for transmission and for the browser rendering meaningful?
 - Model 4: Database interactions of a CMS is our business?
 - Model 2: If q_i requires back-end computation (e.g. Hadoop), is that to be included or separated?
 - Model 5: Where do you put the probes to measure in-OS perfromance?



Common Conceptual Mistakes (CCM)



- Projects often fails due to PE analysis failures
 - The system designed does not perform as it should
 - The software project lags way behind schedule
 - Components have not been analysed
 - The development cycle has not been analysed
 - The nuclear reactor explodes when it should not (Chernobil, Fukushima, Three Mile Islands, Superphoenis, ...)
 - The system reliability has not been properly included in PE (TMI)
 - The system dependability has not been included at all (Fukushima)
- PE Analysis often fails due to initial conceptual mistakes



CCM: Analysis without a Model



- AKA lack of problem understanding
- Without a model we measure a black-box, not a system
- The measures taken may answer "what", but never "how" and "why"
- Artificial Intelligence & Machine Learning try to do some analysis on black-boxes
 - They have sophisticated stochastic models behind them
 - The results remains valid until the stochastic model behind is valid
 - e.g., the model assumes that the output distribution has a single mode,
 if not true the results are not valid



CCM: Search for the "True" Model



- "All Models are Wrong ... some of them are Useful"
 (Mark Tweine, Albert Einstain, George Box or somebody else much more famous than me)
- Alternative version 1: "All Models are Right ... most of them are Useless"

(Unknown, reported by T. Tarpey in a seminar in 2009)

• Alternative version 2: "All Predictions are correct ...

Unless they regard the Future"

(maybe Mark Twaine, if not, myself)



CCM: Biased Goals



- Frequent mistakes in Science & Engineering
 - I want to show that my system is better than his one
 - My boss will fire me if I show that his idea is bogus
- Always present in brochures & commercial datasheets
 - Those are ads, not science
- More often simply lack of pre-analysis
 - I have not given my problem enough time, so my global understanding is not enough to drive the PE process
- Pre-concepts prevent seeing the truth
 - I expect a red ball, and never see the green cube popping out



CCM: No Goals & Random Search



- I don't know what I need to measure/model
 - Measures taken cannot be interpreted
 - The model I make finally yields no meaningful outputs
 - e.g., results are always constant even changing parameters
- I have a model, but change parameters at random
 - Results taken in this way do not offer insights
- I have so many variables and parameters that the search space is practically infinite
 - Any amount of results will always be negligible w.r.t. to space to be analysed
 - Typical mistake in simulations



CCM: Wrong Model or Tool



- Selection of the correct tool (measures, simulations, analysis) is fundamental
 - e.g., we cannot "measure" a disaster, it's too late!
 - More later on
- The model must be tailored for the problem
 - Too much details makes it cumbersome & awkward to use
 - Too few details makes results unreliable



CCM: Uncheked Workload



- Outputs depends on the inputs
 - A different input than "thought" leads to wrong interpretations
- Biased workload
 - a skewer random generator
 - a workload that include rare events that never happen during the simulation/measure ... yet they are possible
- Changing one parameter of the workload (e.g., adding new web queries) changes also other parameters that are not controlled (e.g., interaction with data-bases)
 - Also changing variability I also change the mean ...



CCM: Impossibles & Outliers



- Outliers (results that are discarded as not meaningful) must always be checked
 - Instrument fault?
 - "Unlucky" Simulation?
 - Or a mistake in the model/implementation?
- Events that should not happen must not be ignored
 - Symptoms of a different system behaviour in measures
 - Symptoms of "eisenbugs" in software (simulations)
 - Symptoms of numerical problems in analytical solutions

— ...



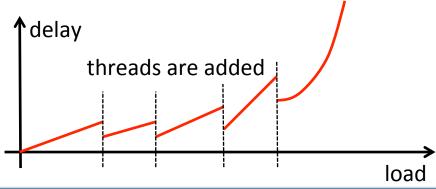
CCM: Overlooking Factors



- This is a "variation" of Impossibles & Outliers
- When building the model some important factor is left out
 - Result change when they should not (the "factor" has changed, but we ignore it)
 - We trust our results ... but they are wrong

 E.g., we model a web-server as a single process, while indeed it is multi-threaded with threads added based on

a performance threshold





CCM: Complexity



- If two models give the same output the simpler is preferred
- If a model has more than 100 parameters you can obtain any result you want ...
 - This becomes PE as "Parameters Engineering" not as "Performance Evaluation"
- An esoteric explanation is often appealing, but not necessarily the the correct one

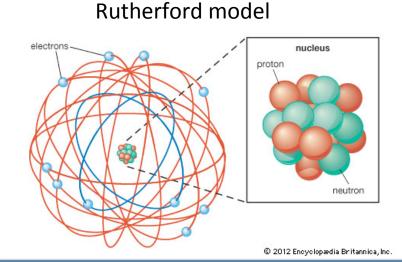


CCM: Lack of Model Checking



- The model is never put in discussion ... but sometimes it is wrong
- Rutherford vs Thomson Atom is the perfect example
 - The interpretation of scattering was simply impossible with Thomson model

Thomson model J.J. Thomson's picture of the atom "corpuscles" (electrons) "...sphere of uniform positive electrification .."





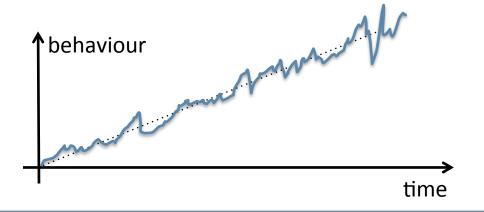
CCM: Stationarity Assumption

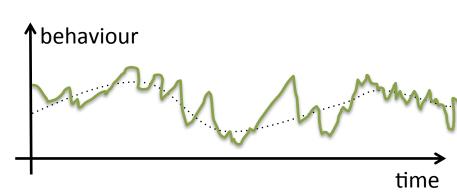


 Stationarity is the property of not changing the behaviour in time:

$$G(t) \approx G(t-t_0)$$
 for all t_0
where \approx indicates a stochastic equivalent behaviour

- Not all systems and phenomena are stationary
 - Interpreting non-stationary systems in light of a stationary model is wrong







CCM: Assumptions & Presentations



- Assumptions and approximations must be very clear (and declared!!)
 - A PE exercise can be OK if I know its goals and validity, but can be rejected entirely if I think it is about the real system
 - E.g., a multi-core CPU approximated as a single core one
- Results presentation is fundamental
 - Graphs are better than tables & numbers
 - Results must be presented with their confidence
 - The results selected for presentation must:
 - Give insight
 - Enable decisions





- The most traditional "scientific experiment"
- The system must be available, accessible and observable
- Workload generation is fundamental and often difficult
 - E.g., loading a 100Gbit/s optical network
- Changing system parameters can be difficult (impossible sometimes)
- Setting probes to take measures might be difficult (impossible sometimes)
 - E.g., adding a hardware counter to a CPU
- Measures must be repeated to achieve confidence

Measures





- They are often considered the most reliable PE form
 - But they do not give predictions!
 - You cannot "measure" disasters, you must avoid them!!
- Measures are intrinsically affected by noise
 - Can be additive, multiplicative, or even distortive
 - In any case it is represented by RVs
- Metrics themselves can be stochastic in nature
 - E.g., replication speed of cells; equilibriums in chemicals at high temperature; completion times of jobs in loaded systems; delays in the Internet; ...



Simulations



- A software that mimics the behaviour of the system
- Requires and abstract model
- State parameters of the system must be carefully selected

$$S(t) = \{s_1, s_2, ..., s_N\}$$

- Parameter ranges are important and must be checked
- t is inherently discrete: t in Z
- The evolution of the system is described as conditional random variables

$$S(t+1) = V | S(t)$$

where V is a vector of stationary R.V.



Simulations



- Workload generation is easier than in measures
- Parametric studies are normally easy
 - But they have a large computational cost
- A simulation "run" is a random walk in the state space of the model
- Simulation runs can be interpreted as Monte-Carlo solutions of the set of couples differential equations that describe the state evolution
- Multiple runs (or batch means techniques) are needed to evaluate the confidence of results



Analysis



- Finding the model is difficult
- Finding its solutions sometimes even more
- Monte-Carlo numerical integration is normally feasible
 - In this case the analysis becomes hybrid with simulations, but the confidence is normally easier to estimate and sometimes it does not require multiple solutions
- Parametric studies are very simple and effective
- Analytical models are often in the form of dynamic equations
 - i.e., coupled differential equations



Analysis: Example of a model



$$\begin{cases} dG(t)/dt = aF(t) & G(0) = G_0 \\ dF(t)/dt = -bG(t) & F(0) = F_0 \end{cases}$$

- A set of coupled differential equations
- This example can describe the evolution of a population G as a function of resources availability F, which in turn decreases as the population G increases
 - Is there an equilibrium?
 - Will G survive or get extinct



Analysis



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Validation



- Measures confirmed with simulations or analysis
 - Insight & Results interpretation
- Analysis confirmed by simulations
 - Be careful not to use the same model!
- Analysis confirmed & tuned with measures
- Simulations confirmed and tuned with measures
- Simulations confirmed by analysis

Never trust a single tool for PE (if possible)



Tool selection & Tradeoff



Criterion/Metric	Measure	Simulation	Analysis
Stage of design	Prototype – Operational	Any	Any – Very Early
Time required	Variable – Long	Variable	Small
Design complexity	High	Low – Medium	Variable
Accuracy	Variable	Variable	Variable
What-If evaluation	Difficult	Computationally expensive	Easy
Cost	Very High	Low to Medium	Low
Interpretation	Difficult	Medium	Easy
Marketing	Easy	Medium	Difficult (lack of understanding)