



Using Chip Simulation to Optimize Engine Control

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7th Conference on
Design of Experiments (DoE) in Engine Development
Berlin, 18.–19.06.2013

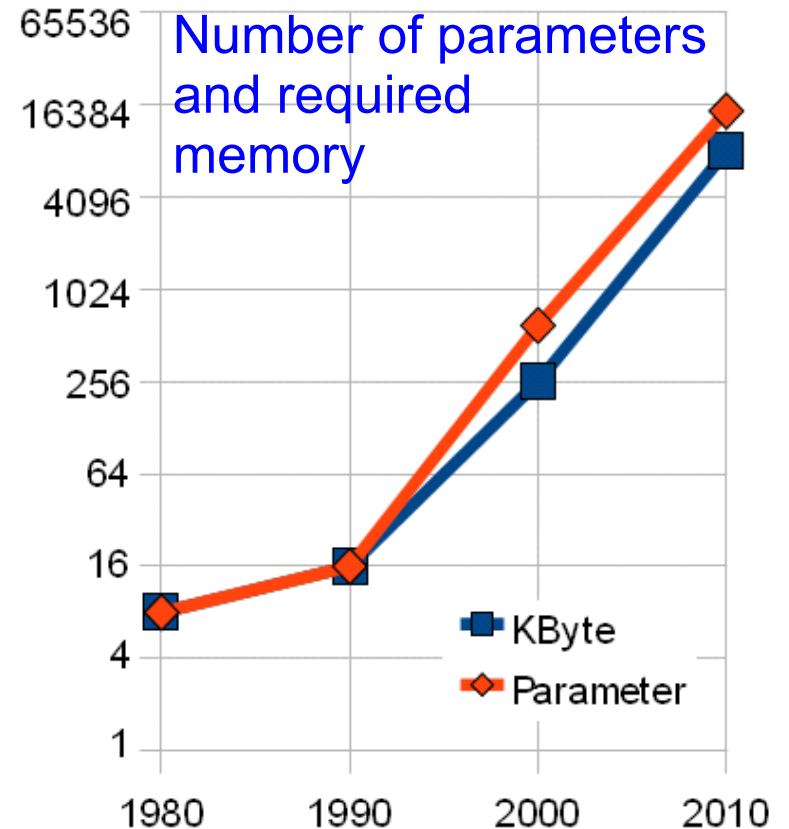
Using Chip Simulation to Optimize Engine Control

1. Motivation
2. Running ECU functions on PC via chip simulation
3. Coupling with least-squares optimization
4. Conclusion

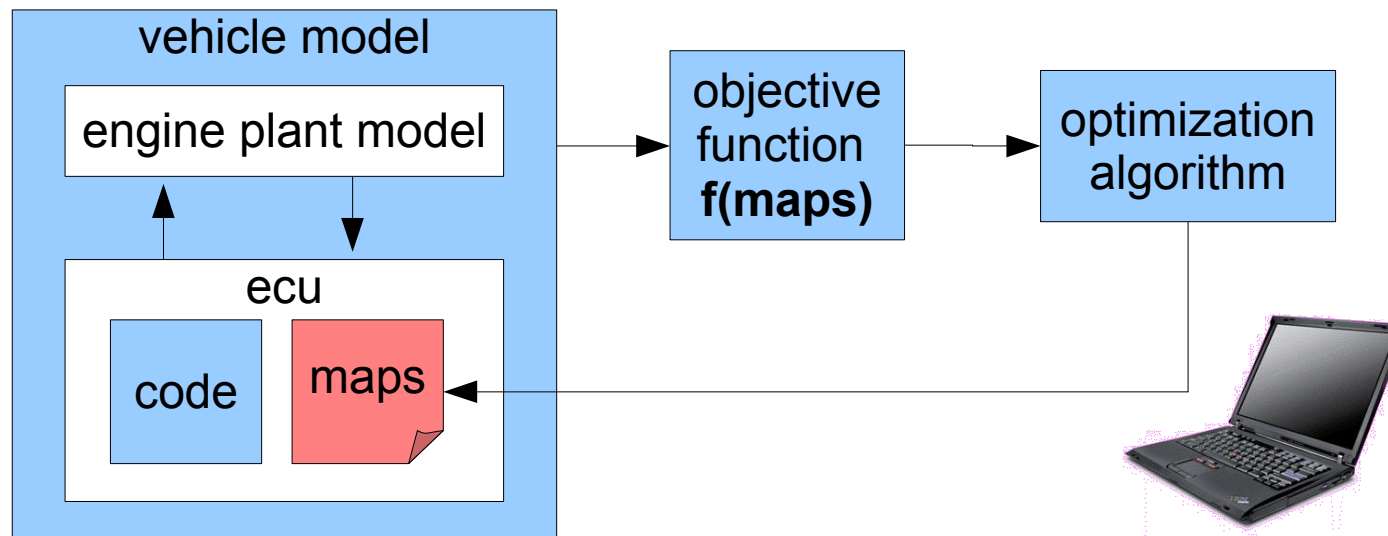
- number of engine control parameters doubles every few years
- budget for engine calibration does not

Idea

- increase degree of automation
- move calibration tasks from test rigs to PC and apply mathematical optimization



source: presentation of S. Ullmann (BMW)
5th Conference on DOE, 2009



ECU source (C, Ascet, or Simulink model) typically not available for OEM

Challenge: how to simulate the ECU on PC?

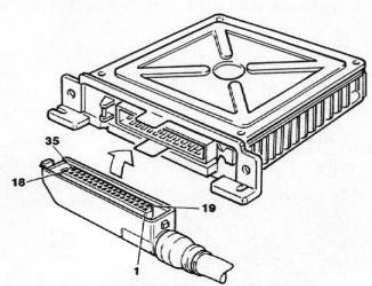
Options:

- **reverse engineer** the ECU function of interest, e. g. with Simulink
→ time consuming, error prone
- **simulate the hex** file of the ECU
→ less work, no modeling error

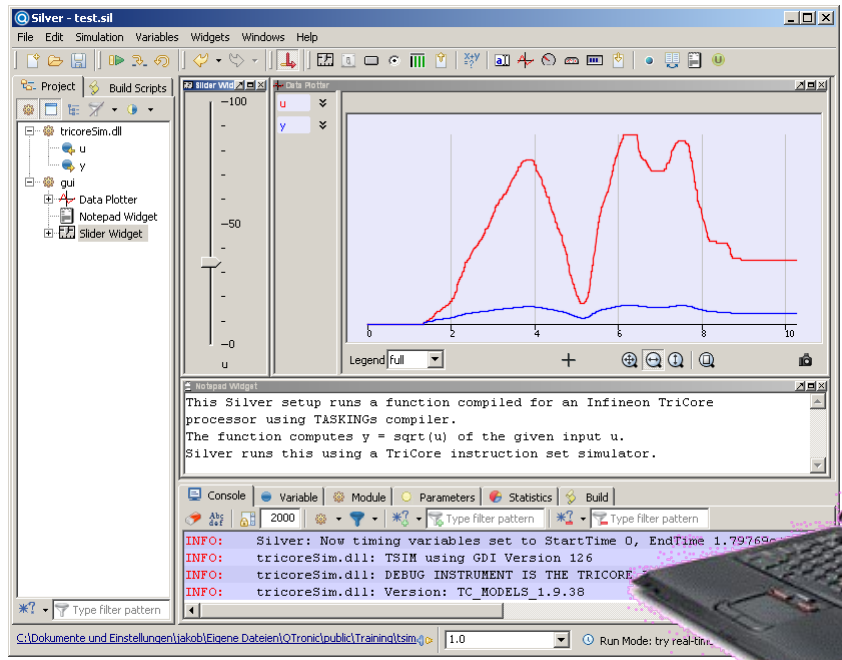
```
:020000040000FA  
:0200000480007A  
:200000003000000004FD00000000028000FD0080080200801C0200800507AF AF AF AF AF AF AF 62  
:20002000AF AF AF AF AF AF AF AF AF AF AF AF AF AF 02000000D35A48223004020000000080FFFC0080C2  
:20004000FECADEF AFEAFFECA0000000074FC008000100000310412007800008003020080C7  
:20006000FECADEF AFEAFFECA0000000000000000000000100000779781BD000000000000000000F  
:200080000000000000000000000000000000000000000000000000000000000000000000060  
:2000A0000000000000000000000000000000000000000000000000000000000000000000040  
000000000000000000000000000000000000000000000000000000000000000000000000020  
00000000000000000000000000000000000000000000000000000000000000000000000000000
```

.hex

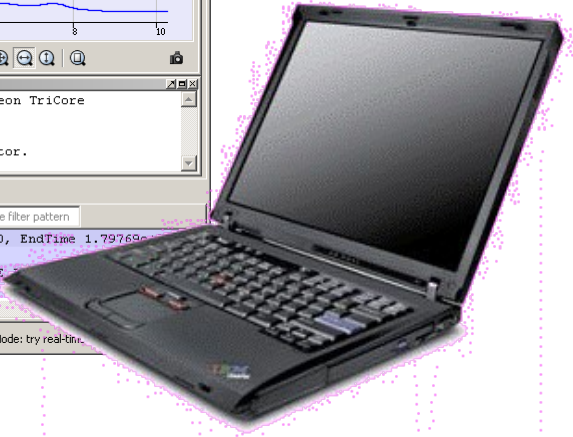
Silver Chip Simulator
for TriCore chip family



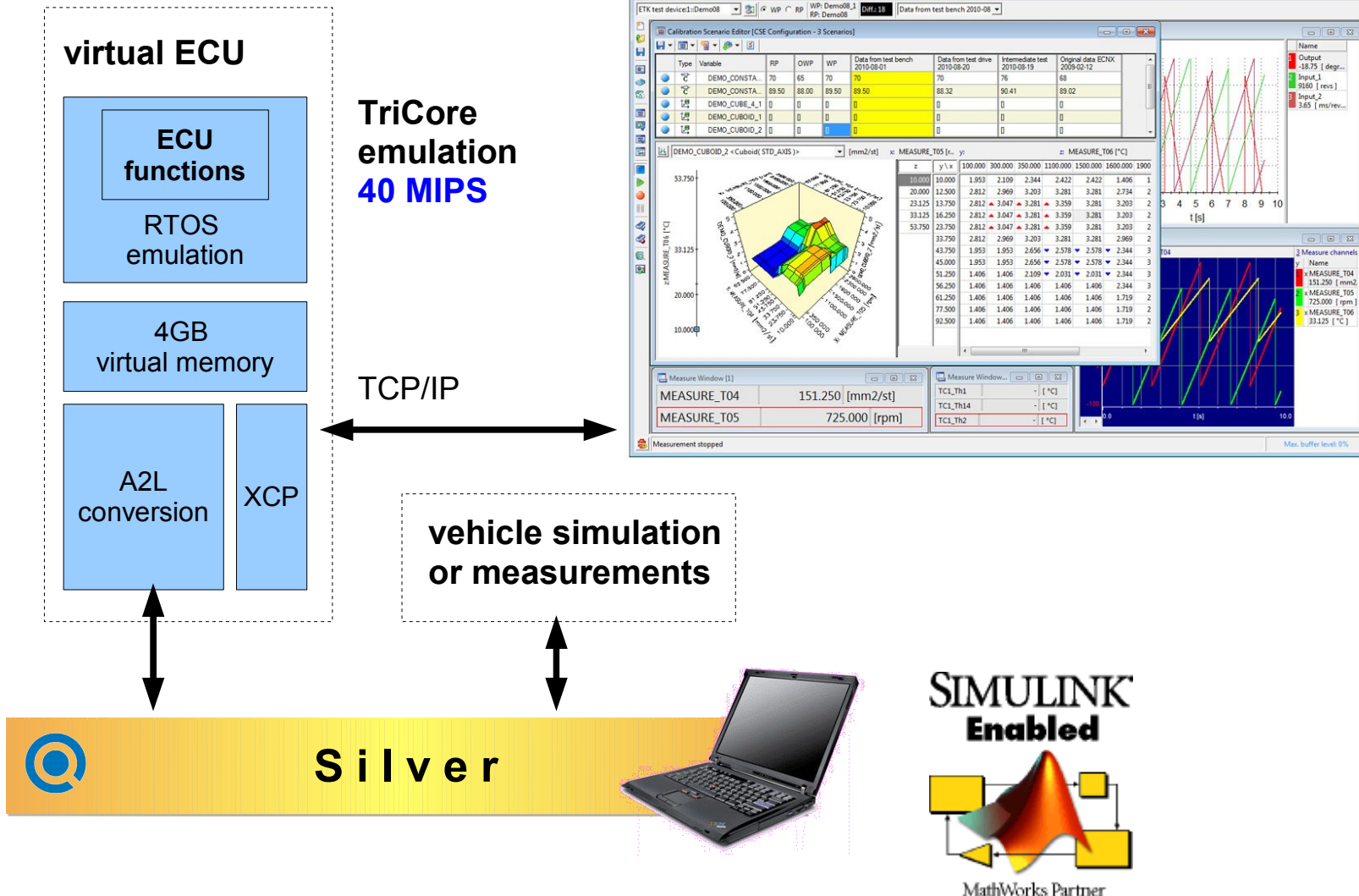
4 MB



- chip simulation runs on PC with about 40 MIPS
- selected function run e.g. 20 times faster than realtime
- simulation can be exported as SFunction



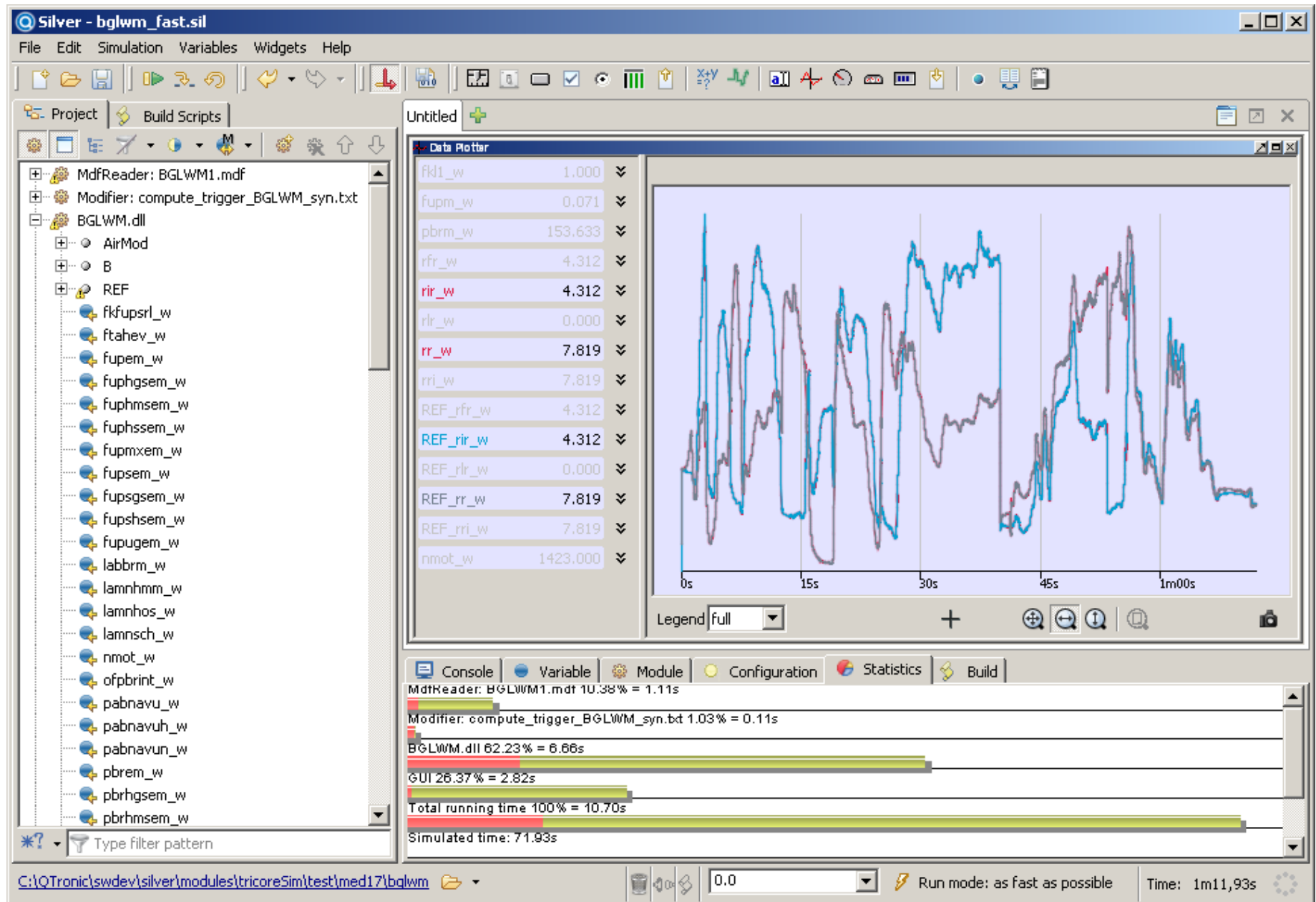
INCA or CANape for on-line calibration:
measure and tune running simulation



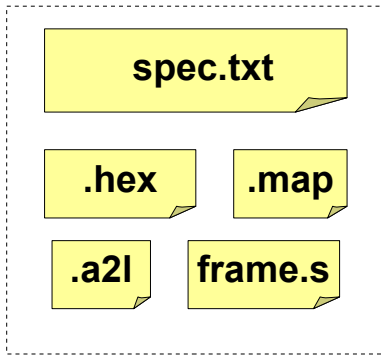
1. write spec.txt to specify what functions to run
2. step and debug the simulation in Silver debug mode
3. generate fast running SFunction or Silver module: runs without a2l and hex

```
01 # specification of sfunction or Silver module
02 hex_file(m12345.hex, TriCore_1.3.1)
03 a2l_file(m12345.a2l)
04 map_file(m12345.map)          # a TASKING or GNU map file
05 frame_file(frame.s)          # assembler code to emulate RTOS
06 frame_set(STEP_SIZE, 10)     # Silver step size in ms
07 frame_set(TEXT_START, 0xa0000000) # location of frame code
08
09 # functions to be simulated, in order of execution
10 task_initial(ABCDE_ini, 0)
11 task_initial(ABCDE_inisyn, 0)
12 task_triggered(ABCDE_syn, trigger_ABCDE_syn)
13 task_periodic(ABCDE_20ms, 20, 0)
14 task_periodic(ABCDE_200ms, 200, 0)
15
16 # interface of the generated sfunction or Silver module
17 a2l_function_inputs(ABCDE)
18 a2l_function_outputs(ABCDE)
19 a2l_function_parameters_defined(ABCDE)
```

Virtual ECU running in Silver: MED17



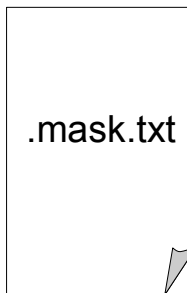
generated SFunction in MATLAB/Simulink



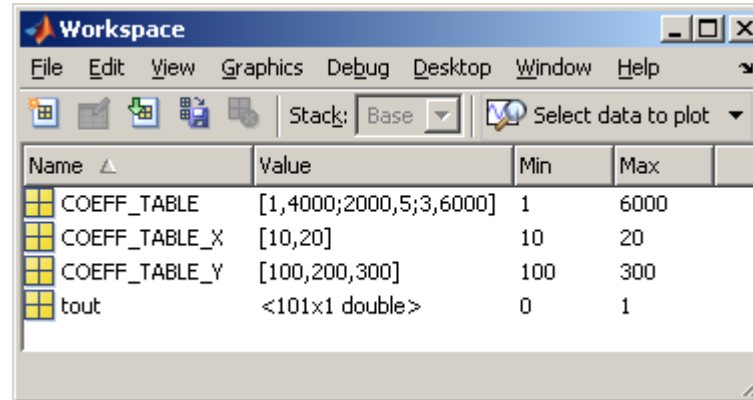
tcbuild



MATLAB/Simulink
S-function
40 MIPS

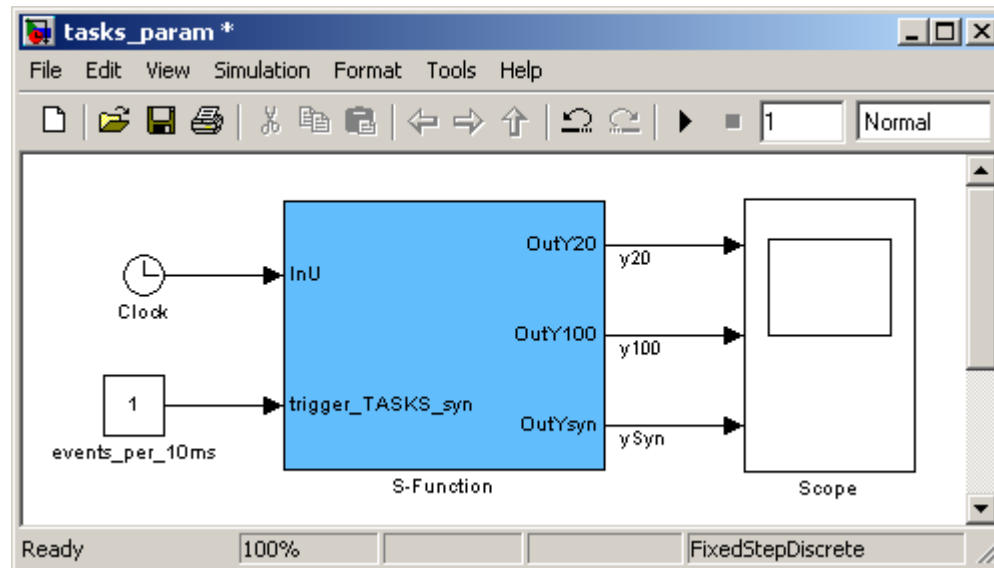


default values for
characteristics from
HEX file as m script,
mask for S-function
block and similar
Simulink snippets



Name	Value	Min	Max
COEFF_TABLE	[1,4000;2000,5;3,6000]	1	6000
COEFF_TABLE_X	[10,20]	10	20
COEFF_TABLE_Y	[100,200,300]	100	300
tout	<101x1 double>	0	1

characteristics turned into
MATLAB workspace variables
- read by S-function
- may be modified by script



Run complex function for a measured scenario, 3.5 minutes

target	execution time	MIPS
Silver in debug mode	919.15 sec	0.41
generated Silver module or MATLAB/Simulink SFunction	9.30 sec	40.80
MED17 with TC1797, 180 Mhz	210.00 sec	270

Limitations

- instruction accurate, but not cycle accurate
- based on TriCore specification: 'silicon bugs' are not simulated
- PCP, CAN controllers and other on chip devices not modeled

Advantages

- no real-time requirement: simulate faster or slower than real-time
- 4 GB virtual memory available in virtual ECU
- zero-execution time model: simulated task runs infinitely fast
hence: deterministic simulation without interrupts: easy to analyze

Engine controller contains steady-state model of the engine

Objective

Tune parameters of the engine model such that it fits given measurements

Least-squares optimization

Minimize goal function

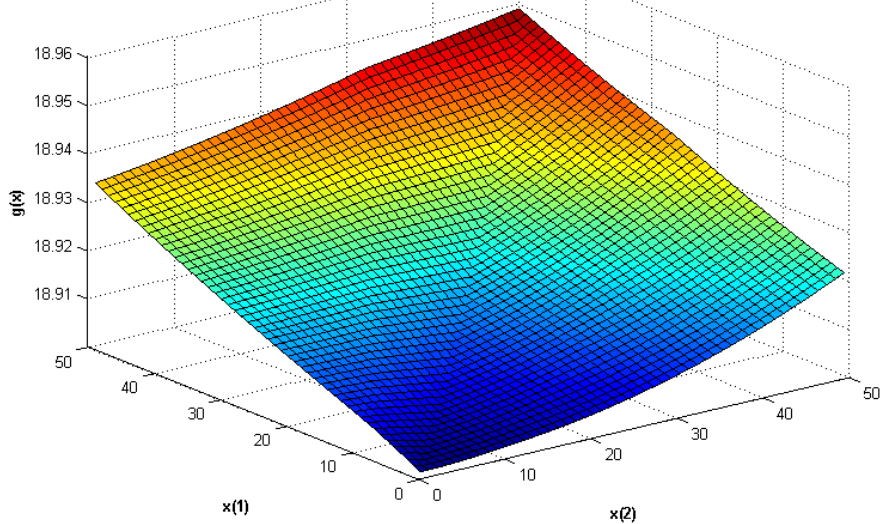
$$g(x) = \sum_{i=1}^m f_i^2(x)$$

where

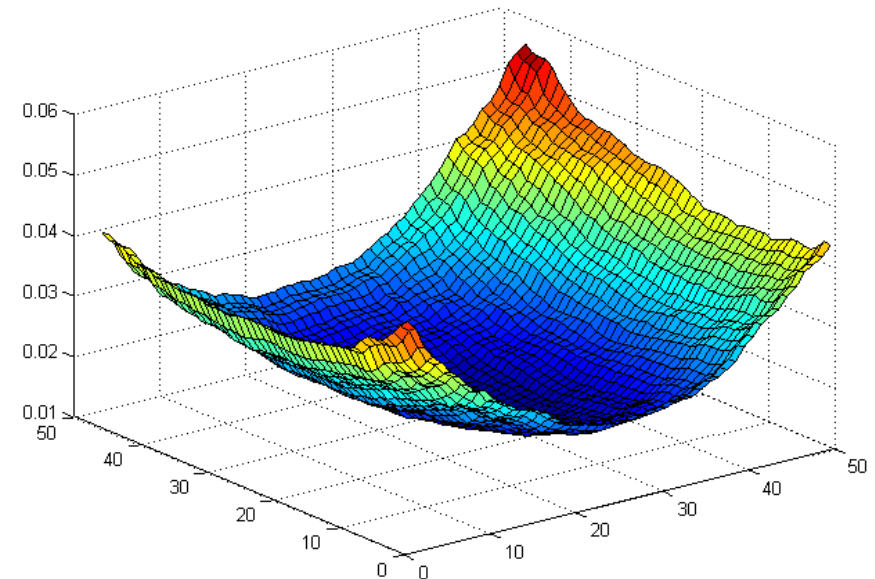
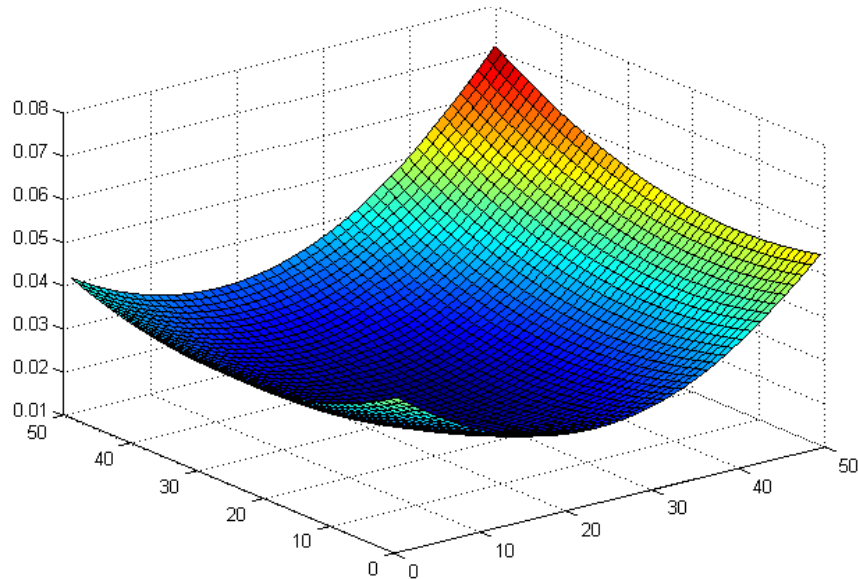
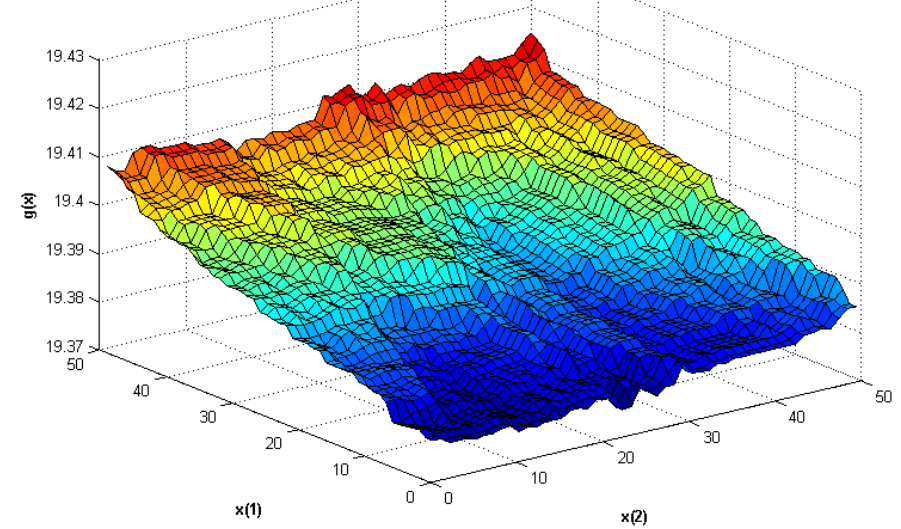
- x is a vector of n real parameters
- $f_i(x) = \text{model}(x, t_i) - \text{measurement}(t_i)$

A problem with chip simulation

goal function: Simulink model

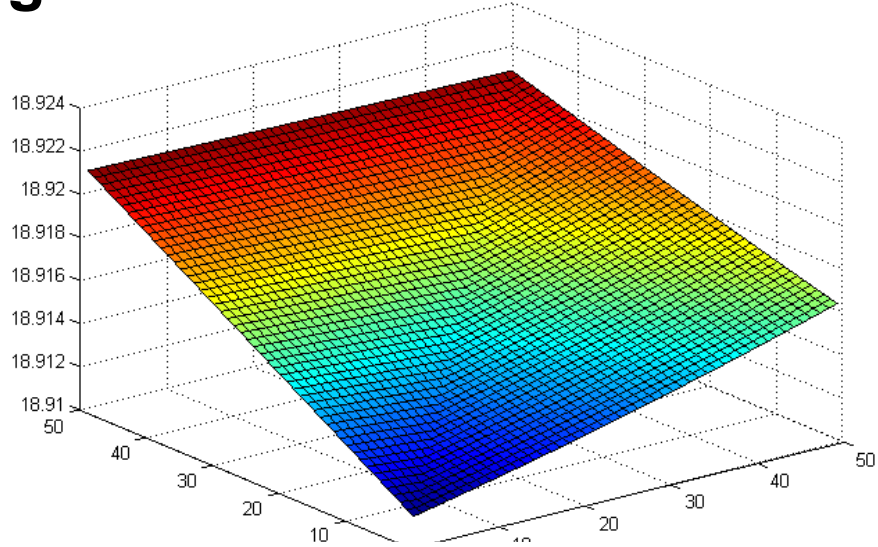


chip simulation

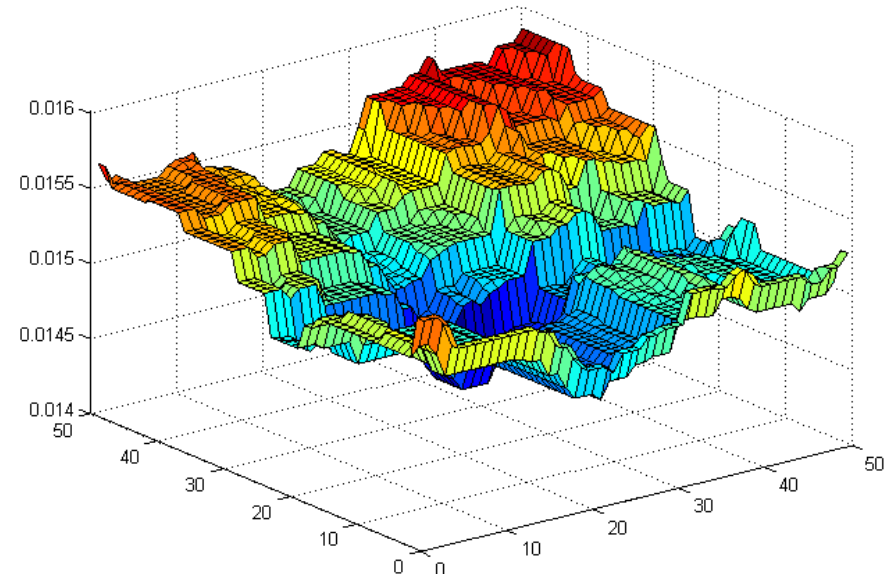
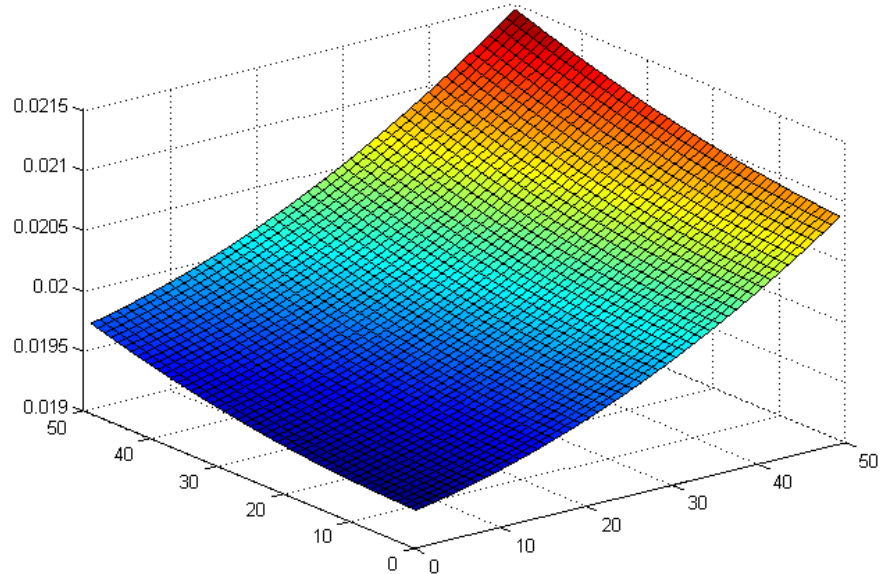
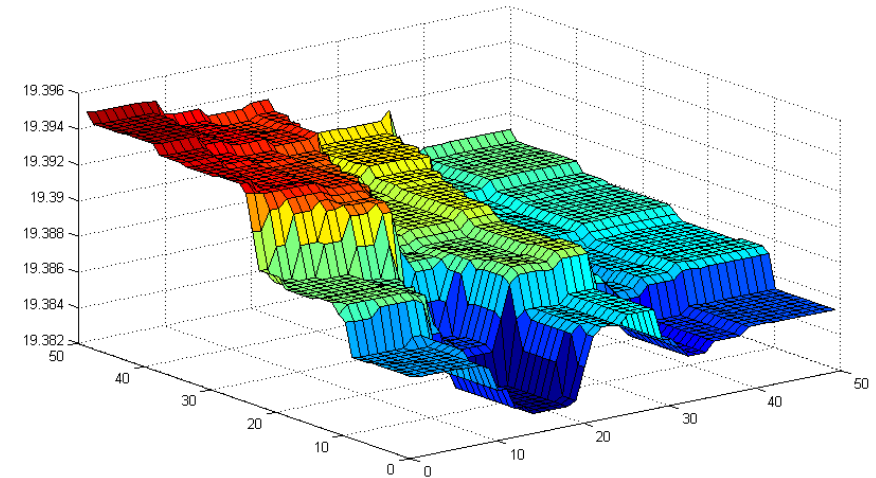


A problem with chip simulation

goal function: Simulink model



chip simulation

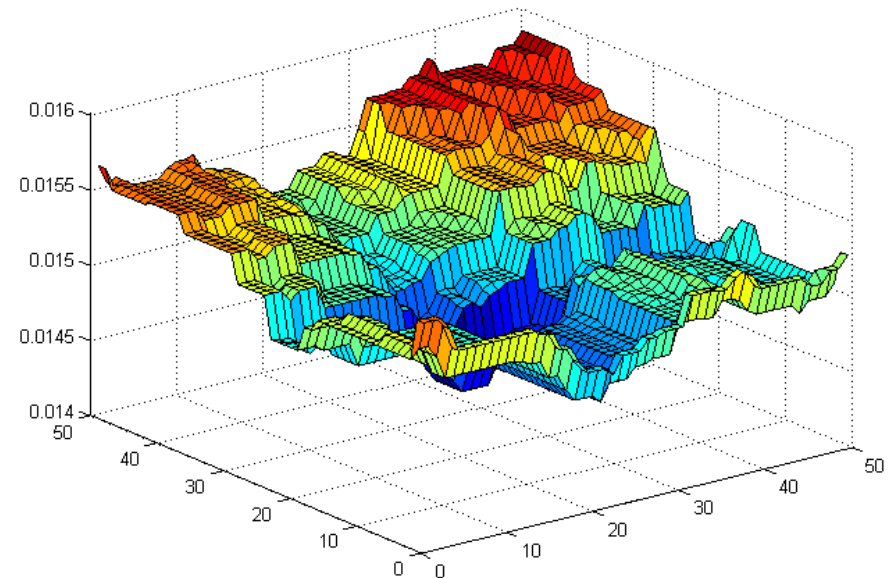
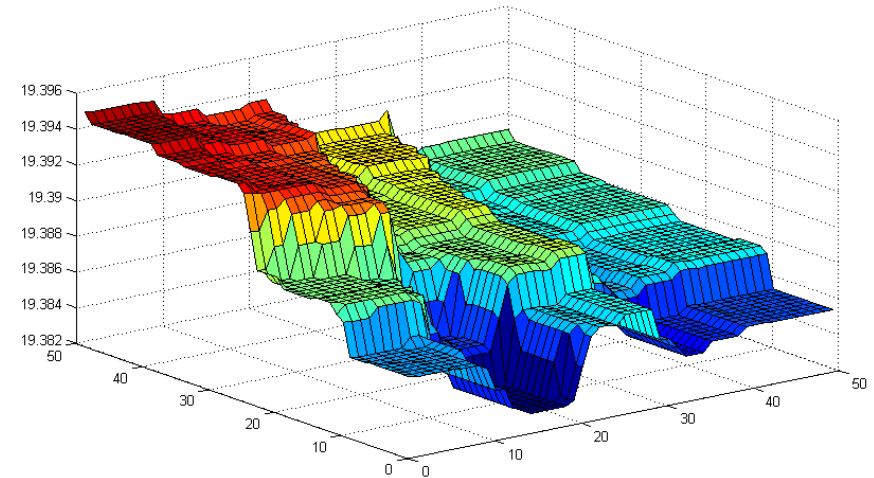


A problem with chip simulation

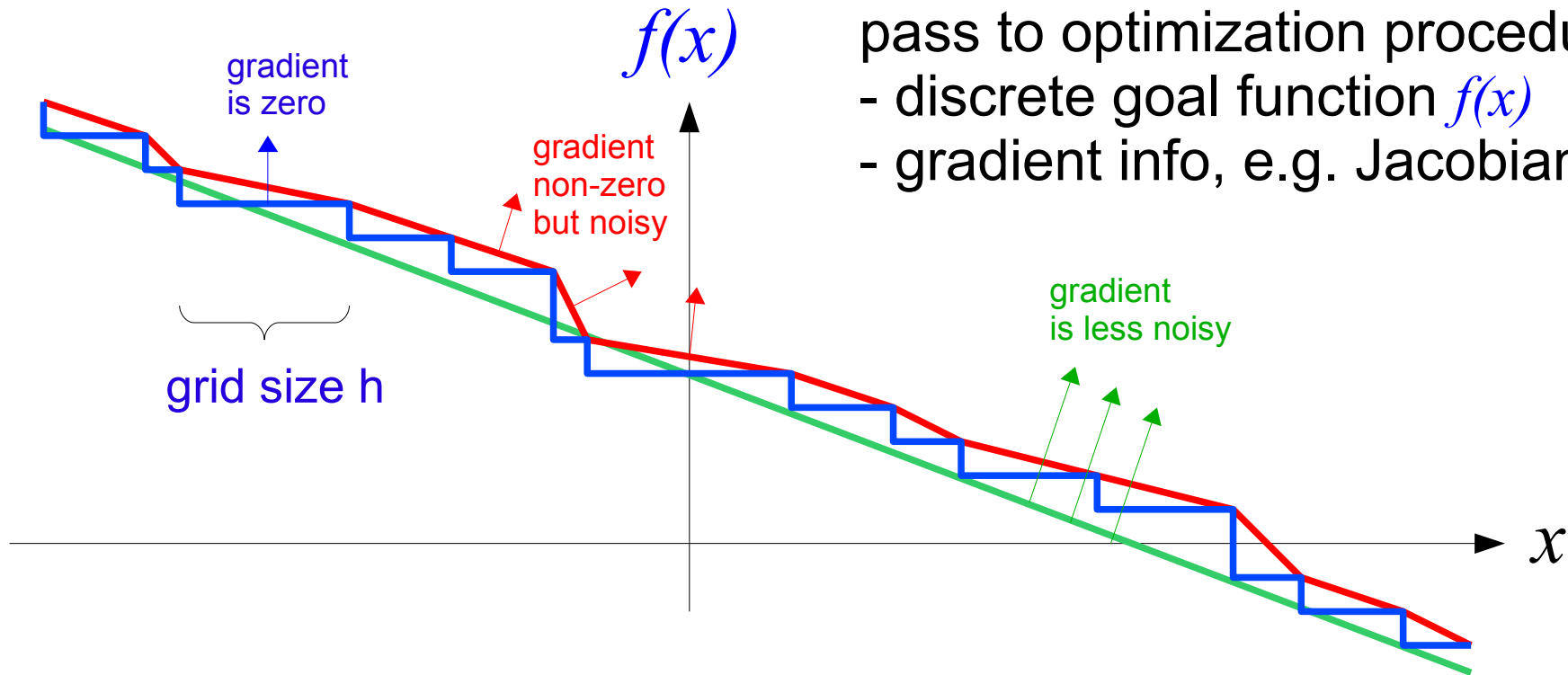
Optimization methods often require **gradients** to guide search

Engine control often implemented using **fixed-point integer** code

- gradients of the goal function are zero (or undefined)
- no guidance
- optimization terminates early at local optimum



Idea 1: construct a smooth goal function



- pass to optimization procedure
- discrete goal function $f(x)$
 - gradient info, e.g. Jacobian J_{ij}

$f(x)$ goal function implemented using chip simulation: zero gradient

$f(x)$ use current grid size h to compute gradient

$$\frac{f(x+h) - f(x)}{h}$$

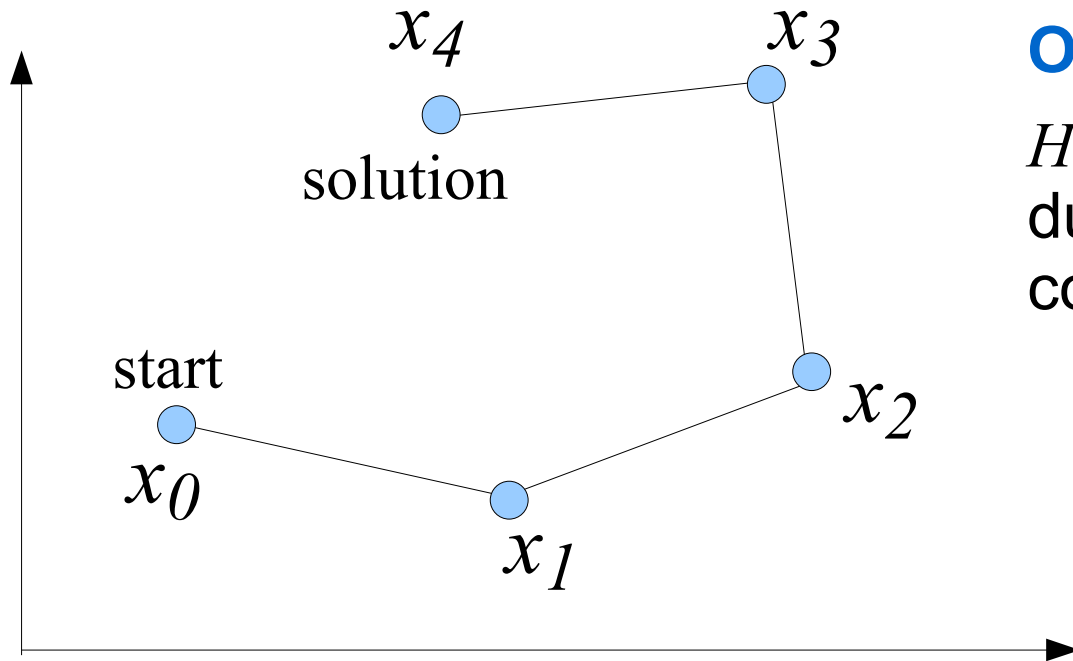
$f(x)$ less noise: use $10h$ to compute the gradient

Idea 2: pre-compute grid sizes

m time points, n parameter

→ $m \times n$ matrix H_{ij} of grid sizes

→ must be computed at each step x_0, x_1, x_2, \dots expensive!



Observation

H_{ij} does not change much during the solution process: compute only for x_0 and reuse

Idea 3: Stochastic model of grid sizes

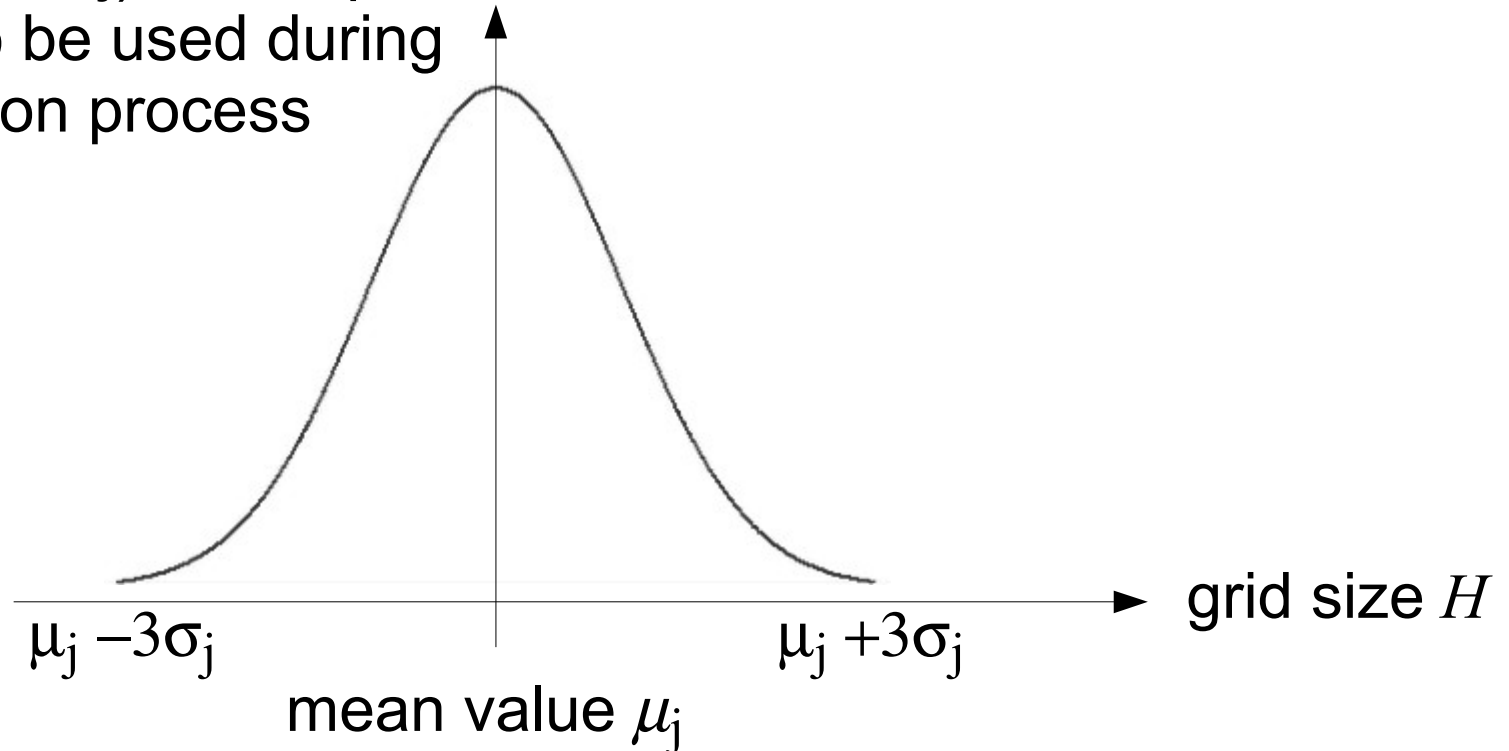
For large problems, do not compute all elements of matrix H_{ij}

Use **stochastic model**: for parameter x_j

- compute H_{ij} for x_0 and some (not all) time points t_i

- estimate **average** μ_j and **standard deviation** σ_j

- use $h_j = 10 (\mu_j + 3\sigma_j)$ to compute gradient of x_j to be used during the entire solution process



Example: Tune engine model used by engine controller

- $m = 202$ measurement time points
- $n = 20$ parameters
- solver: **lsqnonlin** from MATLAB optimization toolbox
- goal function implemented using chip simulation
 - gradient info passed using option FinDiffRelStep
 - stochastic model of grid sizes
- performance validated against hand-coded smooth Simulink model
 - very **similar solutions** found
 - similar number of function evaluations
 - **factor 2 slower** with chip simulation, to compute grid sizes

Using Chip Simulation to Optimize Engine Control

- chip simulation can be used to port ECU functions to PC
- the resulting model
 - runs much faster than real time
 - can be coupled with optimization procedures to automate engine calibration
- derivative-free optimization:
no problem
- otherwise:
compute gradient as
finite difference with
controlled step size

