



# Recursive Trust-Region Method for Unconstrained Minimization: Convergence properties and experience

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# Unconstrained optimization

The unconstrained nonlinear programming problem:

$$\text{minimize } f(x)$$

for  $x \in \mathbb{R}^n$ ,  $f$  smooth.

Many applications:

- variational problems
- surface design

(Generalizations now under study)

## Plan

→ *Introduction*

- Problem
- Algorithm
- Taylor iterations
- Iter. structure
- Initial point
- Test problems
- Some results
- Theory
- Perspectives



# Hierarchy of problem descriptions

Can we use a structure of the form:

Finest problem description

Restriction  $\downarrow R$

$P \uparrow$  Prolongation

Fine problem description

Restriction  $\downarrow R$

$P \uparrow$  Prolongation

...

Restriction  $\downarrow R$

$P \uparrow$  Prolongation

Coarse problem description

Restriction  $\downarrow R$

$P \uparrow$  Prolongation

Coarsest problem description

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# Structured model choice

Consider minimizing at **topmost** (finest) level.  
At each iteration, choose the model as

- a local Taylor expansion (classical)  
→ **Taylor iteration**
- the immediately coarser problem description  
→ **recursive iteration:**

compute fine  $g$  (and  $H$ )

step and trial point

**Restriction**  $\downarrow R$

$P \uparrow$  **Prolongation**

minimize the *coarse* model within the *fine* TR



# A recursive multi-scale algorithm

Until **convergence**:

- Choose either a Taylor or recursive model
  - Taylor model: compute a Taylor step
  - Recursive: **apply the Algo recursively**
- Evaluate change in the objective function
- If achieved reduction  $\approx$  predicted reduction,
  - accept trial point as new iterate
  - (possibly) enlarge the trust region
- else
  - reject the trial point
  - shrink the trust region
- Impose: **current TR  $\subseteq$  upper level TR**

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# (Simple) Taylor iterations

Which solver for the (approximate) solution of the (same level) **trust-region subproblem**?

Simple answer:

- for **low(est) level(s)** (small dimension):  
the exact Moré-Sorensen method
- for **higher levels** (high dimension):  
a truncated conjugate gradient  
(Steihaug-Toint or GLTR)

But...



# Multigrid technology...

Multigrid techniques for PDE problems indicates:

- the **high-frequency** components of **residual** only visible in **fine mesh** (high levels)
- need **two** different methods:
  - reduce **high frequency** components on the **fine mesh**

Smoothing

- reduce **low frequency** components on the **coarse mesh**

Damping

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# ... adapted to optimization

In unconstrained optimization,

residual  $\rightarrow$  gradient

- gradient **smoothing**:
  - TCG not very efficient!
  - adapt Gauss-Seidel smoothing
    - $\rightarrow$  **cyclic coordinate search**
    - (on Taylor's model)
- low frequency **damping**:  
full solution (MS) in low dimension

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# Structure of the recursive iterations

Decision to **stop solving the lower-level subproblem** based on

- subproblem **criticality**  $\rightarrow$  free form (gradient accuracy + TR constraint activity)
- fixed form **cycles** (possibly truncated)
  - **V** cycles
  - **W** cycles
  - **$W_q$**  cycles ( $q > 2$ )

At least one successful iteration per level

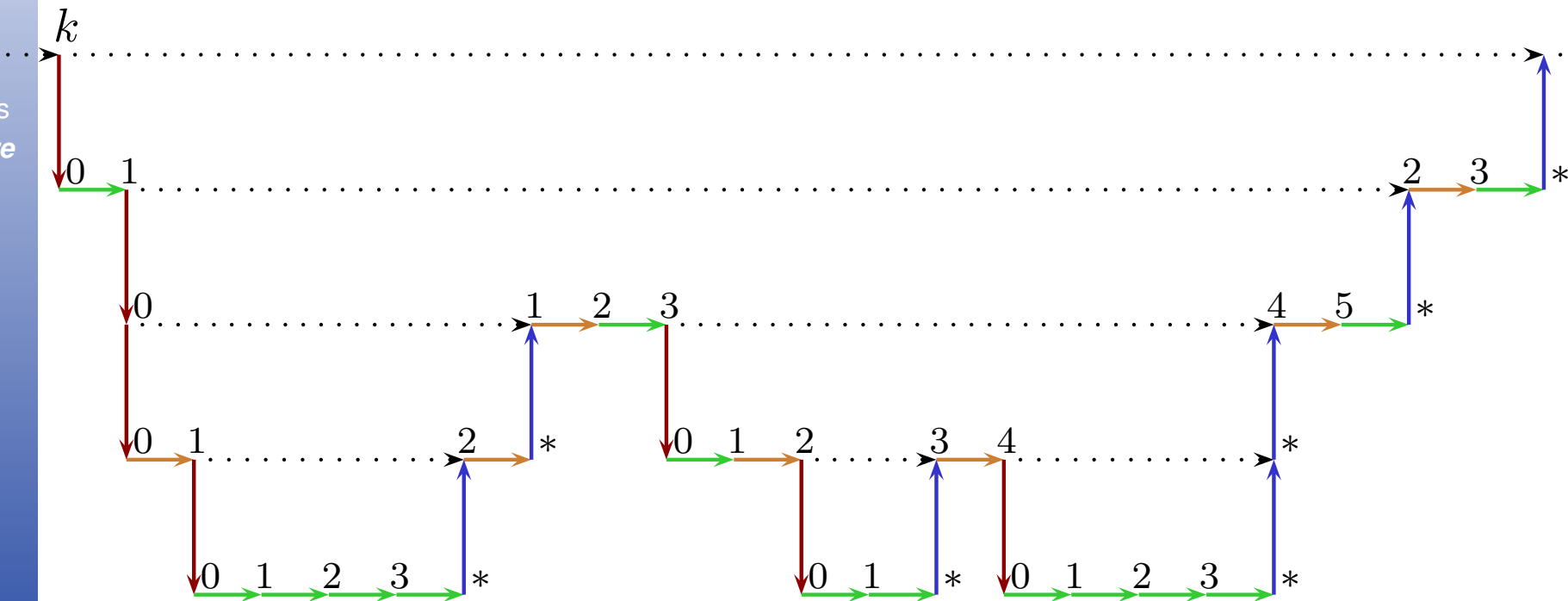
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# Free form iterations

The “iterates view” for the same example:



Prolongation

Restriction

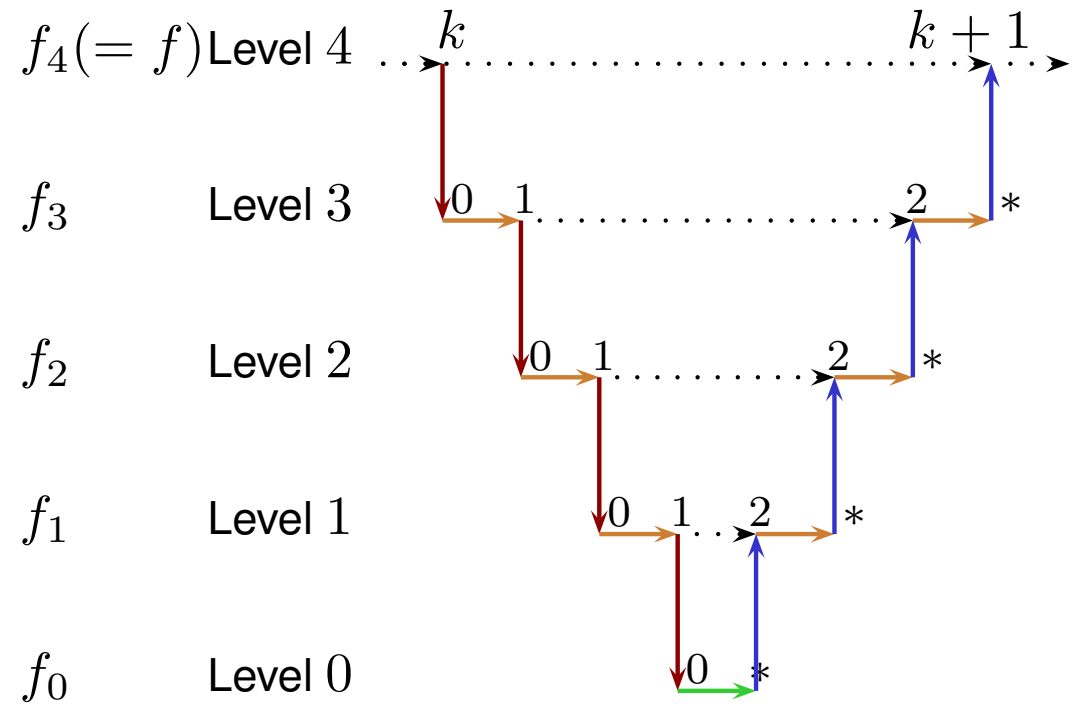
Damping

Smoothing



# V cycles

The “iterates view” for a **V cycle** recursion (5 levels, all iterations successful):



Prolongation    Restriction    Damping    Smoothing

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# Computing the initial point

Need  $x_{r,0}$  (starting point at topmost level):  
→ use a **mesh refinement** technique.

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For  $i = 0, \dots, r - 1,$

- apply the **recursive algorithm** to solve

$$\min_x f_i(x)$$

(with **increasing accuracy**)

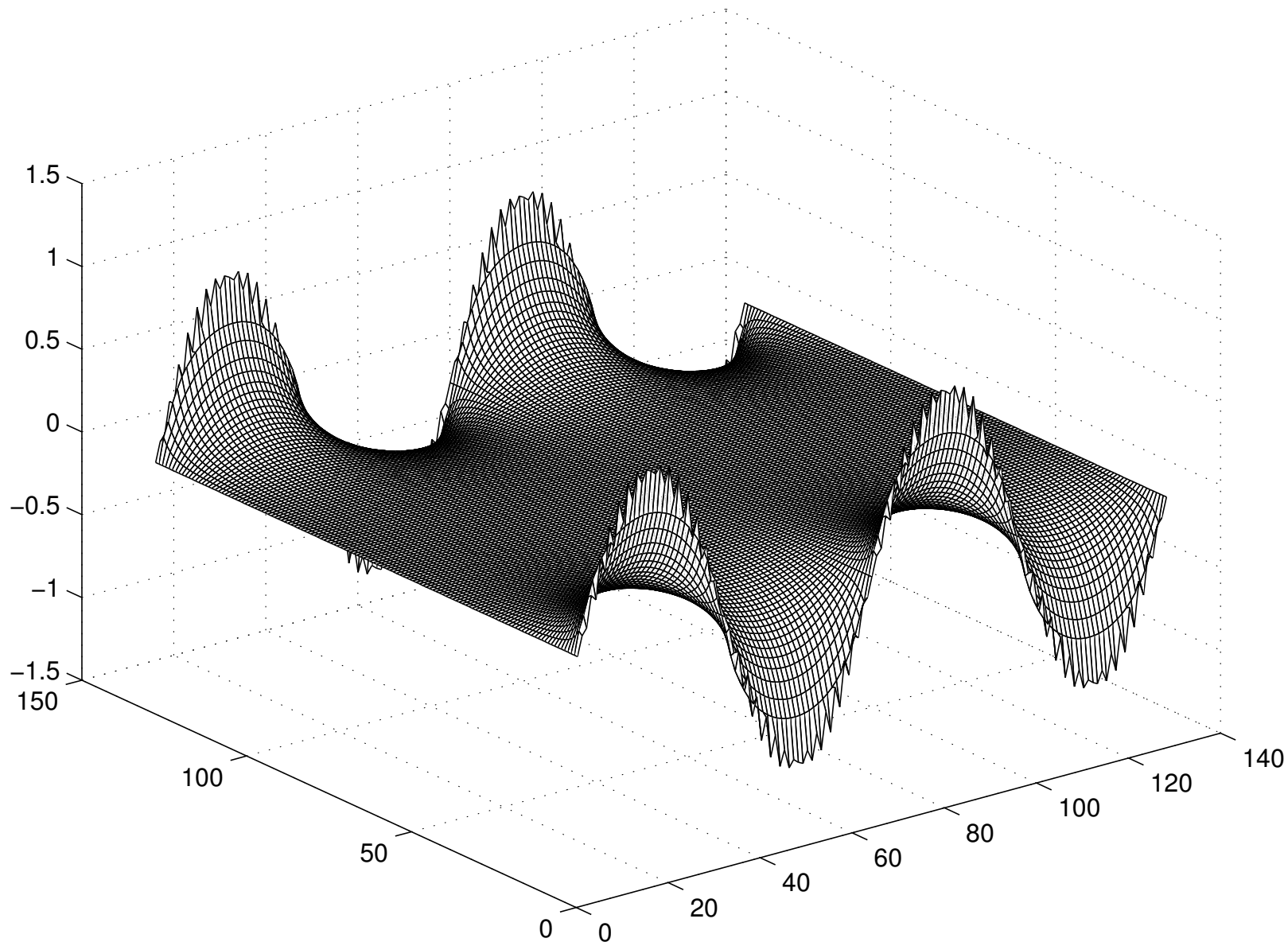
- apply the **prolongation** to obtain the initial point at next level

- reminiscent of the **full multigrid scheme**
- approach of the solution **at coarse levels**



# A minimum surface problem

solution at level 5



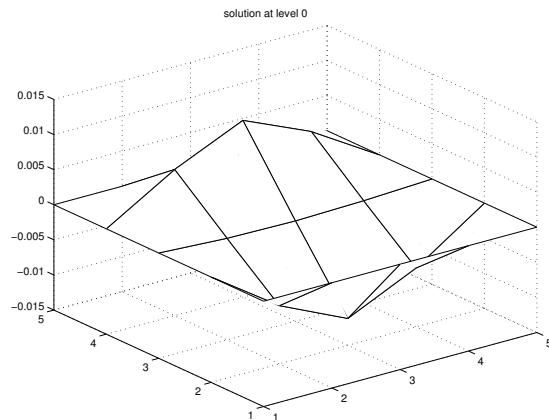
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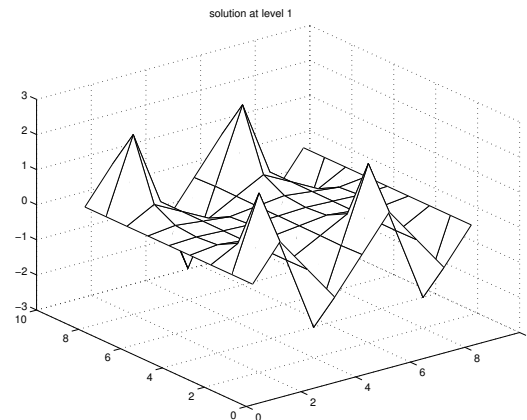
# The level structure

## Plan

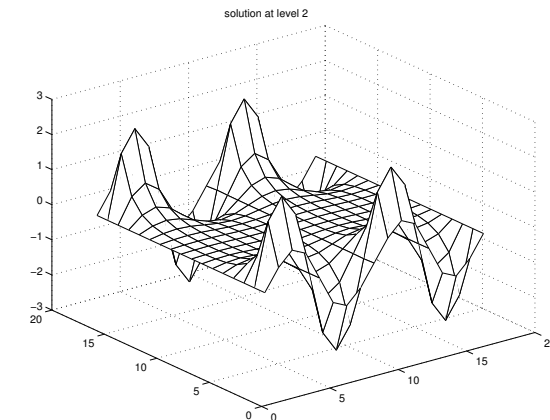
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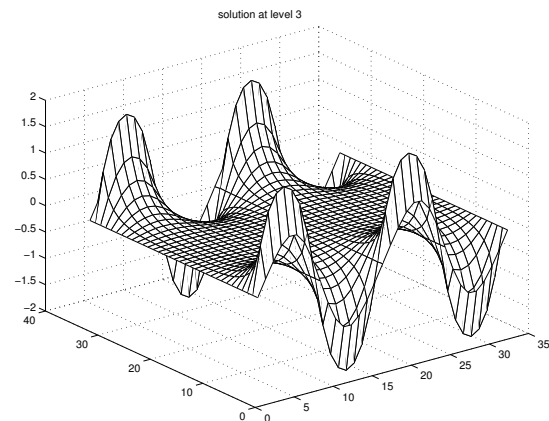
$$n = 3^2 = 9$$



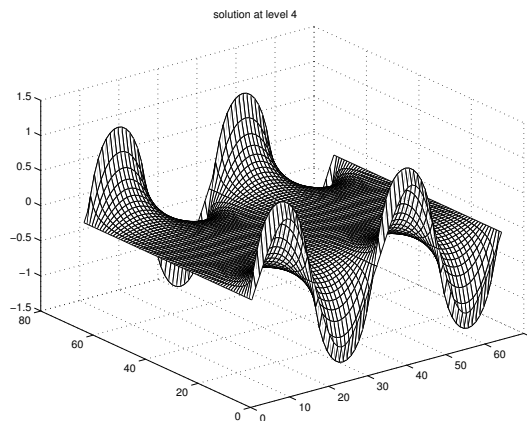
$$n = 7^2 = 49$$



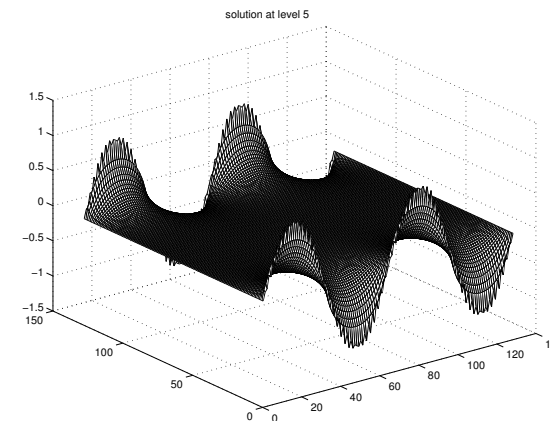
$$n = 15^2 = 225$$



$$n = 31^2 = 961$$



$$n = 63^2 = 3969$$



$$n = 127^2 = 16129$$



# Further problem details

- **structured** level transfer operators
  - $P$  = full weighting interpolation operator
  - $R$  = normalized  $P^T$
- handling the **boundary condition**
  - boundary condition not forced
  - additional **smoothing** “just inside”
- random starting point (at coarsest level)

Contact me for a live demo ...



# Other test problems (1)

- 2D Laplacian (check) problem (5 points FD pencil, unit square)

$$\min -\frac{1}{2}x^T \Delta x - f^T x$$

$$f = \sin[x_1 * (1 - x_1)] * \sin[x_2 * (1 - x_2)]$$

- 2D nonconvex quartic nonlinear least-squares (5 points FD pencil, unit square)

$$\min \int (u - f)^2 + 10^{-2} \int (\gamma - f)^2 + \int (-\Delta u + u\gamma - g)^2$$

$$g = -\Delta f + f^2$$

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# Other test problems (2)

- Lewis and Nash Dirichlet to Neumann transfer problem:

$$a : [0, \pi] \rightarrow \mathbb{R} \int_0^\pi \left( \frac{\partial u}{\partial x_2}(x_1, 0) - \phi(x_1) \right)^2 dx_1$$

with  $S = \{(x_1, x_2), 0 \leq x_1 \leq \pi, 0 \leq x_2 \leq \pi\}$

$$\Gamma = \{(x_1, x_2), 0 \leq x_1 \leq \pi, x_2 = 0\}.$$

$$\phi(x) = \sum_{i=1}^{15} \sin(i x) + \sin(40 x)$$

subject to the boundary value problem

$$\begin{cases} \Delta u = 0 \\ u(x, y) = a(x_1) \text{ on } \Gamma, \quad u(x, y) = 0 \text{ on } \partial S \setminus \Gamma \end{cases}$$

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# Other test problems (3)

- Borzi and Kunish's solid ignition optimal control (unit square):

$$\min_f \left\{ \int_{\Omega} (u - z)^2 + \frac{\beta}{2} \int_{\Omega} (e^u - e^z)^2 + \frac{\nu}{2} \int_{\Omega} f^2 \right\}$$

where

$$\begin{cases} -\Delta u + \delta \exp(u) = f & \text{in } \Omega \\ u = 0 & \text{on } \partial\Omega. \end{cases}$$

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# Other test problems (4)

- Vogel's image deblurring inverse problem:

$$\min \mathcal{J}(f) = \frac{1}{2} \|Tf - d\|_2^2 + TV(f),$$

where  $TV(f)$  is the discretization of the total variation function

$$\int_0^1 \int_0^1 (1 + (\partial_x f)^2 + (\partial_y f)^2)^{\frac{1}{2}} dx dy.$$

## Plan

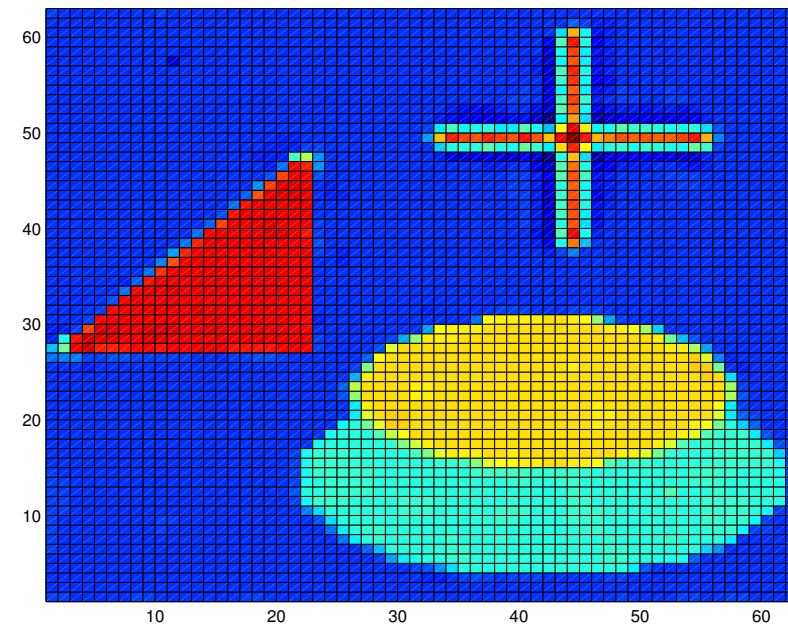
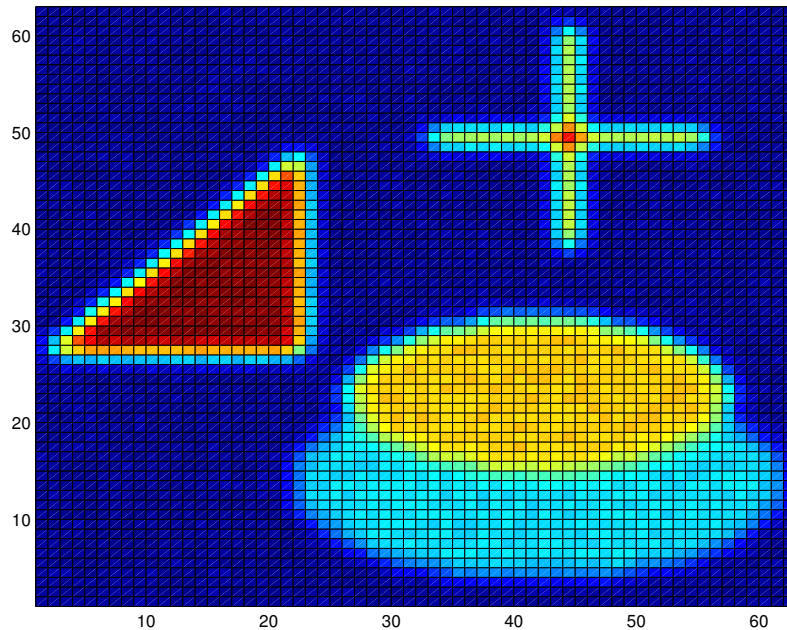
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# Other test problems (5)

Vogel's problem data and result:

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# A typical run (minimum surface)

V style, pure quadratic recursion,  
2 smoothing cycles, gradient accuracy:  $1e-07$

level	3	7	15	31	63	127	255
Tayl. its	10	0	0	0	0	0	0
smooth cyc	0	59	155	232	233	173	70
prolong	0	2	13	25	29	25	20
restric	0	4	26	50	58	50	40
backtrs	0	0	0	0	0	0	1
evals f	12	6	10	18	20	39	72
evals g	6	6	10	18	20	30	42
evals H	6	2	3	6	5	8	16

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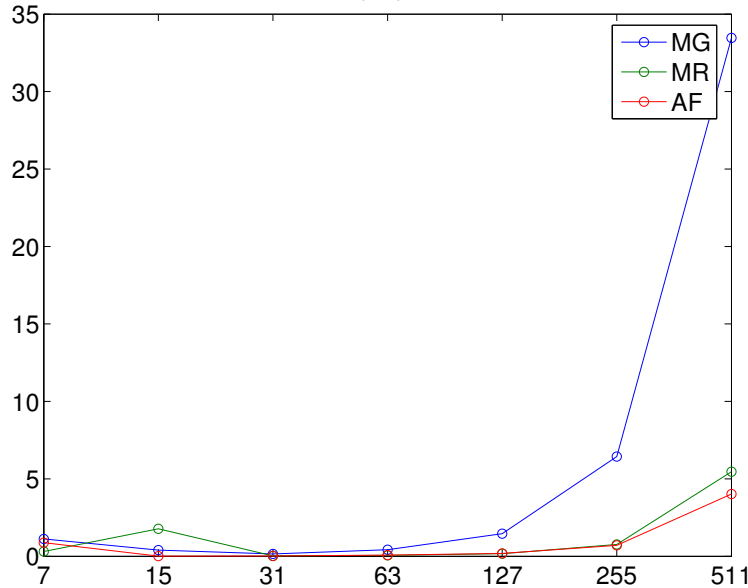


# Problem size and CPU (1)

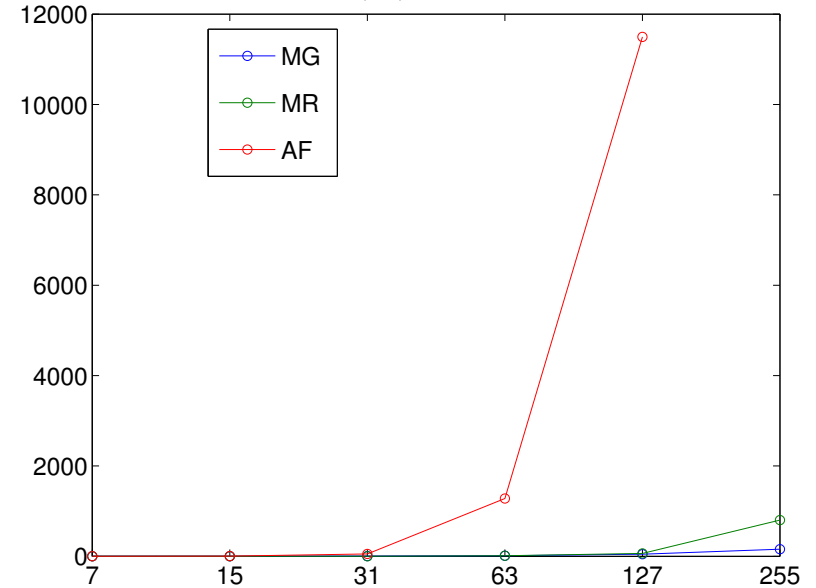
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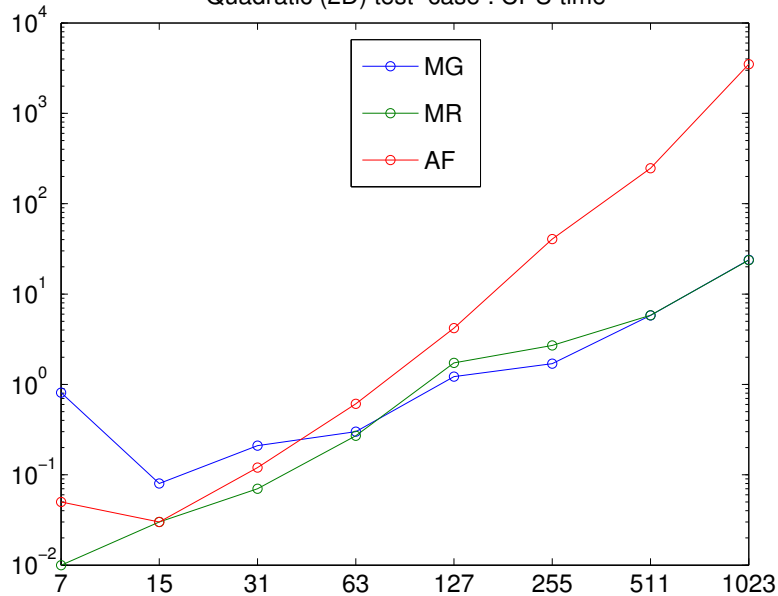
Dirichlet to Neuman (1D) test-case : CPU time



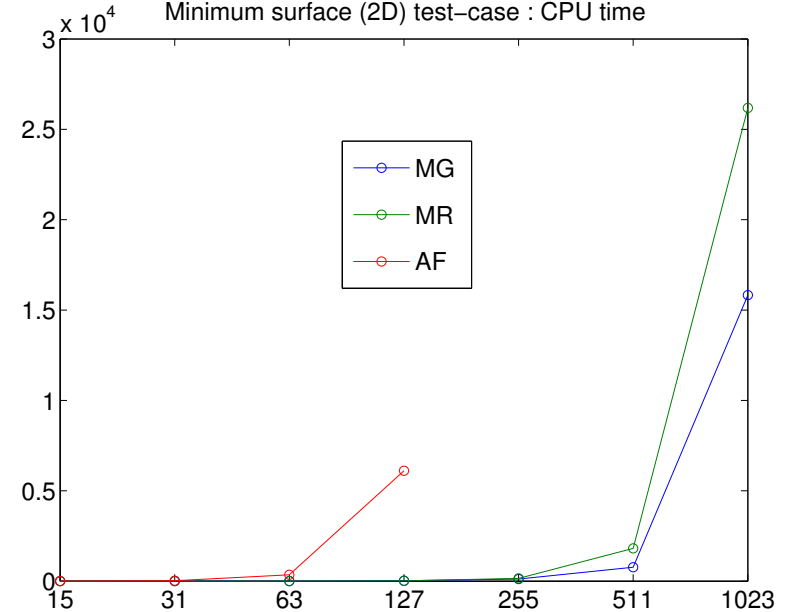
Nonconvex (2D) test-case : CPU time



Quadratic (2D) test-case : CPU time



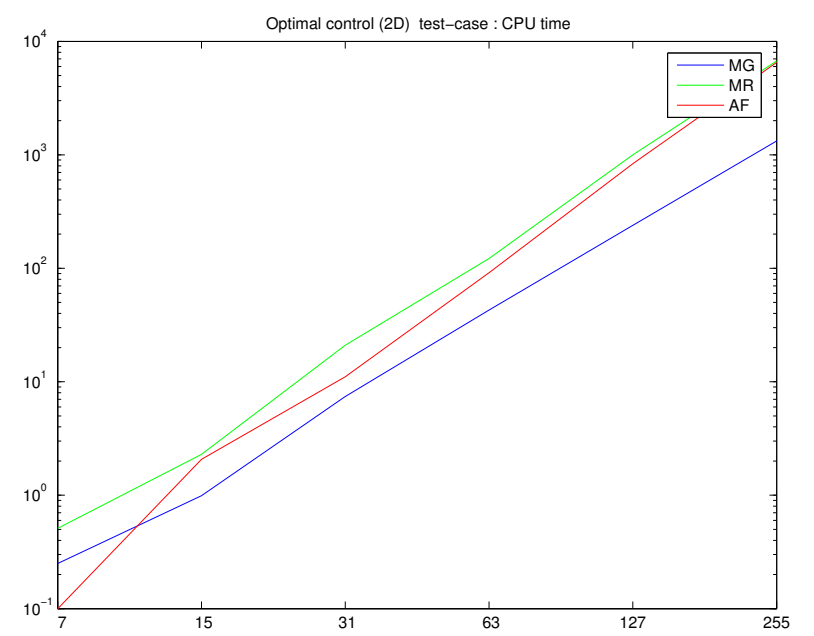
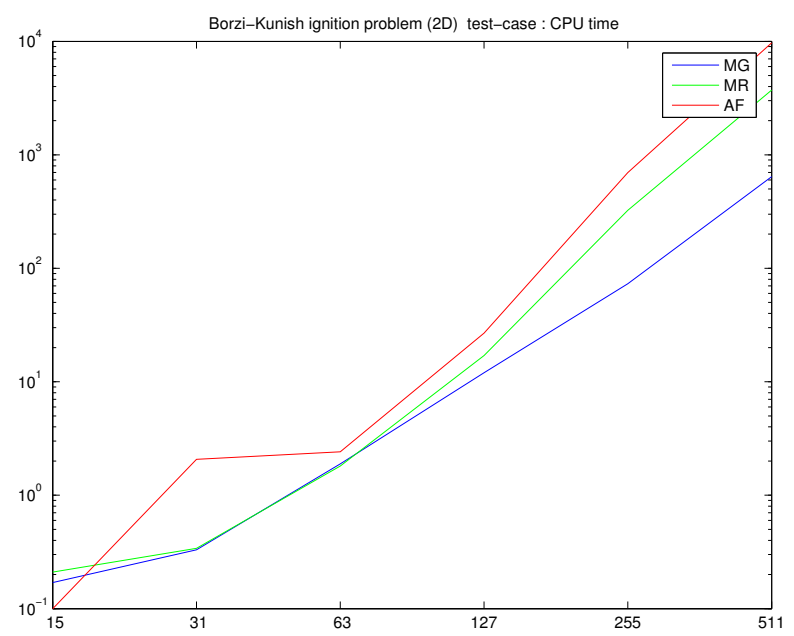
Minimum surface (2D) test-case : CPU time





# Problem size and CPU (2)

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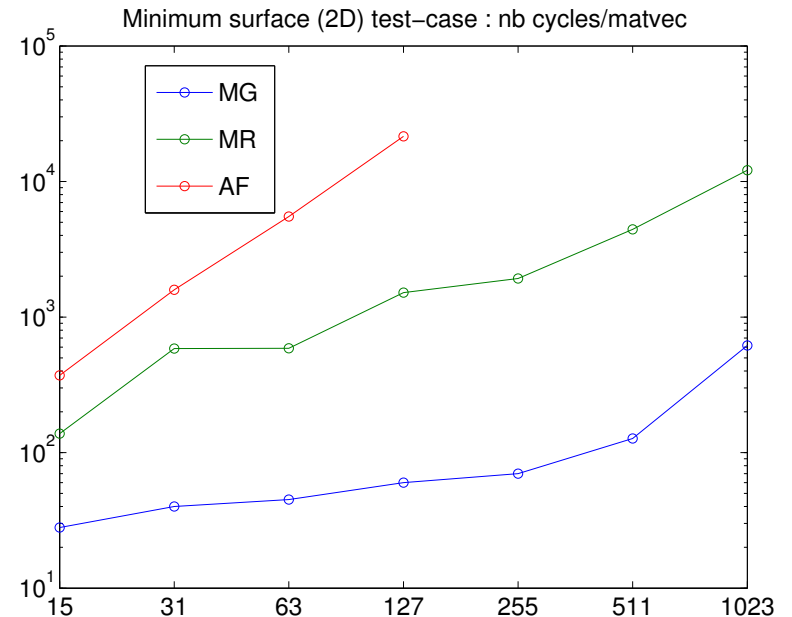
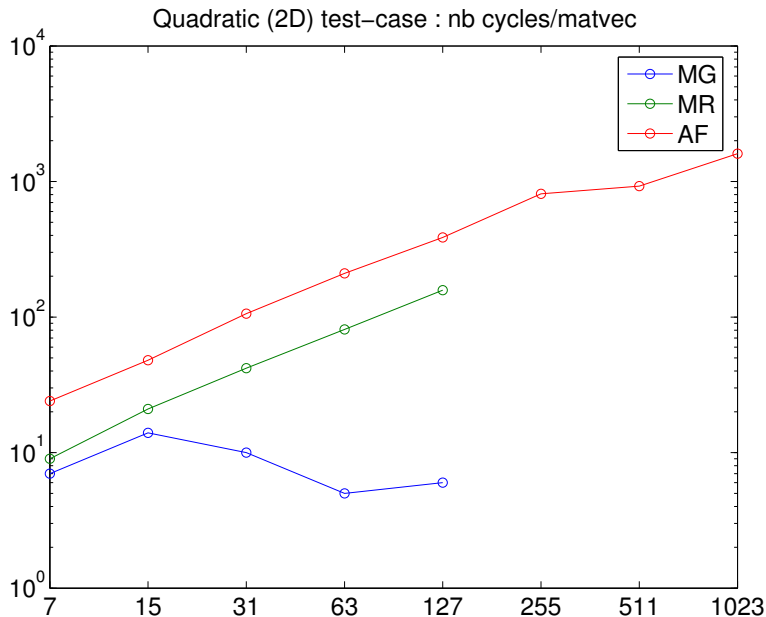
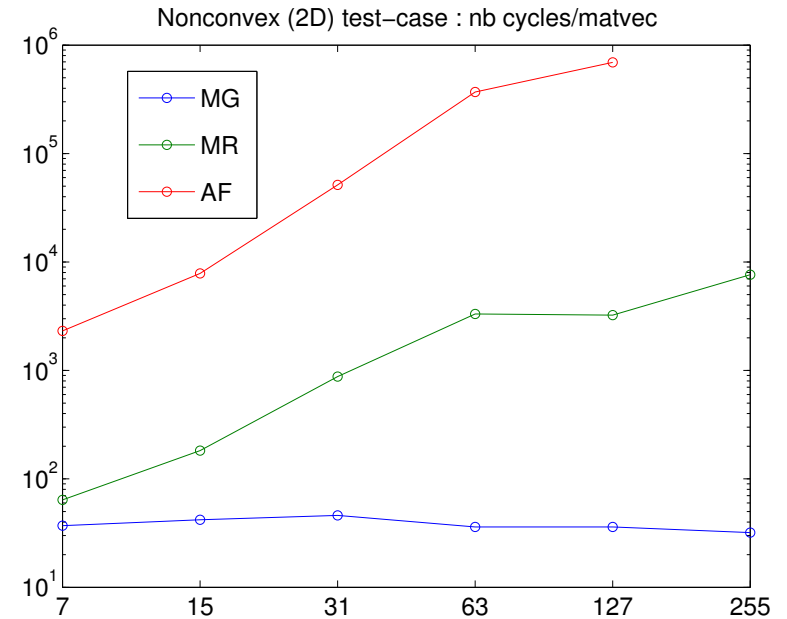
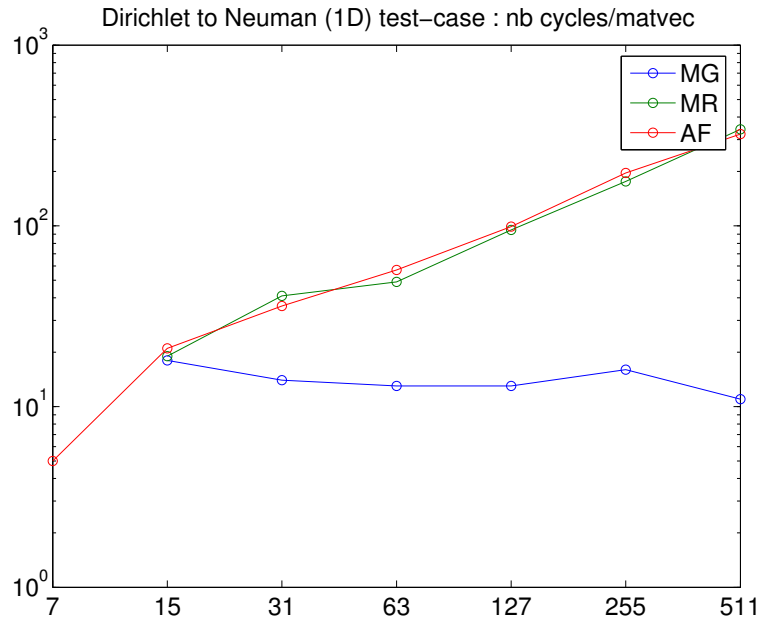




# Problem size and linear algebra

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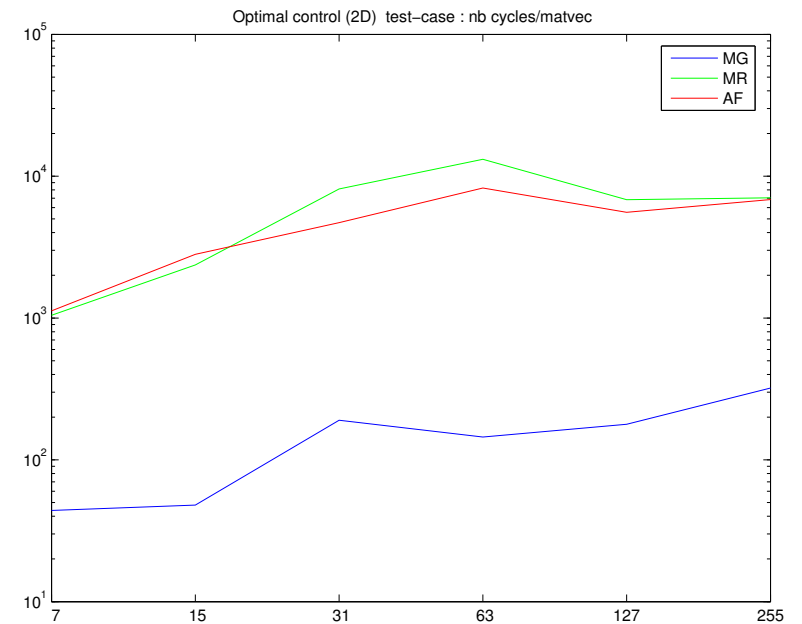
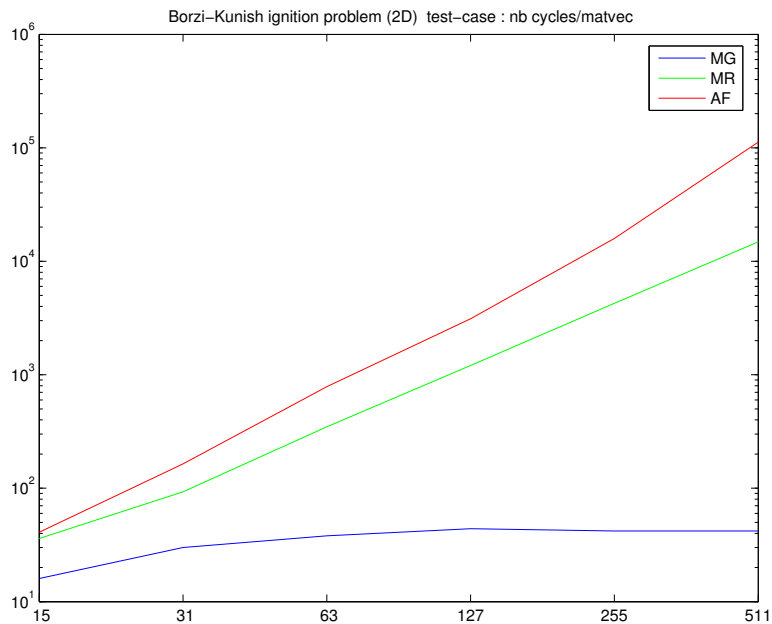




# Problem size and linear algebra (2)

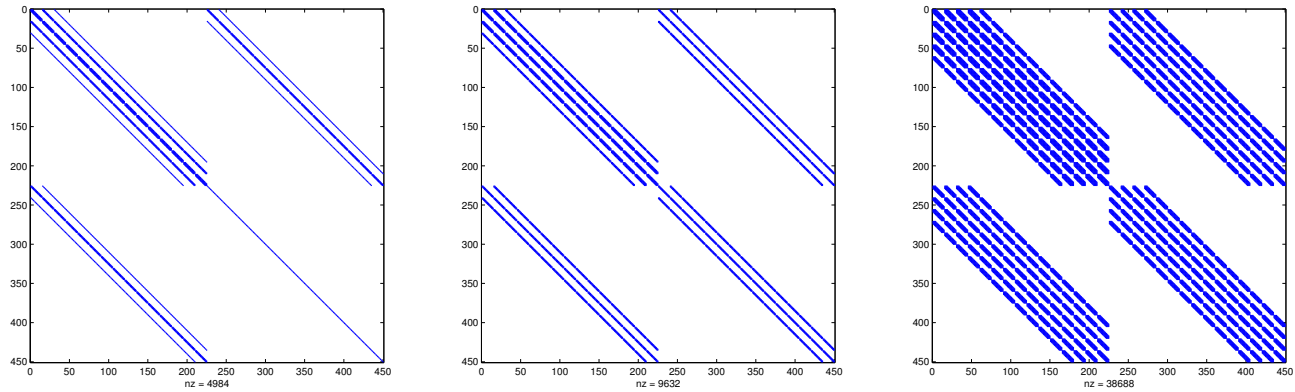
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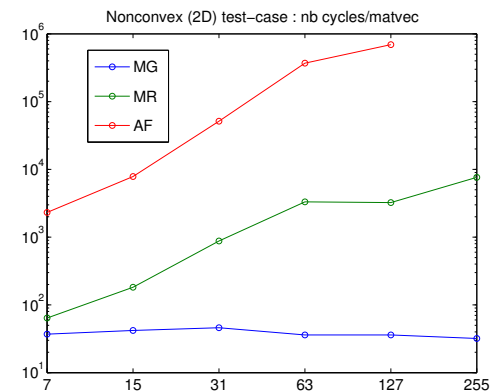
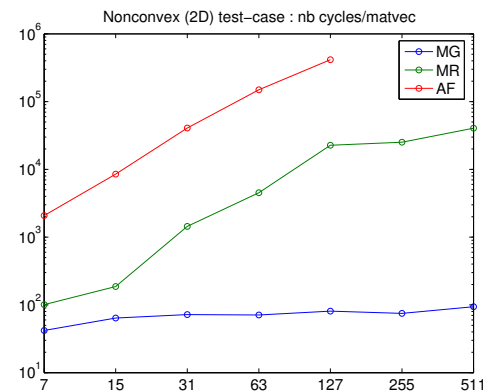
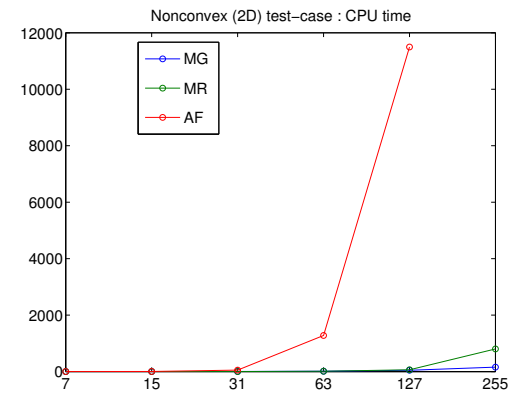
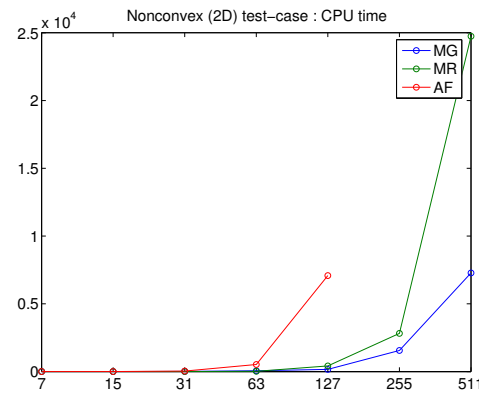


# Interpolation and fill-in



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# Convergence to critical points

Based on trust region technology

- the **sufficient decrease** argument (imposed in Taylor's iterations)
- a conditional coarsening condition

Main results:

- convergence to **first-order critical points at all levels**
- (weak) upper bound on the number of iterations to achieve a given accuracy

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# Convergence to weak minimizers

## Difficulties:

- **curvature condition** too costly to impose a posteriori on recursive iterations: only a subset of iterations can be checked
- Gauss-Seidel only looks at **coordinate directions**

## Main result:

- **asymptotic positive curvature** on coarsest subspace + prolongation of all coarse coordinate directions

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# Current conclusions

- **more** efficient than **mesh refinement** for large instances
- pure quadratic recursion ( $f_i = 0$ , Galerkin) very efficient
- interpolation **degree** crucial
- V cycles most efficient

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# Perspectives

Encouraging (so far)

- more numerical experiments!
- a (more natural)  $l_\infty$  version
- **multigrid-type** developments:  
(algebraic multilevel, p-multigrids, ...)
- constrained problems  
(bounds, equalities, general)
- non-monotone (filter) techniques
- ... and much more!

Thank you for your attention

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→ *Perspectives*