

# Multigrid Implementation in COMSOL Multiphysics

- Comparison of Theory and Practice



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How can you easily destroy your modeling effort?	Choose an inadequate solver or a badly parameterized one!
How can you avoid this?	
	You need knowledge on numerical methods – at least confidence in implemented solver techniques.
How can you obtain this?	
	Let the results for accepted model problems convince you about properties of a particular solver. Here: <b>geometric MG</b> .
If you like this	"COMSOL solver tuning" (May 2014, Berlin). More general, more solver.



#### MG – a combination of **smoothing**

CGC









What would we like to have for a MG solver?

- steady convergence rate (both error and residual reduction) from the first cycle till the last one
- "h-independent" convergence
- fast convergence
- convergence depends on the quality of smoothing
- linear complexity O(N)
- moderate memory requirement<sub>5</sub>\_
- faster than other solver





<b>Poisson equation:</b> $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial u^2} = f(x, y)$ mit $\varepsilon_x = 1.0, \varepsilon_y = 1.0$					
6.008.001 d	.o.f.				
smoother	time	no. of	time per	$\varrho_{comsol}$	v. mem.
	[s]	cycles	cycle		(GB)
SOR	175	10	2.8	0.046	17.0
SOR	181	9	3.9	0.040	15.6
SOR	184	9	4.3	0.040	15.7
SSOR	195	10	3.7	0.053	15.9
SSOR	197	9	5.6	0.048	15.0
SSOR	203	9	6.3	0.048	15.0
Vanka	196	10	4.8	0.053	18.0
Vanka	205	9	6.1	0.048	18.0
Vanka	210	9	6.9	0.048	18.0
	: $\varepsilon_x \frac{\partial^2 u}{\partial x^2} +$ <b>5.008.001</b> d smoother SOR SOR SOR SSOR SSOR SSOR Vanka Vanka Vanka	: $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2}$ 5.008.001 d.o.f. smoother time [s] SOR SOR SOR SOR SOR 175 181 184 184 SSOR 195 SSOR 195 SSOR 197 SSOR 203 Vanka 205 Vanka 210	: $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2} = f(x,$ 5.008.001 d.o.f. smoother time no. of [s] cycles SOR SOR SOR SOR SOR SSOR 195 10 9 10 SSOR 195 10 9 SSOR 197 9 SSOR 203 9 Vanka 205 9 Vanka 210 9	: $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2} = f(x, y)$ mit $\varepsilon_x$ 5.008.001 d.o.f. smoother time no. of time per [s] cycles cycle SOR SOR SOR SOR SOR SOR 175 10 9 4.3 3.9 4.3 SSOR 195 10 3.7 SSOR 195 10 3.7 SSOR 195 10 3.7 SSOR 197 9 5.6 SSOR 203 9 6.3 Vanka 205 9 6.1 Vanka 210 9 6.9	$\begin{array}{c c} \varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2} = f(x,y) & \text{mit } \varepsilon_x = 1.0, \varepsilon_y \\ \textbf{5.008.001 d.o.f.} \\ \textbf{smoother time no. of time per } \varrho_{comsol} \\ \textbf{SOR } & \textbf{[s] cycles cycle} \\ \textbf{SOR } & \textbf{[s] } \textbf{cycles } \textbf{cycle} \\ \textbf{SOR } & \textbf{[s] } \textbf{9} & \textbf{3.9} \\ \textbf{SOR } & \textbf{175 } \textbf{10} & \textbf{3.7 } \textbf{0.046} \\ \textbf{0.040 } \\ \textbf{0.048 } \\ \textbf{0.053 } \\ \textbf{0.048 } \\ \textbf{0.053 } \\ \textbf{0.048 } \\$







Poisson equation: $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2} = f(x, y)$				What predicts theory?		
"free triangular mesh": 8.786.945 (8L), and 35.1				the weighted relaxation fact	Jacobi-me or 1.0 has	ethod with s no
solver	smoother	total time (seconds)	no. cycl	good smoothi Poisson equat	ng proper tion, best	ty for the with 0.8
MG-W(2,1)-7L	Jacobi $\omega = 1.0$		5		divergence	
MG-V(2,1)-8L	Jacobi $\omega = 1.0$	84	29	1.1	0.402	8.7
MG-V(2,1)-8L	Jacobi $\omega=0.8$	71	17	1.1	0.194	8.6
MG-F(2,1)-8L	Jacobi $\omega=0.8$	77	17	1.4	0.192	8.6
MG-W(2,1)-8L	Jacobi $\omega = 0.8$	79	17	1.6	0.192	8.7
MG-V(2,1)-9L	Jacobi $\omega=0.8$	596	17	8.4	0.193	30.7
MG-V(2,1)-9L	SOR	450	10	3.0	0.051	29.3
				0.0		

the 9L problem could not be solved by a direct solver







Anisotropic Poisson equation:  $\varepsilon_x \frac{\partial^2 u}{\partial x^2} + \varepsilon_y \frac{\partial^2 u}{\partial y^2} = f(x, y)$  with  $\varepsilon_x = 0.01, \varepsilon_y = 1.0$ "free triangular mesh": 8.786.945 (8L) d.o.f.

solver	smoother	total time	no. of	time per	$arrho_{comsol}$	v. mem.
		( <u>seconds</u> )	cycles	cycle		(GB)
MG-V(2,1)-8L	SOR	391	298	1.14	0.915	8.4
MG-F(2,1)-8L	SOR	502	297	1.51	0.915	8.4
MG-W(2,1)-8L	SOR	544	297	1.65	0.915	8.4
MG-V(2,1)-8L	SSOR	407	198	1.79	0.877	8.3
MG-V(2,1)-8L	Vanka	450	198	2.00	0.877	9.7
MG-V(2,1)-8L	SORline	441	68	5.08	0.692	10.3



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#### cooling of a thermos, heat transfer, laminar flow, free convection stationary, nonlinear



d.o.f.	solver	total time	v. mem.
	characteristics	(seconds)	(GByte)
1.975.146	MG-V(2,1)-6L SOR	50	7.4
	Pardiso	83	11.1
	MUMPS	157	9.9
	Spooles	254	8.6
5.084.465	MG-V(2,1)-7L SOR	128	8.3
	Pardiso	220	19.2
	MUMPS	410	14.8
	Spooles	857	20.6
20 323 281	MG-V(21)-81 SOR	695	25 5
	Pardiso	cancelled	
	MUMPS	cancelled	43.8
	Spooles	omitted	-



What has been observed?

- numerical complexity of the cycle is reflected well by the time per cycle
- convergence speed of F- and W-cycle are identical
  - F- and W- converge faster than V-cycle
- MG with V(2,1)-cycle usually is the fastest MG-solver (time to solve)
- steady convergence speed for all cycles (first to last)
- convergence speed is almost h-independent
- linear behavior is (almost) given
- moderate memory requirements, especially when compared to direct solver
- Jacobi smoother reacts on relaxation parameter as known from theory
- MG without coloured relaxation pattern or block relaxation behaves as predicted for the anisotropic Poisson equation
- MG beats all direct solver except for the anisotropic problem the direct solver could not solve the very large problems

# MG implementation in COMSOL is reliable – use it



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