Recursively Deflated PCG for mechanical problems

C. Vuik, T.B. Jönsthövel, M.B. van Gijzen

Delft University of Technology

Householder Symposium XVIII: 16 June, 2011

Introduction

problem from the work floor: material analysis

Iterative methods

overview existing solvers deflation method recursive deflation

Numerical experiment: real asphalt core

Questions and references

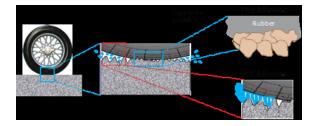


Figure: EU project, SKIDSAFE: asphalt-tire interaction

20th century science

consider materials to be homogeneous

21th century science

shift from MACRO to MESO/MICRO scale

- Obtain CT scan from material specimen
- Convert CT scan to mesh
- Use finite element method for discretization of governing equations

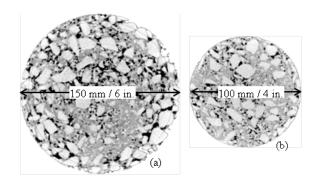


Figure: CT scan of asphalt column



Figure: from CT scan to mesh, approx. 3 mln DOF



governing equations

$$K\Delta u = \Delta f \tag{1}$$

Stiffness matrix K, change in displacement Δu and change of force Δf . The change of force involves evaluation of non-linear equations that depend on displacement field.

properties of stiffness matrix K

- symmetric, positive definite: $\forall \Delta u \neq 0, \ \Delta u^T K \Delta u > 0$
- $K \in \mathbb{R}^{n \times n}$, $n >> 10^6$
- discontinuities in values matrix entries ~ O (10⁶): ill-conditioned

Existing solvers

just some possible methods and preconditioners

- preconditioned conjugate gradient method (PCG) combined with,
 - BIM: Jacobi, SSOR
 - Decomposition methods: (Additive-Schwarz) $ILU(\epsilon)$
- direct solvers: MUMPS, PARDISO, SuperLU
- multigrid: geometric multigrid, algebraic multigrid (smoothed aggregation)

bottom line: no free lunch

no black box solution for large, ill-conditioned systems

- performance of PCG depends on spectrum of K, large jumps induce small eigenvalues, hence performance degrades when number of jumps (different materials) increases
- direct solvers (may) become expensive for large meshes
- AMG can be insensitive to jumps, however to achieve this one has to define the coarse grid specifically

Use deflation

Deflation based operator is not a classical preconditioner, i.e. it is not an approximation of K. The deflation operator is a projection which, by the right choice of the projection vectors, removes eigenvalues from the spectrum of the projected system.

definition

split displacement vector u,

$$u = (I - P^T) u + P^T u, (2)$$

and let us define the projection P by,

$$P = I - KZ(Z^TKZ)^{-1}Z^T, \quad Z \in \mathbb{R}^{n \times m}$$
(3)

the DPCG method

We use deflation based operator in conjunction with preconditioning (e.g. diagonal scaling) to remove those small eigenvalues that correspond to the jumps (discontinuities) in the values of the stiffness matrix.

Deflated Preconditioned Conjugate Gradient (DPCG) method Solve for $M^{-1}PK\Delta u = M^{-1}P\Delta f$

- We have observed in [2]¹ that the rigid body modes of the regions corresponding to the different materials coincide with the eigenvectors of the 'jump' eigenvalues.
- By removing those rigid body modes (RBM) using deflation, we remove the corresponding 'jump' eigenvalues from the system.
- The rigid body modes of sets of finite elements can be easily computed.



How do RBM relate to stiffness matrix K?

The kernel of the element matrix of an arbitrary unconstrained finite element is spanned by the rigid body modes of the element. In 3D six rigid body modes: three translations, three rotations.

Theorem

We assume a splitting K = C + R such that C and R are symmetric positive semi-definite with $\mathcal{N}(C) = \text{span}\{Z\}$ the null space of C [1]². Then

$$\lambda_i(C) \le \lambda_i(PK) \le \lambda_i(C) + \lambda_{max}(PR).$$
 (4)

Moreover, the effective condition number of PK is bounded by,

$$\kappa_{\text{eff}}(PK) \le \frac{\lambda_n(K)}{\lambda_{m+1}(C)}.$$
(5)



00000000

How do RBM relate to stiffness matrix K?

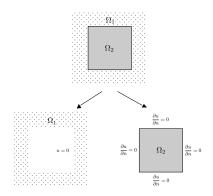


Figure: Principle of rigid body mode deflation

How do RBM relate to stiffness matrix K?

00000000

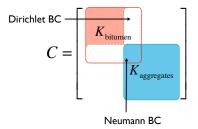


Figure: Principle of rigid body mode deflation: construction of C

How do RBM relate to stiffness matrix K?

0000000

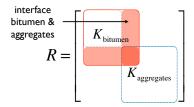


Figure: Principle of rigid body mode deflation: construction of R

Recursive deflation

However, the definition of P given by first theorem does not provide insight in the effect of individual deflation vectors on the spectrum of PK. Introduce a recursive deflation operator which can be used for more extensive eigenvalue analysis of PK.

Definition

$$P^{(k)} = I - KZ_k(Z_k^{\mathrm{T}}KZ_k)^{-1}Z_k^{\mathrm{T}}$$
 with $Z_k = [\tilde{Z}_1, \tilde{Z}_2, ..., \tilde{Z}_k]$, where $\tilde{Z}_j \in \mathbb{R}^{n \times l_j}$ and has rank l_j .

Let $P^{(k)}$ and Z_k as in Definition 2, then

$$P^{(k)}K = P_k P_{k-1} \cdots P_1 K$$

Recursive deflation

where
$$P_{i+1} = I - \tilde{K}_i \tilde{Z}_{i+1} (\tilde{Z}_{i+1}^T \tilde{K}_i \tilde{Z}_{i+1})^{-1} \tilde{Z}_{i+1}^T$$
, $\tilde{K}_i = P_i \tilde{K}_{i-1}$, $\tilde{K}_0 = K$,

 $\tilde{Z}_{i}^{\mathrm{T}}\tilde{K}_{i-1}\tilde{Z}_{i}^{\mathrm{T}}$ and $Z_{k}^{\mathrm{T}}KZ_{k}$ are non-singular because \tilde{Z}_{i} are of full rank and K is a symmetric positive definite matrix.

Recursive deflation

Proof.

by induction,

- i. show $P_1K = P^{(1)}K$ where $Z_1 = \tilde{Z}_1 \in \mathbb{R}^{n \times l_1}$,
- ii. assume $P_{i-1}\tilde{K}_{i-2}=\tilde{K}_{i-1}=P^{(i-1)}K$ where $Z_{i-1}=[\tilde{Z}_{i-1},\tilde{Z}_{i-2},\cdots,\tilde{Z}_1]$, show that $P_i\tilde{K}_{i-1}=P^{(i)}K$ where $Z_i=[\tilde{Z}_i,Z_{i-1}],\ Z_{i-1}\in\mathbb{R}^{n\times l(i-1)},\ \tilde{Z}_i\in\mathbb{R}^{n\times l_i}$ and $I=\sum_{r=i}^i l_i$.

00000000

Recursive deflation: 1D example

Poisson equation,

$$-\frac{d}{dx}\left(c(x)\frac{du(x)}{dx}\right) = f(x), x \in [0, I]$$
$$u(0) = 0, \frac{du}{dx}(I) = 0$$

Recursive deflation: 1D example

Introduce a FE mesh for the line [0,I] including 3 domains $\Omega_1=\{x_1,...,x_4\},\ \Omega_2=\{x_5,...,x_8\}$ and $\Omega_3=\{x_9,...,x_{13}\}.$ For sake of simplicity we will write $c_i=c(x_i)$ where i=1,...,13, $x_1=h$ and $x_{13}=I$. Furthermore because c_i is constant on each material domain we will use $c_i=c_1$, $c_i=c_2$ and $c_i=c_3$ on Ω_1 , Ω_2 and Ω_3 respectively.

After discretization,

Recursive deflation: 1D example

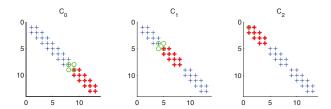


Figure: sparsity pattern C_0 , C_1 and C_2 . Nonzero elements represented by symbols; corresponding to deflated material, interface elements and remaining elements pictured by bold crosses, circles and non bold crosses respectively.

Recursive deflation: 1D example

0000000

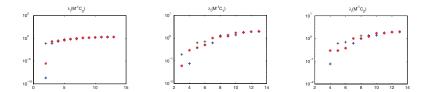


Figure: spectrum of $M^{-1}C_i$ (* correct, + wrong choice deflation vectors) compared to spectrum of $M^{-1}K$ (+)

Consider picture from introduction. Size of system approx. 3 million DOF, material parameters given in table below,

Table:

(a) E modulus materials

aggregate	bitumen	air voids
70000	5000	100

We compare PCG and DPCG combined with three different preconditioners,

- diagonal scaling: low cost, weak properties
- AMG smoothed aggregation, default parameters, no specific information on mesh provided: relative low set up and solve cost, designed for solving elastic equations
- AMG smoothed aggregation, approx. null space of operator and dof-to-node mapping provided: expensive set up and solve cost, high memory usage

	4 CPUs	8 CPUs	64 CPUs	iterations
PCG - diag	nc	nc	nc	nc
DPCG - diag	9883	5456	680	9018
PCG - SA	6687	6906	1123	2018
DPCG - SA	9450	5043	771	1210
PCG - SA ⁺	oom	2200	oom	407

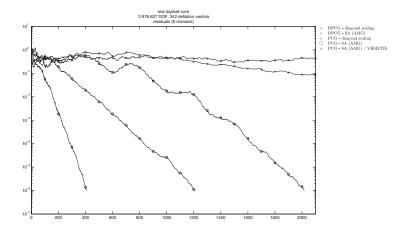
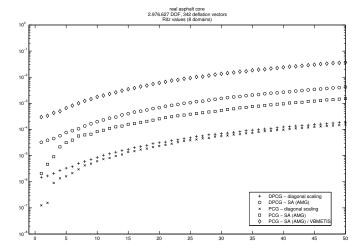
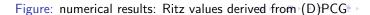


Figure: numerical results: residuals







Questions and references



On the construction of deflation-based preconditioners. SIAM J. Sci. Comput., 23(2):442-462, 2001.



T.B. Jönsthövel, M.B. van Gijzen, C.Vuik, C. Kasbergen, and A. Scarpas.

Preconditioned conjugate gradient method enhanced by deflation of rigid body modes applied to composite materials. Computer Modeling in Engineering and Sciences, 47:97–118. 2009.

http://ta.twi.tudelft.nl/nw/users/vuik/papers/Jon11GVS.pdf http://ta.twi.tudelft.nl/nw/users/vuik/papers/Jon11GMVS.pdf

