



*A Quadratically Constrained  
Eigenvalue Minimization Problem  
Arising from PDE of Monge-Ampère  
Type*

D.C. Sorensen

- ▶ Collaborator: **R. Glowinski**
- ▶ Support: **NSF and AFOSR**

IWASEP VII

Dubrovnik

June 2008

---

*True Title*

---

**Fooling Around and Having Fun  
with  
Numerical Linear Algebra**

---

## Outline

---

- ▶ Motivation – Monge-Ampère Equations
- ▶ FEM Discretization → Quadratic Min Problem
- ▶ Formulation of General Eigenvalue Constrained  
Quadratic Min Problem
- ▶ The General Secular Equation and Its Solution
- ▶ A Surprising Result – Finite  $\mathcal{O}(n^3)$  Complexity
- ▶ Some Numerical Results

---

## Monge-Ampère Equations

---

Applications and Research Reference:

BIRS Workshop (2003) on Monge-Ampère Equation

<http://www.birs.ca/workshops/2003/03w5067/report03w5067.pdf>

Monge-Ampère Equations arise in

Riemannian Geometry

Conformal Geometry

CR Geometry

**Example** : Problem of prescribed Gauss curvature

A real-valued function  $K$  is specified on a domain  $\Omega$  in  $\mathbb{R}^n$

Prescribed Gauss curvature seeks to identify a hypersurface of

$\mathbb{R}^{n+1}$  as a graph  $z = u(x)$  over  $x \in \Omega$  so that

the Gauss curvature is given by  $K(x)$  at every  $x$ .

---

## Monge-Ampère Problem

---

Fully-Nonlinear 3D BVP (of Dirichlet type);

Find  $\psi$  such that

$$\lambda_1\lambda_2 + \lambda_2\lambda_3 + \lambda_3\lambda_1 = f \text{ in } \Omega, \quad \psi = g \text{ on } \partial\Omega,$$

where

- ▶ Function  $\psi$  unknown;  $\Omega$  is a bounded domain of  $\mathbb{R}^3$
- ▶  $\{\lambda_1, \lambda_2, \lambda_3\}$  is the spectrum of

$$\text{Hessian } \mathbf{D}^2\psi = \left( \frac{\partial^2\psi}{\partial x_i \partial x_j} \right)_{1 \leq i, j \leq 3}$$

- ▶  $f$  and  $g$  are two given functions with  $f > 0$ .

---

## Dirichlet Problem for $\sigma_2$ -Operator

---

Rewrite as

$$|\nabla^2\psi|^2 - \mathbf{D}^2\psi : \mathbf{D}^2\psi = 2f \text{ in } \Omega,$$

where

$$\mathbf{A} : \mathbf{B} = \sum_{1 \leq i, j \leq d} a_{ij} b_{ij} = \text{trace} \mathbf{A}^T \mathbf{B} \quad (\text{Frobenius})$$

Fully nonlinear PDE becomes

$$[\text{trace}\{\mathbf{D}^2\psi\}]^2 - \text{trace}\{(\mathbf{D}^2\psi)^2\} = 2f.$$

---

## Solving when $\sigma_2$ -Operator is *Elliptic*

---

Least Squares Approach:

$\sigma_2$ -operator linearized in neighborhood of elliptic  $\psi$  gives:

$$\phi \rightarrow 2[\nabla^2\psi\nabla^2\phi - \mathbf{D}^2\psi : \mathbf{D}^2\phi].$$

Coefficient Matrix

$$2[\nabla^2\psi\mathbf{I} - \mathbf{D}^2\psi],$$

The  $\sigma_2$ -operator *elliptic* in nbhd of  $\psi \iff$  if Matrix is s.p.d. (or n.p.d) , i.,e.,

$$(\lambda_1 + \lambda_2)(\lambda_2 + \lambda_3) > 0, (\lambda_1 + \lambda_3)(\lambda_1 + \lambda_2) > 0.$$

$$\lambda_1 + \lambda_2 > 0, \quad \lambda_2 + \lambda_3 > 0, \quad \lambda_1 + \lambda_3 > 0.$$

---

## PDE Least Squares Problem

---

Find  $(\psi, \mathbf{P}) \in \mathcal{V}_g \times \mathbf{Q}_f$  such that  
 $J(\psi, \mathbf{P}) \leq J(\phi, \mathbf{G}), \quad \forall (\phi, \mathbf{G}) \in \mathcal{V}_g \times \mathbf{Q}_f,$   
where

$$J(\phi, \mathbf{G}) = \frac{1}{2} \int_{\Omega} (\mathbf{D}^2 \phi - \mathbf{G}) : (\mathbf{D}^2 \phi - \mathbf{G}) dx,$$

with  $dx = dx_1 dx_2 dx_3$ .



---

## Block Relaxation Algorithm

---

Given  $\psi^0 \in \mathcal{V}_g$  ;

for  $k = 0, 1, 2, \dots$

- ▶  $\mathbf{P}^{k+1} = \operatorname{argmin}_{\mathbf{G} \in \mathbf{Q}_f} J(\psi^k, \mathbf{G})$  ;
- ▶  $\psi^{k+1/2} = \operatorname{argmin}_{\phi \in \mathcal{V}_g} J(\phi, \mathbf{P}^{k+1})$  ;
- ▶  $\psi^{k+1} = \psi^k + \omega(\psi^{k+1/2} - \psi^k)$ ;

Relaxation Parameter:  $0 < \omega < 2$

Initialization

$$\nabla^2 \psi^0 = \sqrt{3f} \text{ in } \Omega, \quad \psi^0 \text{ on } \partial\Omega.$$

Note that  $\psi^0$  has the  $\mathcal{H}^2(\Omega)$ -regularity if  $\partial\Omega$  is 'sufficiently' smooth and/or  $\Omega$  is convex.

---

## Multiple Minimizations Needed

---

Must solve following at the vertices of a finite element mesh:

$$\text{Find } \mathbf{P}^{k+1}(x) \in \mathbf{E}(x), \quad \mathbf{j}_k(\mathbf{P}^{k+1}(x); x) \leq \mathbf{j}_k(\mathbf{A}; x), \quad \forall \mathbf{A} \in \mathbf{E}(x),$$

where

$$\begin{aligned} \mathbf{E}(x) = \{ \mathbf{A} \mid \mathbf{A} \in \mathbb{R}^{3 \times 3}, \quad \mathbf{A} = \mathbf{A}^T, \\ \lambda_1 \lambda_2 + \lambda_2 \lambda_3 + \lambda_3 \lambda_1 = f(x), \\ \lambda_1 + \lambda_3 > 0, \lambda_2 + \lambda_3 > 0, \lambda_3 + \lambda_1 > 0 \} \end{aligned}$$

and

$$\mathbf{j}_k(\mathbf{A}; x) = \frac{1}{2} \mathbf{A} : \mathbf{A} - \mathbf{D}^2 \psi^k(x) : \mathbf{A},$$

with  $\{\lambda_1, \lambda_2, \lambda_3\}$  being the spectrum of  $\mathbf{A}$ .

---

## Multiple Minimizations Needed

---

Must solve following at the vertices of a finite element mesh: Can normalize using division by  $f(x)$ .

$$\min_{\mathbf{A} \in \mathbf{E}_1} \text{trace}[\mathbf{A}^T (\mathbf{A} - 2\mathbf{B})] = \min_{\mathbf{A} \in \mathbf{E}_1} \text{trace}[\mathbf{A}^2 - 2\mathbf{A}\mathbf{B}]$$

where

$$\mathbf{E}_1 = \{ \mathbf{A} \mid \mathbf{A} \in \mathbb{R}^{3 \times 3}, \mathbf{A} = \mathbf{A}^T, \\ \lambda_1 \lambda_2 + \lambda_2 \lambda_3 + \lambda_3 \lambda_1 = 1, \\ \lambda_1 + \lambda_3 > 0, \lambda_2 + \lambda_3 > 0, \lambda_3 + \lambda_1 > 0 \}$$

(with  $\{\lambda_1, \lambda_2, \lambda_3\}$  being the spectrum of  $\mathbf{A}$ )

---

## Minimization Problem - Matrix Form

---

### Problem Qmin

$$\min \text{trace}\{\mathbf{A}\mathbf{A} - 2\mathbf{B}\mathbf{A}\}$$

s.t.

$$\ell^T \mathbf{M} \ell = 2$$

$$\mathbf{M} \ell \geq \mathbf{0}$$

where

$$\mathbf{M} = \mathbf{e}\mathbf{e}^T - \mathbf{I}, \quad \text{with } \mathbf{e}^T = (1, 1, \dots, 1),$$

and

$$\mathbf{B} = \mathbf{B}^T \text{ is specified,}$$

$$\mathbf{A} = \mathbf{A}^T = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T,$$

$$\mathbf{\Lambda} = \text{diag}(\ell).$$

---

## Diagonalized Problem

---

Note that Problem Qmin is equivalent to

$$\min \text{trace}\{\Lambda\Lambda - 2\hat{\mathbf{B}}\Lambda\} = \min \ell^T \ell - 2\mathbf{b}^T \ell$$

s.t.

$$\ell^T \mathbf{M} \ell = 2$$

$$\mathbf{M} \ell \geq \mathbf{0}$$

where

$$\hat{\mathbf{B}} = \mathbf{Q}^T \mathbf{B} \mathbf{Q} \quad \text{and} \quad \mathbf{b} = \text{diag}(\hat{\mathbf{B}}).$$

---

## Constraints for $n = 3$

---

**Constraints:**

$$\ell^T \mathbf{M} \ell = 2$$

$$\mathbf{M} \ell \geq \mathbf{0}$$

$$\mathbf{M} = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

$$\ell^T \mathbf{M} \ell = 2 \Rightarrow \lambda_1 \lambda_2 + \lambda_2 \lambda_3 + \lambda_3 \lambda_1 = 1$$

$$\mathbf{M} \ell \geq \mathbf{0} \Rightarrow \begin{aligned} \lambda_2 + \lambda_3 &\geq 0 \\ \lambda_1 + \lambda_3 &\geq 0 \\ \lambda_1 + \lambda_2 &\geq 0 \end{aligned}$$

---

## No Active Equality Constraint

---

### Lemma

*If the vector  $\ell \in \mathbb{R}^n$  is finite and feasible, then none of the inequality constraints can be active. In other words,*

$$\ell^T \mathbf{M} \ell = 2 \quad \Rightarrow \quad \mathbf{e}_j^T \mathbf{M} \ell > 0, \text{ for } j = 1, 2, \dots, n.$$

---

## Proof Outline

---

Proof Outline:

If  $\lambda_2 + \lambda_3 + \dots + \lambda_n = 0$ ,

Equality constraint  $\ell^T \mathbf{M} \ell = 2$  provides

$$\begin{aligned} \lambda_1(\lambda_2 + \lambda_3 + \dots + \lambda_n) &= 2 & - & \lambda_2(\lambda_3 + \lambda_4 + \dots + \lambda_n) \\ & & - & \lambda_3(\lambda_2 + \lambda_4 + \lambda_5 + \dots + \lambda_n) \\ & & \dots & \\ & & - & \lambda_n(\lambda_2 + \lambda_3 + \dots + \lambda_{n-1}). \end{aligned}$$

It follows that

$$0 = \lambda_1(\lambda_2 + \lambda_3 + \dots + \lambda_n) = 2 + \lambda_2^2 + \lambda_3^2 \dots + \lambda_n^2 \geq 2$$

which is a contradiction.



---

## Lagrangian for Equality Constraint:

---

Lagrangian:

$$\mathcal{L}(\ell, \mu) = \ell^T \ell - 2\mathbf{b}^T \ell + \mu(\ell^T \mathbf{M} \ell - 2).$$

Setting the grad of Lagrangian to zero gives:

$$(\mathbf{I} + \mu \mathbf{M})\ell = \mathbf{b}.$$

If  $1/\mu$  is not an eigenvalue of  $-\mathbf{M}$  then the equality constraint becomes

$$\mathbf{b}^T (\mathbf{I} + \mu \mathbf{M})^{-1} \mathbf{M} (\mathbf{I} + \mu \mathbf{M})^{-1} \mathbf{b} = 2$$

---

## Eigensystem of $\mathbf{M}$

---

Helps to know eigensystem of  $\mathbf{M} = \mathbf{e}\mathbf{e}^T - \mathbf{I}$  :

Eigenvalues

$$\omega_1 = n - 1 \quad \text{and} \quad \omega_2 = -1, \quad \text{multiplicity } n - 1$$

Eigenvector for  $\omega_1$  is  $\mathbf{e}$

Eigenvector Matrix

$$\mathbf{U} \equiv (\mathbf{I} - 2\mathbf{w}\mathbf{w}^T), \quad \text{with } \mathbf{w} = (\mathbf{e} + \sqrt{n}\mathbf{e}_1) / \|\mathbf{e} + \sqrt{n}\mathbf{e}_1\|,$$

Easily checked:

$$\mathbf{U}^T \mathbf{M} \mathbf{U} = \mathbf{U} \mathbf{M} \mathbf{U}^T = \mathbf{U} \mathbf{e} \mathbf{e}^T \mathbf{U}^T - \mathbf{I} = n \mathbf{e}_1 \mathbf{e}_1^T - \mathbf{I}.$$

---

## The Secular Equation

---

$$\frac{\beta_1^2 \omega_1}{(1 + \mu \omega_1)^2} = 2 + \frac{\beta_2^2}{(1 - \mu)^2}$$

where  $(\beta_1, \mathbf{b}_2^T) = \mathbf{b}^T \mathbf{U}$  and  $\beta_2^2 = \mathbf{b}_2^T \mathbf{b}_2$ .

Note:

$$\beta_1 = \mathbf{e}^t \mathbf{b} / \sqrt{n} \text{ is invariant: } \mathbf{e}^T \mathbf{b} = \text{trace}\{\mathbf{B}\}$$

---

## *Reciprocal Square Root Equation*

---

Better Equivalent Form

$$\pm(1 + \mu\omega_1) = \frac{(1 - \mu)|\beta_1|\sqrt{\omega_1}}{\sqrt{2(1 - \mu)^2 + \beta_2^2}}.$$

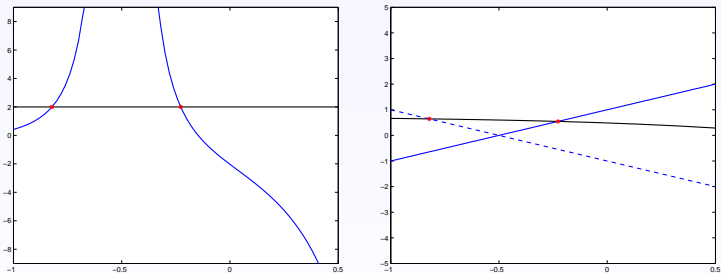
Algorithm: Newton's Method

converges in 3 steps

---

## Graphs of Secular Equations

---



**Figure:** The Secular Equation for Multiplier  $\mu$  (left) and the Reciprocal Square Root Secular Equation (right)

---

## Choice of Root

---

Corresponding to Root  $\mu$ , the vector  $l$  is:

$$l = \mathbf{U}\mathbf{c} \quad \text{with} \quad \mathbf{c}^T = \left( \frac{\beta_1}{(1 + \mu\omega_1)}, \frac{1}{1 - \mu} \mathbf{b}_2^T \right).$$

Pick Solution with Positive Components:

$$0 < \mathbf{M}l = (\mathbf{e}\mathbf{e}^T - \mathbf{I})l = \mathbf{e}(\mathbf{e}^T l) - l$$

---

## Suggested Alternating Min Algorithm

---

- ▶  $\mathbf{b} = \text{diag}(\mathbf{B}); \quad \mathbf{Q} = \mathbf{I};$
- ▶ while ('not converged'),
  - ▶  $\min_{\ell} \ell^T \ell - 2\mathbf{b}^T \ell$  s.t. constraints ;
  - ▶  $\min_{\mathbf{A}=\mathbf{W}\text{diag}(\ell)\mathbf{W}^T} \text{trace}\{\mathbf{A}\mathbf{A} - 2\mathbf{B}\mathbf{A}\}$  s.t. constraints;
  - ▶  $\mathbf{b} = \text{diag}(\mathbf{W}^T \mathbf{B} \mathbf{W});$

ALWAYS CONVERGED IN 2 STEPS !

---

## Surprise Result

---

Suppose  $\mathbf{B} = \mathbf{Q}\text{diag}(\mathbf{b})\mathbf{Q}^T$

If  $\mu$  and corresponding  $\ell = (\mathbf{I} + \mu\mathbf{M})^{-1}\mathbf{b}$  solve

$$\min \ell^T \ell - 2\mathbf{b}^T \ell \quad \text{s.t. constraints}$$

We have

**Lemma**

Let  $\Lambda = \text{diag}(\ell)$ . Then  $\mathbf{A}_Q = \mathbf{Q}\Lambda\mathbf{Q}^T$  solves

$$\min \text{trace}\{\mathbf{A}\mathbf{A} - 2\mathbf{B}\mathbf{A}\} \quad \text{s.t. constraints}$$

over all  $\mathbf{A} = \mathbf{W}\Lambda\mathbf{W}^T$  with  $\mathbf{W}^T\mathbf{W} = \mathbf{I}$ .



---

## Proof Outline

---

If  $\hat{\mathbf{b}} = \mathbf{Q}^T \mathbf{B} \mathbf{Q}$  then  $\mathbf{W}^T \mathbf{B} \mathbf{W} = \hat{\mathbf{Q}}^T \hat{\mathbf{B}} \hat{\mathbf{Q}}$ .

Let  $\hat{\mathbf{b}} = \text{diag}(\hat{\mathbf{Q}}^T \hat{\mathbf{B}} \hat{\mathbf{Q}})$ .

$$\hat{\beta}_j \equiv \hat{\mathbf{b}}(j) = \mathbf{q}_j^T \hat{\mathbf{B}} \mathbf{q}_j = \sum_{i=1}^n \beta_i \gamma_{ij}^2,$$

where  $\beta_j = \mathbf{b}(j)$ ,  $\hat{\mathbf{Q}} = (\gamma_{ij})$ ,  $\mathbf{q}_j = \hat{\mathbf{Q}} \mathbf{e}_j$

Hence,

$$\hat{\mathbf{b}} = \mathbf{G}^T \mathbf{b}, \quad \text{where the } i, j \text{ - th entry of } \mathbf{G} \text{ is } \gamma_{ij}^2.$$

We show

$$\ell^T \ell - 2\mathbf{b}^T \ell \leq \ell^T \ell - 2\hat{\mathbf{b}}^T \ell.$$

Since  $\ell$  is fixed, sufficient to show

$$\ell^T \hat{\mathbf{b}} - \ell^T \mathbf{b} = \ell^T (\hat{\mathbf{b}} - \mathbf{b}) \leq 0.$$

---

## Proof Outline Contd.

---

$\mathbf{G}\mathbf{e} = \mathbf{G}^T\mathbf{e} = \mathbf{e}$  implies

$$\hat{\mathbf{b}} - \mathbf{b} = (1 - \mu)(\mathbf{G}^T - \mathbf{I})\ell,$$

Therefore,

$$\ell^T(\hat{\mathbf{b}} - \mathbf{b}) = (1 - \mu)\frac{1}{2}(\ell^T(\mathbf{G}^T - \mathbf{I})\ell + \ell^T(\mathbf{G} - \mathbf{I})\ell)$$

It follows that

$$\ell^T(\hat{\mathbf{b}} - \mathbf{b}) = -(1 - \mu) \sum_{i \neq j} \gamma_{ij}^2 (\lambda_j - \lambda_i)^2 \leq 0,$$

since  $(1 - \mu) > 0$ .

---

## Converse

---

### Lemma

Suppose  $\mathbf{A} = \mathbf{W}\mathbf{\Lambda}\mathbf{W}^T$  solves Problem Qmin. Then there is an orthogonal  $\hat{\mathbf{Q}}$  such that  $\mathbf{Q} = \mathbf{W}\hat{\mathbf{Q}}$  diagonalizes  $\mathbf{B}$  and  $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T$ . In other words, if  $\mathbf{A}$  solves Problem Qmin, then  $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T$  with  $\mathbf{Q}^T\mathbf{B}\mathbf{Q}$  diagonal.

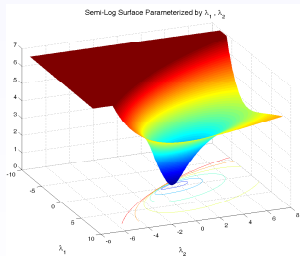
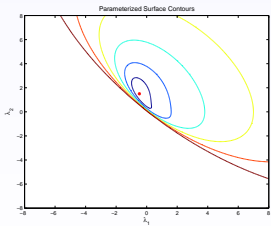
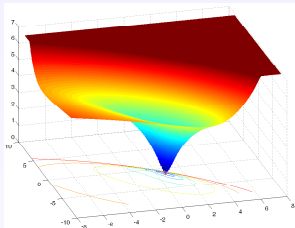
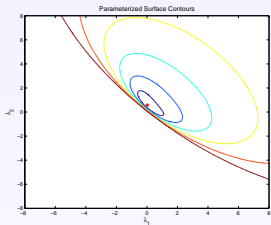
---

## Final Algorithm

---

- ▶  $[\mathbf{Q}, \mathbf{L}] = \text{eig}(\mathbf{B}); \quad \mathbf{b} = \text{diag}(\mathbf{L})$
- ▶  $\min_{\ell} \ell^T \ell - 2\mathbf{b}^T \ell \quad \text{subject to constraints}$
- ▶  $\mathbf{A} = \mathbf{Q} \text{diag}(\ell) \mathbf{Q}^T;$

Works for ANY  $n$



---

## Summary

---

- ▶ Monge-Ampère important to Diff Geometry
- ▶ Glowinski - Dean algorithm requires MANY eigenvalue constrained min problems
- ▶ We provided a very simple and efficient method with some surprising properties.

Report:

**A Quadratically Constrained Minimization Problem Arising from PDE of Monge-Ampère Type**

CAAM TR08-02, DCS and R. Glowinski

---

## References

---

:

- ▶ E.J. Dean and R. Glowinski, Numerical solution of the two-dimensional Monge-Ampère equation with Dirichlet boundary conditions: a least-squares approach, *C. R. Acad. Sci. Paris, Ser. I*, **339**(12), 887-892, (2004).
- ▶ E.J. Dean and R. Glowinski, On the numerical solution of a two-dimensional Pucci's equation with Dirichlet boundary conditions: a least-squares approach, *C. R. Acad. Sci. Paris, Serie I*, **341**, 375-380, (2005).
- ▶ R. Glowinski, E.J. Dean, G. Guidoboni, L.H. Juarez and T.W. Pan, Applications of operator-splitting methods to the direct numerical simulation of particulate and free-surface flows and to the numerical solution of the two- dimensional Monge-Ampère equation, *Jap. J. Industr. Appl. Math.*, **25**, 1-63, (2008).

---

## References

---

:

- ▶ D.B.Fairlie and A.N. Leznov, General solutions of the Monge-Ampère equation in n-dimensional space, *Journal of Geometry and Physics*, **16** , 385-390, (1995).
- ▶ D.B.Fairlie and A.N. Leznov, The General Solution of the Complex Monge-Ampère Equation in two dimensional space, preprint August (1999),  
<http://arxiv.org/abs/solv-int/9909014/> (accessed 27 May 08).
- ▶ BIRS Workshop (2003) on Monge-Ampère Equation  
<http://www.birs.ca/workshops/2003/03w5067/report03w5067.pdf>