

On the Forsythe conjecture

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joint work with

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George E. Forsythe

Godfather of “Computer Science”



1917–1972

- National Bureau of Standards (1948), Standards Western Automatic Comp.
- Stanford University (1957), founded Computer science department (1965). He hired Gene H. Golub in 1962.
- *“It is generally agreed that he, more than any other man, is responsible for the rapid development of computer science in the world’s colleges and universities.”*
[Donald Knuth]
- 17 Ph.D., e.g. Beresford N. Parlett, Cleve B. Moler, James M. Varah, Richard P. Brent, J. Alan George.

Problem

Problem

A is symmetric and positive definite, b given

Minimize

$$f(x) = \frac{1}{2}x^T Ax - x^T b$$

using the **steepest descent** method

```
input  $A, b, x_0$   
for  $k = 0, 1, 2, \dots$  do  
  
     $g_k = Ax_k - b$   
     $x_{k+1} = x_k - \alpha_k g_k$   
  
end for
```

Asymptotic behavior of normalized gradients?

Asymptotic behavior

Forsythe and Motzkin conjecture

- Consider the **steepest descent** method and denote

$$v_k \equiv \frac{g_k}{\|g_k\|}.$$

Note that $v_k \perp v_{k+1}$.

- [Forsythe & Motzkin, 1951] conjectured that vectors v_k asymptotically alternate between **two directions**,

$$v_{2k} \rightarrow v, \quad v_{2k+1} \rightarrow w.$$

- [Akaike 1959]: Proof using methods from probability theory.
[Forsythe 1968]: Algebraic proof and generalization.
- [Zou & Magoulés, 2022, SIREV]: Still of interest in optimization.

Problem

Forsythe 1968

Minimize

$$f(x) = \frac{1}{2} x^T A x - x^T b$$

using the ***s*-gradient method**:

```
input  $A, b, x_0$ 
```

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

$$g_k = Ax_k - b$$

$$x_{k+1} = \arg \min_{y \in \mathcal{K}_s(A, g_k)} f(x_k + y)$$

```
end for
```

This is nothing but restarted CG \rightarrow CG(*s*).

Asymptotic behavior

and the Forsythe conjecture

- Consider the $\text{CG}(s)$ method applied to $Ax = b$, $s > 1$.
- Let x_0 be such that $d(A, g_0) > s$. Then

$$v_k \equiv \frac{g_k}{\|g_k\|}$$

are well defined.

Forsythe's conjecture

Vectors v_k asymptotically alternate between **two directions**,

$$v_{2k} \rightarrow v, \quad v_{2k+1} \rightarrow w.$$

- Observation: v_k are the Lanczos vectors and $v_k \perp v_{k+1}$.

Arnoldi projection of v

with respect to A and s

- $A \in \mathbb{R}^{n \times n}$, $v \in \mathbb{R}^n$, and $s \geq 1$, we define $w \in \mathbb{R}^n$:

$$w \in \underbrace{A^s v + \mathcal{K}_s(A, v)}_{p(A)v} \quad \text{and} \quad w \perp \mathcal{K}_s(A, v).$$

- $w \neq 0$ is unique if $d(A, v) > s$, denote

$$w = P_s(A; v) v.$$

- w can be computed using the **Lanczos** algorithm (if A is symmetric) or the **Arnoldi** algorithm.
- Note that $P_s(A; v)$ is independent of scaling of v .

A more general formulation

of the Forsythe conjecture via the Lanczos (Arnoldi) process

- $A \in \mathbb{R}^{n \times n}$ **symmetric**, $v \in \mathbb{R}^n$ with $d(A, v) > s \geq 1$
- **Conjecture:** Consider the algorithm

$$w_0 = v$$

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

$$v_k = w_k / \|w_k\|$$

$$w_{k+1} = P_s(A; v_k) v_k$$

end for

Then the sequence $\{v_{2k}\}$ has a **single limit** vector.

- The vectors v_k are well defined.

Symmetric matrices

Norms of w_k

It holds that

$$w_{k+1} = P_s(A; v_k) v_k$$

and

$$\|w_{k+1}\| = \min_{p \in \mathcal{M}_s} \|p(A)v_k\| \leq \min_{p \in \mathcal{M}_s} \|p(A)\|.$$

Theorem

It holds that

$$\|w_k\| \leq \|w_{k+1}\| \quad k = 0, 1, 2, \dots$$

with equality iff $v_k = v_{k+2}$.

Consequence:

$$\|w_k\| \longrightarrow \tau \quad \text{as } k \rightarrow \infty.$$

Distance between v_{k+2} and v_k

$$\begin{aligned}1 - \frac{1}{2} \|v_{k+2} - v_k\|^2 &= \langle v_{k+2}, v_k \rangle = \frac{1}{\|w_{k+2}\|} \langle w_{k+2}, v_k \rangle \\&= \frac{1}{\|w_{k+2}\|} \langle P_s(A; v_{k+1}) v_{k+1}, v_k \rangle \\&= \frac{1}{\|w_{k+2}\|} \langle v_{k+1}, P_s(A; v_{k+1}) v_k \rangle \\&= \frac{1}{\|w_{k+2}\|} \langle v_{k+1}, A^s v_k \rangle \\&= \frac{1}{\|w_{k+2}\|} \langle v_{k+1}, \underbrace{P_s(A; v_k) v_k}_{w_{k+1}} \rangle \\&= \frac{\|w_{k+1}\|}{\|w_{k+2}\|} \rightarrow 1\end{aligned}$$

A short summary

$A \in \mathbb{R}^{n \times n}$ **symmetric**, $v \in \mathbb{R}^n$ with $d(A, v) > s \geq 1$

$$w_0 = v$$

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

$$v_k = w_k / \|w_k\|$$

$$w_{k+1} = P_s(A; v_k) v_k$$

end for

We know that

$$\|w_k\| \leq \|w_{k+1}\|, \quad \|w_k\| \rightarrow \tau,$$

and

$$\|v_{k+2} - v_k\| \rightarrow 0.$$

Bolzano-Weierstraß $\rightarrow \{v_{2k}\}$ has a convergent subsequence.

Example

- The property

$$\|v_{k+2} - v_k\| \rightarrow 0$$

is **not sufficient** for the existence of a **single** limit vector.

- Complex points

$$\mu_k = e^{i\omega_k}, \quad \omega_k = \sum_{j=1}^k \frac{\pi}{j}$$

satisfy $|\mu_k - \mu_{k+1}| \rightarrow 0$, but $\{\mu_k\}$ does not converge.

- It may be **difficult** to find a counterexample **numerically**.

The set of limit vectors

- Let Σ^A be the set of unit norm vectors such that $d(A, v) > s$.
- Define the **transformation** $T_A : \Sigma^A \rightarrow \Sigma^A$

$$v \mapsto T_A(v) \equiv \frac{P_s(A; v) v}{\|P_s(A; v) v\|}$$

so that

$$v_{k+2} = T_A(T_A(v_k)).$$

- $T_A \circ T_A : \Sigma^A \rightarrow \Sigma^A$ is well defined and **continuous**.

Theorem

The set Σ_*^A of limit vectors of the sequence $\{v_{2k}\}$ satisfies:

- (1) Σ_*^A is a **closed** and **connected** set in \mathbb{R}^n .
- (2) $\Sigma_*^A \subseteq \Sigma^A$, and each $v_* \in \Sigma_*^A$ satisfies $v_* = T_A(T_A(v_*))$.

Degree of limit vectors v_*

$A \in \mathbb{R}^{n \times n}$ **symmetric**, $v \in \mathbb{R}^n$ with $d(A, v) > s \geq 1$

$w_0 = v$

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

$$v_{k+1} = T_A(v_k)$$

end for

Theorem

Each limit vector v_* of $\{v_{2k}\}$ satisfies

$$s < d(A, v_*) \leq 2s.$$

Proof based on $v_* = T_A(T_A(v_*))$.

The case $s = 1$

Without loss of generality A is diagonal

- \forall limit vector v_* of $\{v_{2k}\}$ we have $d(A, v_*) = 2$,

$$v_* = \alpha e_i + \beta e_j$$

for some canonical basis vectors e_i and e_j , $\alpha\beta \neq 0$, and

$$\tau = \left\| Av_* - (v_*^T Av_*) v_* \right\|$$

giving

$$\tau^2 = \alpha^2 (1 - \alpha^2) (\lambda_i - \lambda_j)^2. \quad (*)$$

- **Finitely many** combinations of distinct $i, j \in \{1, 2, \dots, n\}$, for each such combination **finitely many** values of α satisfying $(*)$.
- Σ_*^A is **connected** \Rightarrow there is just **one** limit vector.

The same approach

does not work for $s = 2$

- $\tau = \|P_s(A; v)v\|$ gives

$$\tau^2 = v^T A^4 v + \frac{(v^T A^3 v)^2 - (v^T A^2 v)^3}{(v^T A v)^2 - v^T A^2 v}.$$

- $d(A, v_*) = 3$ or $d(A, v_*) = 4$:

$$v_* = \alpha e_i + \beta e_j + \gamma e_\ell$$

$$\alpha^2 + \beta^2 + \gamma^2 = 1.$$

- **One** nonlinear equation with **two** degrees of freedom.
- **Infinitely many** solutions.

The case $s = 2$

$$\|w_{k+1}\| v_{k+1} = w_{k+1} = P_s(A; v_k)v_k$$

so that

$$\underbrace{\|w_{k+1}\| \|w_{k+2}\|}_{\rightarrow \tau^2} v_{k+2} = \underbrace{P_s(A; v_{k+1})P_s(A; v_k)}_{Q_{2s}(A; v_k)} v_k$$

and each limit vector v_* of $\{v_{2k}\}$ satisfies

$$\tau^2 v_* = Q_{2s}(A; v_*)v_*,$$

where v_* has either 3 or 4 nonzero components.

The case $s = 2$

and results of [Zhuk and Bondarenko, 1983]

- If v_* has **4 nonzero** components with indexes i_1, \dots, i_4 , then

$$\tau^2 = Q_{2s}(\lambda_{i_j}; v_*), \quad j = 1, \dots, 4.$$

4 interpolation conditions $\rightarrow Q_{2s}$ is determined **uniquely**.

- If v_* has **3 nonzero** components and if A is positive definite, then Q_{2s} is again **unique**. [Zhuk and Bondarenko, 1983]
- Finitely many combinations of sets of $i_j \in \{1, 2, \dots, n\} \Rightarrow$ **finitely many** polynomials Q_{2s} that correspond to v_* 's.
- Quoting [Zabolotskaya, 1979] they use as a proven fact that the convergence of the coefficients of Q_{2s} implies the existence of a **single limit vector** v_* .
- We consider the case $s = 2$ to be still open.

Nonsymmetric matrices

Worst-case GMRES

and the cross equality

- For a given s , there exists a unit norm vector b such that

$$\|r\| = \min_{p \in \pi_s} \|p(A)b\| = \max_{\|v\|=1} \min_{p \in \pi_s} \|p(A)v\|.$$

Theorem

[Zavorin '02; Faber, Liesen, T. '13]

If b is a worst-case GMRES initial vector for A and s , then

$$b \xrightarrow{\text{GMRES}(A, b, s)} r \xrightarrow{\text{GMRES}(A^T, r, s)} \|r\|^2 b.$$

- We say that b satisfies the **cross equality** for A and s if

$$b \xrightarrow{\text{GMRES}(A, b, s)} r \xrightarrow{\text{GMRES}(A^T, r, s)} z \in \text{span}\{b\}.$$

GMRES Cross iteration algorithm

and the Forsythe conjecture [Faber, Liesen, T., 2013]

Given A , s , and b , it seems that b_k **converge** to a vector satisfying the **cross equality** for A and s :

```
 $b_0 = b$   
for  $k = 1, 2, \dots$  do  
  
     $r_k = \text{GMRES}(A, b_{k-1}, s)$   
     $c_k = r_k / \|r_k\|$   
     $z_k = \text{GMRES}(A^T, c_k, s)$   
     $b_k = z_k / \|z_k\|$   
  
end for
```

$$\|r_k\| \leq \|z_k\| \leq \|r_{k+1}\| \leq \|z_{k+1}\|$$

The algorithm does **not** find a **worst-case** initial vector in general.

Arnoldi Cross iteration algorithm

and generalization of the Forsythe conjecture for nonsymmetric matrices

Given $A \in \mathbb{R}^{n \times n}$, $v \in \mathbb{R}^n$ such that $d(A, v) > s \geq 1$

$$w_0 = v$$

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

$$v_k = w_k / \|w_k\|$$

$$w_{k+1} = P_s(B; v_k) v_k$$

end for

where $B = A$ (for k even), $B = A^T$ (for k odd).

Conjecture

The subsequence $\{v_{2k}\}$ has a **single limit** vector.

Results [Faber, Liesen, T., 2023]

for nonsymmetric matrices

$$\|w_k\| \leq \|w_{k+1}\| \quad \text{and} \quad \|v_{k+2} - v_k\| \rightarrow 0.$$

$$T_A(v) \equiv \frac{P_s(A; v) v}{\|P_s(A; v) v\|}$$

Theorem

The set Σ_*^A of limit vectors of the sequence $\{v_{2k}\}$ satisfies:

- (1) Σ_*^A is a **closed** and **connected** set in \mathbb{R}^n .
- (2) $\Sigma_*^A \subseteq \Sigma^A$, and each $v_* \in \Sigma_*^A$ satisfies $v_* = T_{A^T}(T_A(v_*))$.

It holds that $s < d(A, v_*)$, but it **does not** hold in general that

$$d(A, v_*) \leq 2s.$$

Orthogonal matrices

Arnoldi Cross Iteration

for orthogonal matrices and $s = 1$

Given $A \in \mathbb{R}^{n \times n}$ **orthogonal**, $v \in \mathbb{R}^n$ such that $d(A, v) > s = 1$

$$w_0 = v$$

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

$$v_k = w_k / \|w_k\|$$

$$\alpha_k = v_k^T A v_k$$

$$w_{k+1} = (B - \alpha_k I)v_k$$

end for

where $B = A$ (for k even), $B = A^T$ (for k odd).

$$\|w_{k+1}\|^2 = 1 - \alpha_k^2 \Rightarrow |\alpha_k| \geq |\alpha_{k+1}|.$$

Without loss of generality

A is block diagonal

$A \in \mathbb{R}^{n \times n}$ can be orthogonally block-diagonalized $A = UGU^T$
with U **orthogonal** and

$$G = \begin{bmatrix} G_1 & & & & & & & \\ & \ddots & & & & & & \\ & & G_m & & & & & \\ & & & [\pm 1] & & & & \\ & & & & \ddots & & & \\ & & & & & [\pm 1] & & \\ & & & & & & \ddots & \\ & & & & & & & [\pm 1] \end{bmatrix}$$

where

$$G_j = \begin{bmatrix} c_j & s_j \\ -s_j & c_j \end{bmatrix}$$

with $c_j^2 + s_j^2 = 1$ and $s_j \neq 0$.

Convergence

for orthogonal matrices and $s = 1$

For simplicity

$$A = \begin{bmatrix} G_1 & & & \\ & G_2 & & \\ & & \ddots & \\ & & & G_m \end{bmatrix}, \quad v_k = \begin{bmatrix} v_k^{(1)} \\ v_k^{(2)} \\ \vdots \\ v_k^{(m)} \end{bmatrix}, \quad v_k^{(j)} \in \mathbb{R}^2.$$

Lemma

[Faber, Liesen, T., 2023]

Let $0 < c_1 < \dots < c_m$ and $d(A, v) > 1$ and $v^{(1)} \neq 0$.

For k sufficiently large there exists $0 < \varrho < 1$ such that

$$\|v_{2k+2}^{(j)}\| \leq \varrho \|v_{2k}^{(j)}\|, \quad j = 2, \dots, m,$$

and

$$\|v_{2k+2}^{(1)} - v_{2k}^{(1)}\| \leq \varrho^k.$$

Convergence result

Orthogonal matrices, $s = 1$

Theorem

[Faber, Liesen, T., 2023]

Let $0 < c_1 < \dots < c_m$ and $d(A, v) > 1$ and $v^{(1)} \neq 0$.

Then the sequence $\{v_{2k}\}$ converges to a **single limit vector**.

Proof. Using the previous

$$\|v_{2k+2} - v_{2k}\|^2 = \sum_{j=1}^m \|v_{2k+2}^{(j)} - v_{2k}^{(j)}\|^2 \leq 3 \varrho^{2k}$$

which implies

$$\sum_{k=0}^{\infty} \|v_{2k+2} - v_{2k}\| < \infty.$$

Connection to worst-case Arnoldi problem

Orthogonal matrices, $s = 1$

$$\begin{aligned}\max_{\|v\|=1} \min_{\alpha \in \mathbb{R}} \|Av - \alpha v\|^2 &= \max_{\|v\|=1} \|Av - \langle v, Av \rangle v\|^2 \\ &= 1 - \min_{\|v\|=1} \langle v, Av \rangle^2\end{aligned}$$

and the optimal α_* is given by

$$\alpha_* = \min_{\|v\|=1} |\langle v, Av \rangle| = \min_{z \in F(A)} |z| = c_1.$$

We can prove that α_k in the Cross Iteration algorithm satisfy

$$\lim_{k \rightarrow \infty} \alpha_k = c_1.$$

Hence, Cross Iteration algorithm finds a **worst-case vector**.

Conclusions

- We revised Forsythe's results and **generalized** them for symmetric and nonsymmetric matrices.
- **Conjecture** for symmetric and nonsymmetric matrices.
- For $s = 1$, we proved the existence of a **single limit** vector of the sequence $\{v_{2k}\}$ for symmetric and orthogonal matrices.
- We proved several new results about the limiting behavior of the sequence $\{v_{2k}\}$, but the conjecture still **remains open**.

Related papers

V. Faber, J. Liesen and P. Tichý, [On the Forsythe conjecture, submitted to SIMAX, 2022: <https://arxiv.org/abs/2209.14579>.]

- M. Afanasjew, M. Eiermann, O. G. Ernst, and S. Güttel, [A generalization of the steepest descent method for matrix functions, *Electron. Trans. Numer. Anal.*, 28 (2007/08), pp. 206-222.]
- V. Faber, J. Liesen and P. Tichý, [Properties of worst-case GMRES, *SIAM J. Matrix Anal. Appl.*, 34 (2013), pp. 1500-1519.]
- G. E. Forsythe, [On the asymptotic directions of the s-dimensional optimum gradient method, *Numer. Math.*, 11 (1968), pp. 57-76.]
- P. F. Zhuk and L. N. Bondarenko, [A conjecture of G. E. Forsythe, *Mat. Sb. (N.S.)*, 121(163) (1983), pp. 435-453.]

Thank you for your attention!