

# Zyzyva: The Symmetric Tridiagonal Eigenproblem

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# Problem Definition

Given

$$T = \begin{bmatrix} a_1 & b_1 & & & & & \\ b_1 & a_2 & b_2 & & & & \\ & b_2 & \cdot & \cdot & & & \\ & & \cdot & \cdot & \cdot & & \\ & & & \cdot & a_{n-1} & b_{n-1} & \\ & & & & b_{n-1} & a_n & \end{bmatrix},$$

Solve

$$Tv = \lambda v$$

for  $\lambda$  and  $v \neq 0$ .

This is a **central** problem in any symmetric eigenproblem.

Requirements in finite precision :

- $\|T\hat{v}_i - \hat{\lambda}_i\hat{v}_i\| = O(\varepsilon\|T\|), \quad i = 1, 2, \dots, n$
- $(\hat{v}_i, \hat{v}_j) = O(\varepsilon), \quad i \neq j.$

# History

- Jacobi [1846]
- Bargmann, Montgomery and von Neumann [1946]
- Lanczos [1950], Givens [1954], Householder [1958]
- Wielandt, Wilkinson [1940/1950s], perhaps earlier
- Francis/Kublanovskaja [1961]
- Golub and Kahan [1965]
- Wilkinson [1965]
- Cuppen [1981], Gu and Eisenstat [1995]
- All algorithms cost  $O(n^3)$  in the worst case

# Ideal Tridiagonal Eigensolver

## DESIRED GOALS:

- **Minimum output complexity** —  $O(n)$  per eigenvector
- **Provably numerically accurate**
- Each eigenpair is **independently computable**

# Difficulties

- Problem: Given  $\hat{\lambda}$ , solve

$$(T - \hat{\lambda}I)x \approx 0.$$

- Inverse iteration: Given  $\hat{\lambda}$ , compute:

$$(T - \hat{\lambda}I)x_{i+1} = x_i, \quad i = 0, 1, 2, \dots$$

- Costs  $O(n)$  per iteration
- Typically, 1-3 iterations are enough
- **BUT**, inverse iteration only guarantees

$$\|T\hat{v} - \hat{\lambda}\hat{v}\| = O(\varepsilon\|T\|)$$

# Fundamental Limitations

## Gap Theorem :

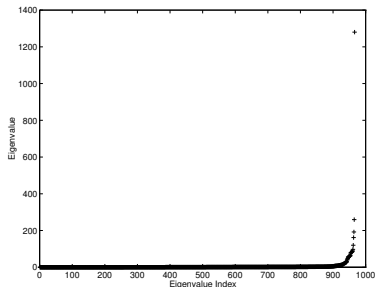
$$\sin \angle(v, \hat{v}) \leq \frac{\|T\hat{v} - \hat{\lambda}\hat{v}\|}{\text{Gap}(\hat{\lambda})}.$$

Gap( $\hat{\lambda}$ ) can be small :

$$\begin{bmatrix} 1 & \varepsilon_1 & & \\ \varepsilon_1 & 1 & \varepsilon_2 & \\ & \varepsilon_2 & 1 & \varepsilon_3 \\ & & \varepsilon_3 & 1 \end{bmatrix}$$

When eigenvalues are close, independently computed eigenvectors **WILL NOT** be mutually orthogonal.

# Eigenvalues of Biphenyl Matrix



- LAPACK — one big “cluster”  $\lambda_1, \lambda_2, \dots, \lambda_{939}$ .
- Tridiagonal solution takes 80% of Total Time.

# Difficulties

- All eigenvalues of  $T$  are easily computed in  $O(n^2)$  time.
  - QR, Root-free QR (Pal, Walker & Kahan), Bisection + Laguerre
- But eigenvector computation costs  $O(n^3)$ 
  - QR was slow
  - D&C resulted in inaccurate eigenvectors
  - Inverse Iteration expensive when eigenvalues were clustered

# Developments at Berkeley

- Ming Gu solves D&C eigenvector problem and comes to Berkeley
- Jeff Rutter writes D&C software to add to LAPACK
- Jim Demmel makes observation:
  - D&C fast when eigenvalues cluster
  - Inverse Iteration fast when eigenvalues are isolated
- Modest GOAL: Combine them to get faster software



# Initial Ideas

- Handling Clusters
  - Different twists for different eigenvalues in the cluster
- Ming Gu produced  $5 \times 5$  counter-example:
  - Leftmost  $\lambda \approx 1$
  - 3 clustered (not tightly) around 2
  - Rightmost  $\lambda \approx 3$
- Struggled with it over many sessions on the computer

# The Major Breakthrough

- LAPACK test matrices with geometrically distributed eigenvalues
  - One type was breezing through
  - Another type was failing miserably
- High Relative Accuracy
  - Kahan (1966)
  - Demmel & Kahan (1985)
- But tridiagonals do not always determine eigenvalues to high relative accuracy
- REPLACE Tridiagonal by Bidiagonal
- Many dead-ends/false leads
- Representation Tree.....

# Software

- Osni Marques
- Christof Vömel's torture tests
- Software is now in LAPACK
- Zyzzyva??

# Early References

- Fernando [1995], Godunov et al. [1985], Parlett & Dhillon[1997]
- Strang (2001)
  - Formula for  $\text{Diag}(T^{-1})$  known in literature on Kalman Filtering

- Birthday Gift for Beresford?