


Solvers for Systems of Nonlinear Algebraic Equations; Where are we TODAY?



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Presentation Plan



- Introduction
- Problem description
- Search for the ultimate solver
 - algorithms and solvers
 - test problems
 - method
 - small problems
 - large problems
- Original problem revisited

Background (1)



- Work originated in early 1990's
- Computer simulation of the behavior of airplanes under action of atmospheric gusts
- Two possible approaches
 - method of harmonic (S. O. Rice)
 - method of filtration (N. Wiener and independently Y. A. Kchinchin)
- Method of filtration used in original study

Background (2)



- Result → system of nonlinear algebraic equations
 - central/computational/difficult part of the problem
 - “independent” of other parts (well-defined and self-contained)
 - no preexisting knowledge about the solution
 - no simple way to suggest a starting vector
 - no simple way to reduce the “search area”

The Avionics Problem → Filter Equation

- Impulsive characteristics of a non-recursive filter $h(t)$ as related to the correlation function $K(\tau)$ of the stochastic process $\{y(t)\}$ obtained via filtering of the input white noise $\{x(t)\}$:

- $$K(\tau) = E \left[\sum_{k=0}^N h(k)x(t-k) \sum_{k=0}^N h(n)x(t+\tau-n) \right]$$

Filter Equation (2)

- Since input $\{x(t)\}$ is a white noise we can rewrite the problem as:
- $$K(\tau) = \sum_{n=0}^N h(n)h(n+\tau) \quad \text{for } \tau=0,1,\dots,N$$
- In its explicit form we have a system of nonlinear algebraic equations

Explicit Problem Formulation

$$K(0) = h(0)h(0) + h(1)h(1) + h(2)h(2) + \dots + h(N)h(N)$$

$$K(1) = h(0)h(1) + h(1)h(2) + h(2)h(3) + \dots + h(N-1)h(N)$$

$$K(2) = h(0)h(2) + h(1)h(3) + h(2)h(4) + \dots + h(N-2)h(N)$$

...

$$K(N-2) = h(0)h(N-2) + h(1)h(N-1) + h(2)h(N)$$

$$K(N-1) = h(0)h(N-1) + h(1)h(N)$$

$$K(N) = h(0)h(N)$$

Initial Work (1)



- Requirements for success (engineering-estimate)
 - minimum $N = 512$
 - potentially $N = 1024$ (?)
 - reached $N = 64$ using modified Powell's Method (1994)
- Encountered problems
 - computer hardware
 - for $N = 64$ solution time 10 minutes on a PC-486
 - robustness of solution methods

Initial Work (2)



- Research re-started → questions:
 - how to solve the system for very (?) large N ?
 - how to select the starting vector?
- Search directions
 - to find (an) ultimate solver(s)
 - use modern hardware

Back to the Basics



- System of N nonlinear algebraic equations
- Large number of algorithms and implemented solvers
 - iterative methods
- How to find the best?
 - use existing/agreed on test problems
 - compare performance
- **NAÏVE!**

“Available” Algorithms



- Newton’s method
- Powell’s algorithm
- Brown’s method
- Secant method
- Bisection
- Steepest Descent
- Trust Region
- Line Search
- Continuation
- Homotopy
- Augmented Lagrangian
- Reduced-Gradient
- Tensor

“Available” Solvers (1)

| Solver | Algorithm | Source |
|---------------|-------------------------------------|-------------------------|
| Brown | Brown's method | In-house implementation |
| QuasiA | Hybrid Powell (Newton/Trust Region) | In-house implementation |
| CHABIS | Characteristic Bisection | ACM TOMS |
| SOS | Brown's method | NETLIB |
| HOMPACK | Homotopy | ACM TOMS/NETLIB |

"Available" Solvers (2)

| Solver | Algorithm | Source |
|---------------|---|--------------------|
| CONTIN | Continuation | ACM TOMS/NETLIB |
| HYBRD | Hybrid Powell (Newton/Trust Region) | NETLIB |
| TENSOLVE | Tensor/Line Search | NETLIB |
| LANCELOT | Projected Gradient | NEOS |
| MINOS | Reduced Gradient | NEOS |

Test Problems



- **LACK of an “all-agreed” test library !**
- Test set → 22 frequently used problems
 - some artificially generated (with properties not typical for real-life applications)
 - most popular → More Test Set
 - typically small systems → $N \leq 10$
 - only few can be extended to larger N
 - no problems with absolute value

22 Test Problems



- Rosenbrock's
- Powell singular
- Powell badly scaled
- Wood
- Helical Valley
- Watson*
- Chebyquad*
- Brown Almost-linear*
- Discrete Boundary Value*
- Discrete integral equation*
- Trigonometric*
- Variably dimensioned*
- Broyden tri-diagonal*
- Broyden banded*
- Exponential/Sine Function
- The Freudenstein-Roth Function
- Semiconductor Boundary Condition
- Brown Badly Scaled
- Powell singular Extended
- Rosenbrock Extended
- Matrix Square Root Problem
- Dennis, Gay, Vu Problem

Methodological Considerations (1)

- Default settings for all solvers used
 - “engineering” approach
 - solver as black-box software
 - controversial choice
- Test problems contain default starting vectors
 - additional starting vectors used
 - zero
 - one
 - random $[0,1]$

Methodological Considerations (2)



- Computational cost
 - number of iterations
 - number of function evaluations
 - time
- Two steps
 - problems in their default formulation
 - increasing the size of amenable problems

Results (1)



- **Easy test problems** (solved easily by all solvers)
 - Rosenbrock's
 - Helical valley
 - The Freudenstein-Roth Function
 - Powell badly scaled
 - Broyden banded function
- **USELESS(?!) as test problems**
- Time so short that practically immeasurable
- In house codes only slightly less efficient than library solvers
 - implementation is not the "important" issue

Results (2)



- Weakest solvers (solve only few problems outside of the easy five):
 - bisection
 - variations of Newton's method
- TENSOLVE results are slightly less accurate
- Homotopy should not be used as a black-box solver
 - requires proper problem mapping
 - makes it less of a "general-purpose" solver than others

Solution of Large Problems



- Test problems that can have the default number of equations increased
 - Watson
 - Brown Almost-linear
 - Discrete integral equation
 - Variably dimensioned
 - Broyden banded
 - Chebyquad
 - Discrete Boundary Value
 - Trigonometric
 - Broyden tri-diagonal
- Five solvers left after initial selection
 - CONTIN
 - HYBRD
 - MINOS
 - LANCELOT
 - TENSOLVE

Results



- Test problems can be divided into three groups (results for any tried starting vector):
 - Difficult Problems - solution only for small N (≤ 31)
 - Medium Difficult Problems - some solvers fail to reach $N=1000$
 - Easy Problems - all solvers reach $N=1000$

Difficult Problems

| Solver | Watson Max N | Chebyquad Max N |
|----------|-----------------|--------------------|
| CONTIN | 6 | 2 |
| HYBRD | 14 | 9 |
| TENSOLVE | 23 | 18 |
| LANCELOT | 31 | 18 |
| MINOS | 31 | 20 |

Difficult Problem Example

Watson Function



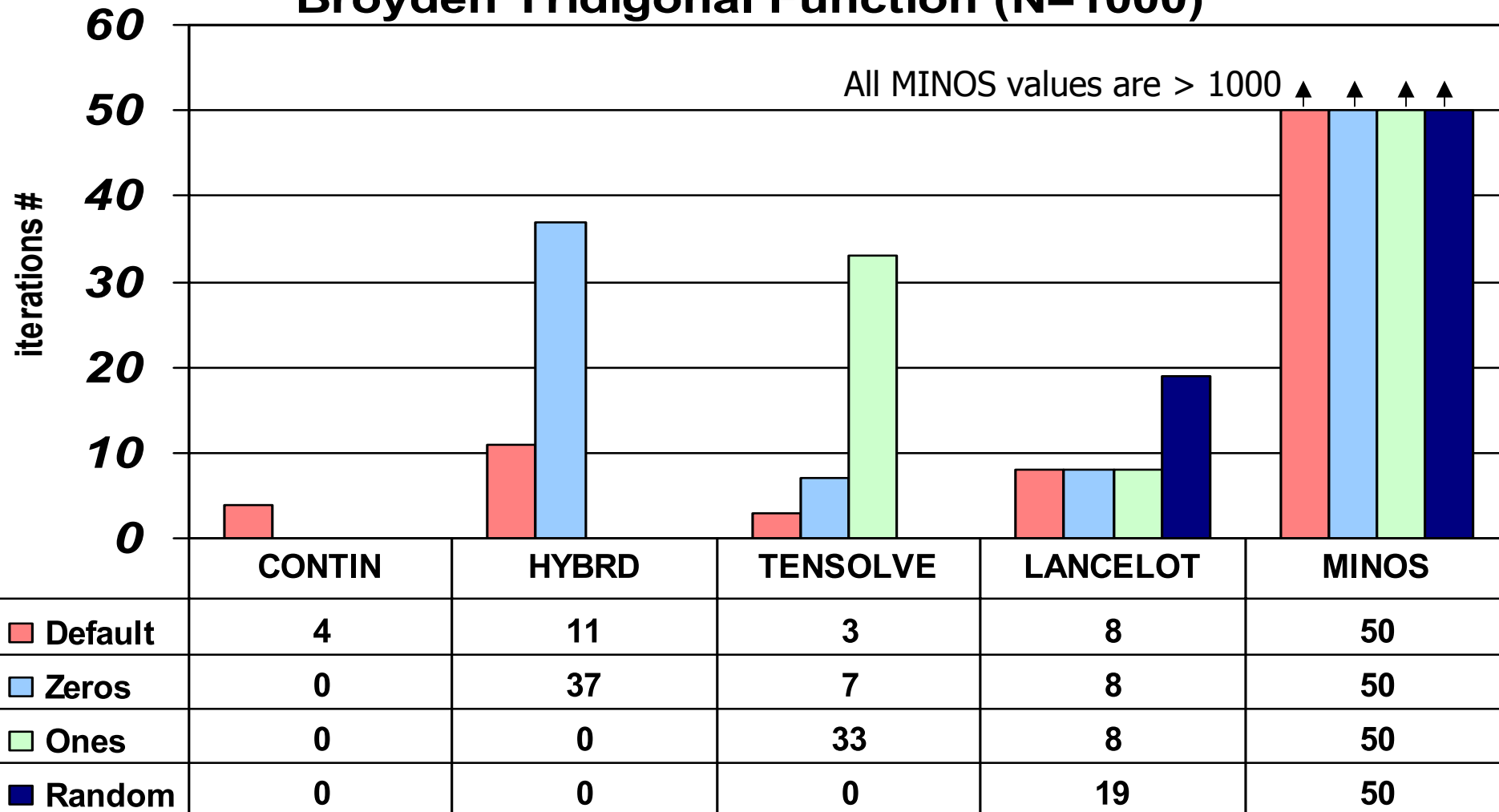
| Solver | Max N | IT | FC | Time/ Sec |
|---------------|--------------|-----------|-----------|----------------------|
| CONTIN | 6 | 6 | 256 | 0 |
| HYBRD | 14 | 54 | 120 | 0 |
| TENSOLVE | 23 | 180 | 5351 | 2.00 |
| LANCELOT | 31 | 44 | 44 | 1.03 |
| MINOS | 31 | 49 | 115 | 0.16 |

Medium Difficult

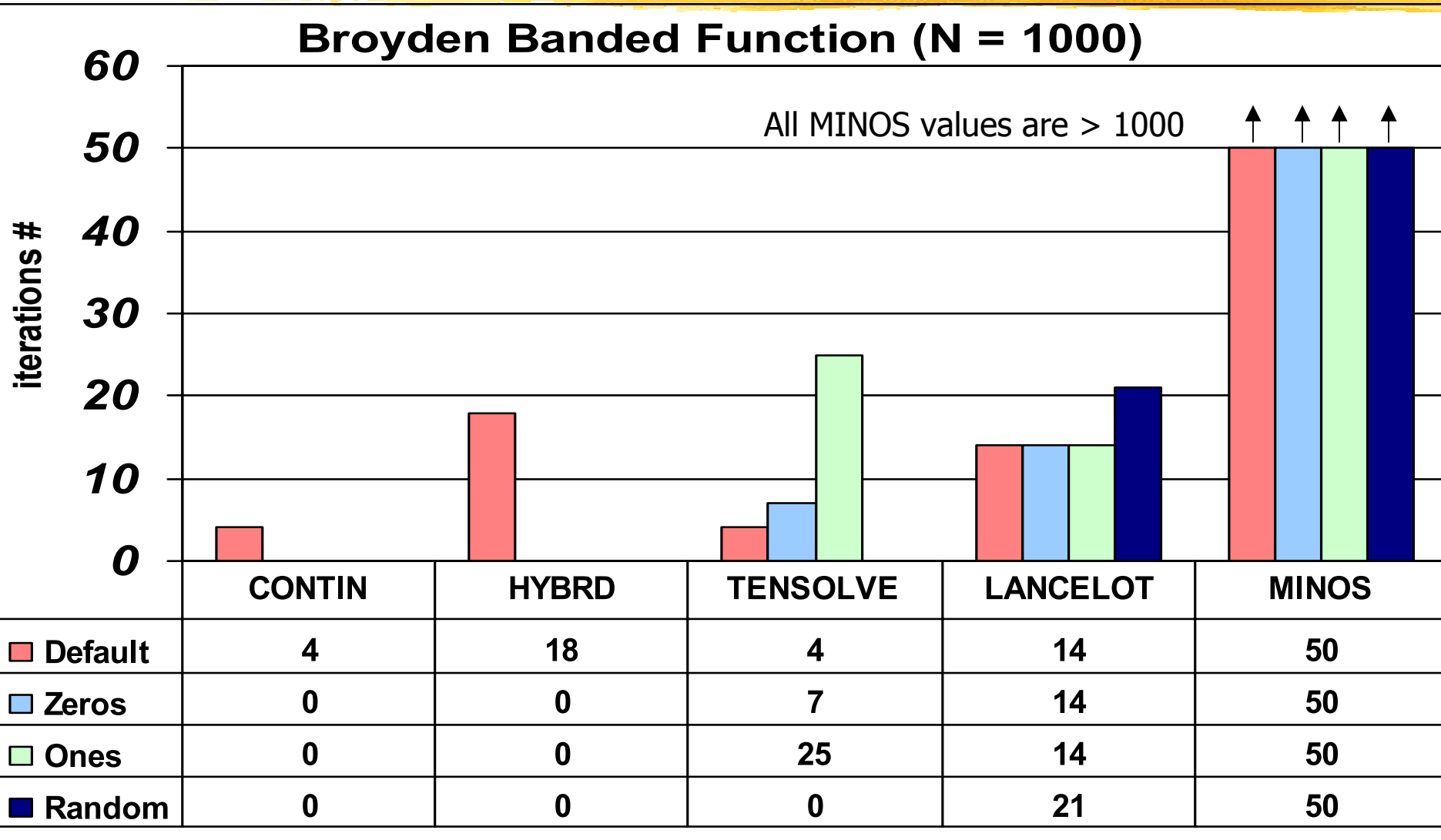
| | Brown Almost Linear | Discrete Boundary Value | Discrete Integral Equation | Trigono- metric | Variably Dimen- sioned |
|---------------|------------------------------------|--|---|----------------------------|---------------------------------------|
| Solver | Max N/ Time/Sec | Max N/ Time/Sec | Max N/ Time/Sec | Max N/ Time/Sec | Max N Time/Sec |
| CONTIN | 1000/ 300 | 11/ 0 | 4/ 0 | 23/ 0 | 4/ 0 |
| HYBRD | 22/ 0 | 1000/ 350 | 1000/ 497 | 40/ 0 | 42/ 0 |
| TENSOLVE | 1000/ 228 | 1000/ 1 | 1000/ 547 | 1000/ 2 | 200/ 31 |
| LANCELOT | 1000/ 220 | 1000/ 61 | 600/ 2826 | 1000/ 10506 | 1000/ 15 |
| MINOS | 1000/ 1094 | 1000/ 4 | 600/ 1804 | 1000/ 3737 | 1000/ 234 |

Sensitivity to starting vector (1)

Broyden Tridigonal Function (N=1000)

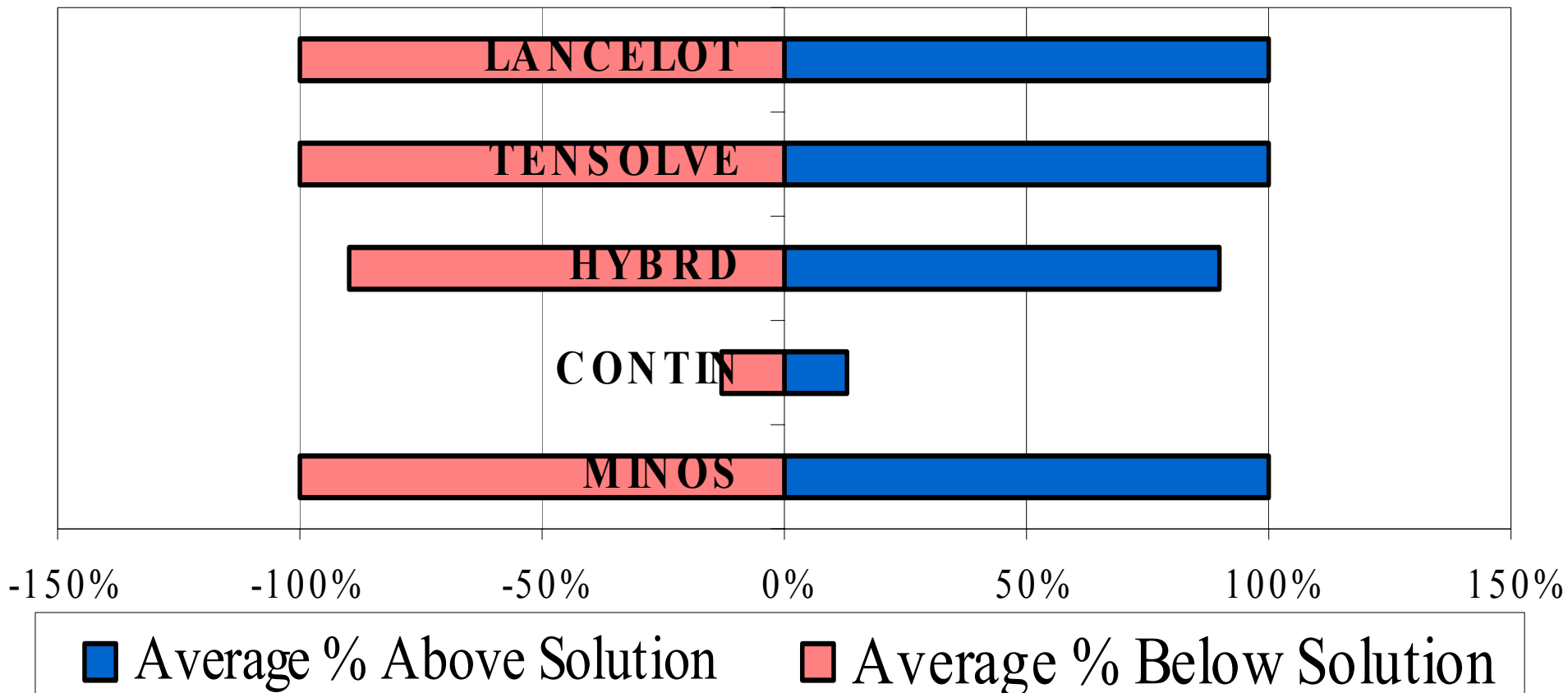


Sensitivity to starting vector (2)



Summary

Average Convergence Distance from Solution of All Test Problems



Observations (1)



- Solvability of the problem depends on interaction between
 - problem
 - solver
 - starting vector
- Problems not solvable using one solver with one starting vector may be solvable by another solver, or when a different starting vector is used

Observations (2)



- How to detect **true lack of solution** to the problem? (A. Grievank, Rouse 2000)
- Of the solvers tested, TENSOLVE, LANCELOT, and MINOS are most robust followed by HYBRD
 - CONTIN and HYBRD converge best with "default"
 - TENSOLVE converges best with "one"
 - LANCELOT and MINOS converges best with "zero"

Observations (3)



- Six popular test problems:

- Rosenbrock's
- Helical valley
- The Freudenstein-Roth function
- Powell badly scaled
- Broyden banded function
- Broyden tridiagonal

are “easily solvable” by more robust solvers → provide no useful information for performance measuring

- Watson and Chebyquad seem to be very hard to solve → can be recommended as real benchmarks for the robustness of new solvers
- New test problems needed(?)

Original Problem Revisited



- Four most robust solvers found earlier to be used:
 - HYBRD
 - TENSOLVE
 - LANCELOT
 - MINOS
- Other solvers tried and similar behavior/weakness as for the test problems observed

Initial Numerical Test

- Example used in original papers has solutions expressed by integers; for example
- For $N=2$:
 - $K(1) = 34$ and $K(2) = 5$
 - Basic solution: $h(1) = 5$ and $h(2) = 3$
- For $N=4$:
 - $K(1)=30, K(2) = 20, K(3) = 11, \text{ and } K(4)=4$
 - Basic solution: $h(1)=1, h(2)=2, h(3)=3, h(4)=4$
 - alternate solution(!)
 - $h(1)\approx 1.65, h(2)\approx 1.58, h(3)\approx 4.35, h(4)\approx 2.24$

Test Cases

- System of equations:
 - $K(\tau) = \sum_{i=0}^N h(i)h(j)$ for $j=0,1,\dots,N$
- Problem 1 \rightarrow Integer Data
 - coefficient vector created so that:
 - exact solution: $h_f(i) = i$
 - starting vector: $h_0(i) = 1$
- Problem 2 \rightarrow Floating Point Data
 - coefficient vector: real world data
 - starting vector: $h_0(i) = 1$
- Other starting vectors used; "one" seems best-overall

Results for Problem 1 (1)

| N | HYBRD | | | LANCELOT | | |
|------|-------|-----|--------------|----------|-----|--------------|
| | IC | FC | Time/ sec | IC | FC | Time/ sec |
| 2 | 8 | 10 | 0 | 10 | 11 | .01 |
| 4 | 19 | 27 | 0 | 8 | 9 | .02 |
| 8 | 8 | 16 | 0 | 11 | 12 | .04 |
| 32 | 10 | 42 | 0 | 14 | 15 | .40 |
| 64 | 10 | 74 | 0 | 23 | 24 | 2.89 |
| 128 | 10 | 110 | 1 | 49 | 50 | 34.68 |
| 256 | 10 | 160 | 2 | 161 | 162 | 905.4 |
| 512 | 10 | 210 | 3 | nc | | |
| 1024 | 10 | 310 | 12 | nc | | |

Results for Problem 1 (2)

| N | MINOS | | | TENSOLVE | | |
|------|-------|-------|--------------|----------|------|--------------|
| | IC | FC | Time/ sec | IC | FC | Time/ sec |
| 2 | 11 | 27 | .01 | 9 | 23 | .01 |
| 4 | 18 | 42 | .01 | 18 | 42 | .01 |
| 8 | 40 | 97 | .01 | 40 | 94 | .01 |
| 32 | 274 | 624 | .32 | 88 | 201 | .04 |
| 64 | 602 | 1357 | 2.47 | 274 | 624 | 2.97 |
| 128 | 1977 | 4182 | 19.2 | 602 | 1367 | 2.44 |
| 256 | 7340 | 14771 | 564.07 | 1216 | 2646 | 11.75 |
| 512 | 24205 | 47580 | 8346.04 | 4591 | 9422 | 171.02 |
| 1024 | nc | | | nc | | |

Comments



- **HYBRD**

- Converges for the largest number of equations ($N = 1024$)
- Extremely fast
- Best accuracy

- **LANCELOT**

- Converges only for up to $N = 256$
- Slowest
- Different solution(!)

- **MINOS**

- Converges for up to $N = 512$
- Different solution(!)

- **TENSOLVE**

- Converges for up to $N = 512$
- Second fastest
- Solution less accurate than HYBRD

Results for Problem 2 (1)

| | HYBRD | | | LANCELOT | | |
|-----|-------|------|--------------|----------|-----|--------------|
| N | IC | FC | Time/ sec | IC | FC | Time/ sec |
| 128 | 10 | 110 | 1 | 161 | 162 | 111.5 |
| 256 | 10 | 160 | 2 | 164 | 165 | 633.09 |
| 512 | 21 | 2095 | 90 | 201 | 206 | 5140.05 |

Results for Problem 2 (2)

| | MINOS | | | TENSOLVE | | |
|------------|--------------|-------------|----------------------|-----------------|-------------|----------------------|
| N | IC | FC | Time/ sec | IC | FC | Time/ sec |
| 128 | 1068 | 2737 | 19.22 | 28 | 3818 | 4 |
| 256 | 1383 | 3881 | 131.52 | 16 | 4381 | 20 |
| 512 | 2310 | 6508 | 1068.88 | 12 | 6692 | 117 |

Error Estimate

- Error estimate → comparison between the actual coefficient vector $K(i)$ and the computed $\underline{K}(i)$ obtained by substituting computed $\underline{h}(i)$ to the problem

| Solver | Minimum Error | Maximum Error | Average Error |
|----------|---------------|---------------|---------------|
| HYBRD | 4.40E-07 | 7.15E-05 | 1.87E-05 |
| LANCELOT | 1.25E-07 | 5.42E-05 | 1.83E-05 |
| MINOS | 4.16E-07 | 4.45E-05 | 1.80E-05 |
| TENSOLVE | 4.51E-07 | 2.83E-03 | 2.86E-04 |

Partial View of Solution Vectors

(N = 512)

| | LANCELOT | MINOS | TENSOLVE | HYBRD |
|----------|-----------------|--------------|-----------------|--------------|
| i | x(i) | x(i) | x(i) | x(i) |
| 1 | -0.0596631 | 0.141907 | -0.135745775 | -0.137628574 |
| 2 | -0.0336641 | 0.072495 | -0.080370211 | -0.079030645 |
| 3 | -0.0176408 | 0.077313 | -0.074520022 | -0.075754926 |
| 4 | 0.0450654 | 0.054729 | -0.044292743 | -0.041200875 |
| ⋮ | | | | |
| 509 | 0.289799 | -0.02806 | 0.044774790 | 0.042478349 |
| 510 | 0.417939 | -0.03148 | 0.043859563 | 0.046436000 |
| 511 | 0.139752 | -0.02624 | 0.027990174 | 0.027022607 |
| 512 | 0.133402 | -0.06305 | 0.054905619 | 0.056642195 |

Observations



- **HYBRD**
 - Convergence only if TENSOLVE results used as starting vector
 - very fast then
- **LANCELOT**
 - Converges to a different solution
- **MINOS**
 - Converges to yet another different solution
 - Slowest of the three globally converging solvers
- **TENSOLVE**
 - Converges fast
 - Least accurate solution
 - Accuracy improved by HYBRD as post-processor

Future Work



- Analysis of the three results (engineering aspects)
 - which/any/all solutions are "correct"?
 - is $N = 512$ enough?
 - what do these results mean?
- Interval solver (INTLIB) for verification
- Commercial solvers → will they introduce anything new?

Big Picture



- PROBLEM vs. SOLVER vs. STARTING VECTOR
- EXISTENCE vs. NON-EXISTENCE vs. SOLUTION #
- UNCONSTRAINED vs. CONSTRAINED PROBLEMS

THANK YOU!

A horizontal brushstroke in a vibrant yellow color, with a slightly textured, painterly appearance, extending across the width of the slide.

QUESTIONS?