



Transformed Companion Matrices as a Theoretical Tool in the Numerical Analysis of Differential Equations

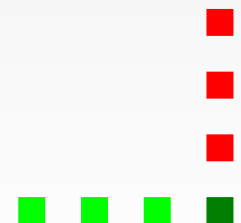
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(work in progress, with O. Koch and G. Schranz-Kirlinger)

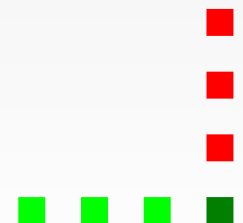
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Companion matrix

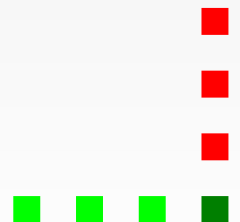
- ▶ Companion (or Frobenius) matrix (nonderogatory):

$$C = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & \ddots & \ddots & \\ & & & 0 & 1 \\ -\gamma_0 & -\gamma_1 & \dots & -\gamma_{n-2} & -\gamma_{n-1} \end{pmatrix} \in \mathbb{C}^{n \times n}$$

- ▶ ... associated with characteristic polynomial

$$\text{charpoly}(C) = p(\zeta) = \sum_{i=0}^n \gamma_i \zeta^i \quad (\gamma_n := 1)$$

in monomial (or 'Taylor') representation



Algorithmic relevance

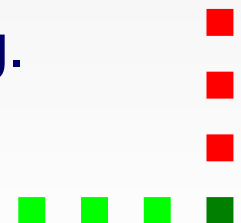
- ▶ Solving the eigenvalue problem for C (lower Hessenberg) gives the zeros of p
- ▶ Monomial representation of p may be numerically unfavorable
- ▶ Use other representations (Lagrange, Newton, ...)
→ transformed companion matrix
- ▶ Literature: “Numerical Polynomial Algebra”,
by Hans J. Stetter (SIAM Press, 2004)
... multivariate polynomial systems
- ▶ We do not consider such algorithmic aspects here
- ▶ **Rather:** Companion matrices as a theoretical tool



Companion matrix as a theoretical tool

- ▶ Example 1: n - step Linear Multistep Method
 - characterized by a polynomial p of degree n
(p not identical with stability function)
 - Companion matrix C represents equivalent one-step method in a higher dimensional space
- ▶ Example 2: Linear ODE of order n
Companion matrix C represents equivalent 1st order system
- ▶ In such cases, the characteristic polynomial p is a ‘symbol’ for the method or the problem
- ▶ Stability estimates reduce to norm estimates for $\varphi(C)$, e.g.
$$\varphi(C) = C^\nu \quad \text{or} \quad \varphi(C) = \exp(tC)$$

(= constant coefficient case)



Linear Multistep Methods

- ▶ ODE: $y'(t) = f(t, y(t))$
- ▶ n - step linear multistep method (stepsize h)

$$\sum_{k=0}^n \alpha_k y_{\nu+k} = h \sum_{k=0}^n \beta_k f(t_{\nu+k}, y_{\nu+k})$$

$$(y_{\nu+k} \approx y(t_{\nu+k}))$$

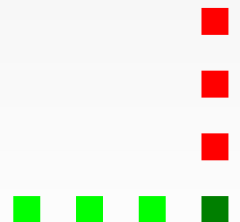
- ▶ Backward Differentiation Formulas (BDF):

$$\beta_k = 0, \quad k < n, \quad \text{and} \quad \beta_n = 1$$

- ▶ simplest cases ($n = 1, n = 2$, A-stable):

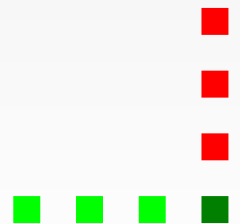
- $-y_{\nu} + y_{\nu+1} = hf(t_{\nu+1}, y_{\nu+1})$ (Backward Euler)

- $\frac{1}{2} y_{\nu} - 2 y_{\nu+1} + \frac{3}{2} y_{\nu+2} = hf(t_{\nu+2}, y_{\nu+2})$ (BDF 2)



Example: Stability of BDF 2 (i)

- ▶ To begin with: model problem $y' = \lambda y$, $\text{Re } \lambda < 0$, stiff
- ▶ Note: BDF 2 is A-stable \Rightarrow
 - $|y_\nu| \rightarrow 0$ for $\nu \rightarrow \infty$
 - **But:** Estimates for finite ν not directly available
 - ? $|y_\nu| \leq ?$
- ▶ ... This is **not** an open problem ...
 - Write BDF 2 as a one-step method in \mathbb{C}^2 using companion matrix
 - Apply the **Kreiss Matrix Theorem**, or
 - Estimate using **G-stability**
- ▶ **But:** Both approaches have a very restricted scope



Example: Stability of BDF 2 (ii)

- ▶ Let $\mu := h\lambda$, $Y_\nu := (y_\nu, y_{\nu+1})^T$
- ▶ BDF 2 $\iff Y_{\nu+1} = C(\mu) Y_\nu$, with companion matrix

$$C(\mu) = \begin{pmatrix} 0 & 1 \\ -\frac{1}{3-2\mu} & \frac{4}{3-2\mu} \end{pmatrix}, \quad |\text{EV}| \leq 1 \quad (\text{Re } \mu \leq 0)$$

- ▶ **Wanted:** Estimate $\|C(\mu)\| \leq 1$, uniformly for $\text{Re } \mu \leq 0$
 - For $\|\cdot\| = \|\cdot\|_2$ this does not hold (consider $\mu = 0$)
 - **G-stability:** Estimate O.K. for $\|Y\| = \langle G Y, Y \rangle_2$, $G = \dots$
 - **But:** \nexists G-stable n -step LMM for $n > 2$
 - \Rightarrow not generalizable
 - **Kreiss Matrix Theorem:** Restricted to const. coeff.

Example: Stability of BDF 2 (iii)

- ▶ ... look for general, more flexible approach to stability
- ▶ Jordan canonical form for $C = C(\mu)$:
 - C is diagonalizable for $\mu \neq -1/2$, $C = X \Xi X^{-1}$
 - X = Vandermonde matrix $\begin{pmatrix} 1 & 1 \\ \xi_1 & \xi_2 \end{pmatrix}$, ξ_k = EV of C
 - X becomes **singular** (confluent) for $\mu \rightarrow -1/2$
 - Estimate $\|C\| \leq \|X\| \|\Xi\| \|X^{-1}\|$ useless near $\mu = -1/2$
- ▶ Note ($\text{charpoly}(C) = p(\zeta) = (\zeta - \xi_1)(\zeta - \xi_2)$):
 - Jordan form of C is **discontinuous** w.r.t. parameter μ
 - corresponds to Lagrange representation of p for $\xi_1 \neq \xi_2$
 - **undefined** for $\xi_1 = \xi_2$ ($\mu = -1/2$)



Example: Stability of BDF 2 (iv)

► Our approach:

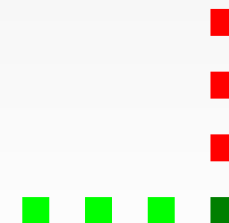
- Consider LU-decomposition of X , $X = LU$, with

$$L = \begin{pmatrix} 1 & 0 \\ \xi_1 & 1 \end{pmatrix}$$

- Then, for all μ : $C = L J L^{-1}$, with

$$J = \begin{pmatrix} \xi_1 & 1 \\ 0 & \xi_2 \end{pmatrix}$$

- ... “*Bidiagonal canonical form*” of C
- continuous w.r.t. parameter μ
- $\text{cond}(L)$ uniformly bounded for $\text{Re } \mu \leq 0$



Example: Stability of BDF 2 (v)

► We can prove:

- After appropriate diagonal scaling, $J \rightarrow \tilde{J} = D J D^{-1}$,

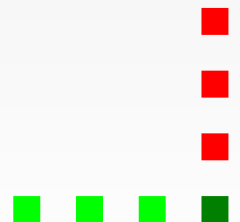
$$\|\tilde{J}\|_{\infty} \leq 1 \quad \forall \operatorname{Re} \mu \leq 0$$

- \Rightarrow uniform stability estimate

$$\|C^{\nu}\|_{\infty} \leq K, \quad \forall \nu > 0, \quad \forall \operatorname{Re} \mu \leq 0$$

- but: ... does **not** work for $\|\cdot\| = \|\cdot\|_2$
- Generalization: Transform C to “*Bidiagonal-Frobenius form*”, $C = L H L^{-1}$, with appropriate η_1, η_2 and

$$H = \begin{pmatrix} \eta_1 & 1 \\ -p[\eta_1] & -p[\eta_1, \eta_2] + \eta_2 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 \\ \eta_1 & 1 \end{pmatrix}$$



Example: Stability of BDF 2 (vi)

► We can prove:

- With $\eta_1 = \eta_2 = \frac{1}{2} \text{trace}(C)$, and after appropriate diagonal scaling, $H \rightarrow \tilde{H} = D H D^{-1}$,

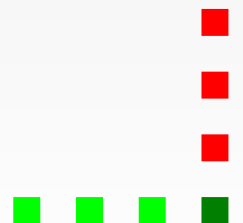
$$\|\tilde{H}\|_2 \leq 1 \quad \forall \text{Re } \mu \leq 0$$

(and: scaled version \tilde{L} of L uniformly well-conditioned)

- Proof: Apply the Cohn-Schur-criterion to $\tilde{H}^T \tilde{H}$

► Interpretation in terms of $p = \text{charpoly}(C)$:

- \tilde{H} is associated with a (scaled)
Newton-Taylor representation of p
- = Newton representation allowed to be confluent
(always well-defined)

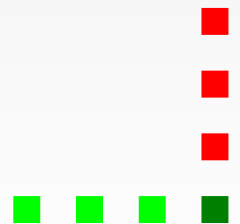


Bidiagonal-Frobenius form (i)

- ▶ For each $n \times n$ companion matrix C : $C = L H L^{-1}$,

$$H = \begin{pmatrix} \eta_1 & 1 & & & \\ & \eta_2 & 1 & & \\ & & \ddots & \ddots & \\ & & & \eta_{n-1} & 1 \\ -p_{[1]} & -p_{[1..2]} & \cdots & -p_{[1..n-1]} & -p_{[1..n]} + \eta_n \end{pmatrix}$$

- ▶ η_1, \dots, η_n arbitrary (not necessarily distinct)
- ▶ $p_{[j..k]}$... [confluent] divided differences of p w.r.t. the η_j
- ▶ L from LU-decomposition of Vandermonde matrix
 $X = X(\eta_1, \dots, \eta_n)$



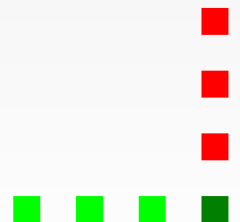
Confluent divided differences

- ▶ η_1, \dots, η_n arbitrary (not necessarily distinct)
- ▶ [Confluent] divided differences (φ smooth enough, $j \leq k$):

$$\varphi[\eta_j, \dots, \eta_k] :=$$

$$\left\{ \begin{array}{ll} \frac{\varphi[\eta_{j+1}, \dots, \eta_k] - \varphi[\eta_j, \dots, \eta_{k-1}]}{\eta_k - \eta_j}, & \eta_j \neq \eta_k \\ \lim_{\varepsilon \rightarrow 0} \frac{\varphi[\eta_{j+1}, \dots, \eta_k + \varepsilon] - \varphi[\eta_j, \dots, \eta_{k-1}]}{\varepsilon}, & \eta_j = \eta_k \end{array} \right.$$

- ▶ Includes all possible combinations of confluent ('derivative-like') and nonconfluent cases



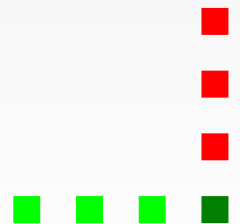
Bidiagonal-Frobenius form (ii)

- ▶ BF-form always well-defined, continuous w.r.t. varying C
- ▶ L (lower triangular):

$$L = \begin{pmatrix} | & | & & | \\ x[\eta_1] & x[\eta_1, \eta_2] & \dots & \dots & x[\eta_1, \dots, \eta_n] \\ | & | & & | \end{pmatrix}$$

where $x[\eta] = (1, \eta, \dots, \eta^{n-1})^T$, $x[\eta_j] = j$ -th column of X

- ▶ Interpretation in terms of $p = \text{charpoly}(C)$:
 - H associated with Newton-Taylor representation of p
 - = Newton representation allowed to be confluent



Bidiagonal-Frobenius form (iii)

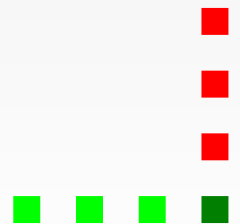
- ▶ Special case: $\eta_j \equiv \xi_j = \text{EV of } C \Rightarrow \text{Bidiagonal form:}$

$$H = J = \begin{pmatrix} \xi_1 & 1 & & & & \\ & \xi_2 & 1 & & & \\ & & \ddots & \ddots & & \\ & & & \xi_{n-1} & 1 & \\ & & & & & \xi_n \end{pmatrix}$$

- ▶ Remark: Matrix functions $\varphi(J)$ are upper tridiagonal, with

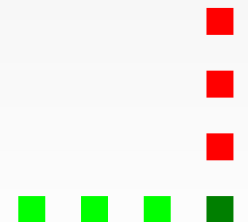
$$(\varphi(J))_{j,k} = \varphi[\xi_j, \dots, \xi_k]$$

- ▶ Trivial case: $\eta_j \equiv 0 \Rightarrow H = C$
- ▶ BF-form can be generalized to companion matrix w.r.t. arbitrary basis polynomials



Stability of Linear Multistep Methods

- ▶ Approach much more versatile than G-stability or Kreiss Matrix Theorem
- ▶ Bidiagonal canonical form has been used for stability analysis of $A(\alpha)$ -stable BDF methods applied to stiff ODEs
 - $y' = \lambda(t)y$
 - $y' = A(t)y + \phi(t, y)$
(A. Eder, G. Schranz-Kirlinger)
- ▶ To be done: Stability/convergence analysis for highly nonlinear problems



Further applications, Remarks

- ▶ Sharp growth estimates for higher order ODEs
- ▶ [Numerical] analysis of singular ODE systems:
 - Well-posedness of BVP depends on Jordan structure of boundary conditions
 - ... Use of bidiagonal form simplifies the analysis, e.g. for parameter-dependent problems
- ▶ General linear methods: [generalized] companion matrices play a role in the analysis – may be worth considering (Example: Butcher, Wright – ESIRK methods characterized by doubly companion matrices)
- ▶ General matrices: Frobenius canonical form → can be transformed blockwise

