

Explicit General Linear Methods with quadratic stability

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The problem

Ordinary Differential Equations of first order

$$\begin{cases} y'(t) = f(y), & t \in [t_0, T] \\ y(t_0) = y_0, \end{cases}$$

$$f : \mathbb{R}^m \rightarrow \mathbb{R}^m, y_0 \in \mathbb{R}^m.$$

model a huge number of phenomena, like

- the growth of populations and of infectious diseases ¹,
- phenomena in the field of medicine and genetics ²,
- economic processes ³

¹Norberto *et al* (2005), Zhao (2003), Jamrog *et al* (2006)

²Beuter *et al* (2002), Iyengar (2008), Spencer *et al* (2004), Varnai *et al* (2004), S.Yamasaki *et al* (2009)

³Gregory *et al* (2004)

Numerical solution of ODEs

Numerical simulation of real problems modeled by systems of ODEs needs for methods which are:

- accurate
- efficient
- reliable, i.e. capable of facing with 'difficult' problems, like moderately and highly stiff problems
- ...

General Linear Methods

$$t_0 < t_1 < \dots < t_N = T, \quad t_n - t_{n-1} = h$$

$$0 \leq c_1 < \dots < c_s \leq 1 \text{ collocation abscissae}$$

GLM ($\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{V}, \mathbf{B}$)

$$\begin{cases} Y_i^{[n]} &= h \sum_{j=1}^s a_{ij} f(Y_j^{[n]}) + \sum_{j=1}^r u_{ij} y_j^{[n-1]}, \quad i = 1, \dots, s, \\ y_i^{[n]} &= h \sum_{j=1}^s b_{ij} f(Y_j^{[n]}) + \sum_{j=1}^r v_{ij} y_j^{[n-1]}, \quad i = 1, \dots, r, \end{cases}$$

- $Y_i^{[n]} \approx y(t_{n-1} + c_i h)$ with order \mathbf{q}
- $y_i^{[n]} \approx$ linear combinat. of the derivatives of the solution at t_n , with order \mathbf{p}
- \mathbf{s} internal stages, \mathbf{r} external stages

Representation of GLM in vector form

Vector form in the scalar case

$$\begin{bmatrix} Y^{[n]} \\ y^{[n]} \end{bmatrix} = \left[\begin{array}{c|c} \mathbf{A} & \mathbf{U} \\ \mathbf{B} & \mathbf{V} \end{array} \right] \begin{bmatrix} hF(Y^{[n]}) \\ y^{[n-1]} \end{bmatrix}$$

$$y^{[n]} = \begin{bmatrix} y_1^{[n]} \\ \vdots \\ y_r^{[n]} \end{bmatrix}, \quad Y^{[n]} = \begin{bmatrix} Y_1^{[n]} \\ \vdots \\ Y_s^{[n]} \end{bmatrix}, \quad F(Y^{[n]}) = \begin{bmatrix} f(Y_1^{[n]}) \\ \vdots \\ f(Y_s^{[n]}) \end{bmatrix}.$$

Explicit methods: \mathbf{A} strictly lower triangular;
 Singly Implicit methods: \mathbf{A} lower triangular with $\text{diag}(\mathbf{A}) = (\lambda \dots \lambda)$

Examples of GLM: Runge Kutta methods

$$\begin{cases} Y_i^{[n]} = y_{n-1} + h \sum_{j=1}^s a_{ij} f(Y_j^{[n]}), & i = 1, 2, \dots, s, \\ y_n = y_{n-1} + h \sum_{j=1}^s b_j f(Y_j^{[n]}), \end{cases}$$

can be represented in GLM form with $r = 1$ external stage:

Runge-Kutta methods

abscissae vector: (c_1, \dots, c_s) and coefficients matrix:

$$\left[\begin{array}{c|c} A & e \\ \hline b^T & 1 \end{array} \right]$$

with $e = (1, \dots, 1)^T \in \mathbb{R}^s$.

Examples of GLM: linear multistep methods

$$y_n = \sum_{j=1}^k \alpha_j y_{n-j} + h \sum_{j=0}^k \beta_j f(y_{n-j}), \quad n = k, \dots, N$$

can be represented as a GLM with $r = 2k$ and $s = 1$:

$$\left[\begin{array}{c|c} \mathbf{A} & \mathbf{U} \\ \hline \mathbf{B} & \mathbf{V} \end{array} \right] = \left[\begin{array}{c|cccccccc} \beta_0 & \alpha_1 & \dots & \alpha_{k-1} & \alpha_k & \beta_1 & \dots & \beta_{k-1} & \beta_k \\ \beta_0 & \alpha_1 & \dots & \alpha_{k-1} & \alpha_k & \beta_1 & \dots & \beta_{k-1} & \beta_k \\ 0 & 1 & \dots & 0 & 0 & 0 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & \dots & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & \dots & 0 & 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 0 & 0 & 0 & \dots & 1 & 0 \end{array} \right]$$

Order and stage order

We assume that there exist $q_k = (q_{1k}, \dots, q_{rk})^T$, $k = 1, \dots, p$:

$$y_i^{[n-1]} = \sum_{k=0}^p q_{ik} h^k y^{(k)}(t_{n-1}) + O(h^{p+1}), \quad i = 1, \dots, r$$

Definitions

The GLM has order \mathbf{p} and stage order \mathbf{q} if

$$y_i^{[n]} = \sum_{k=0}^p q_{ik} h^k y^{(k)}(t_n) + O(h^{p+1}), \quad i = 1, \dots, r.$$

$$Y_i^{[n]} = y(t_{n-1} + c_i h) + O(h^{q+1}), \quad i = 1, \dots, s$$

Order conditions

Zero-stability: $\|V^n\| \leq C, \quad \forall n = 1, 2, \dots$

Theorem (Butcher 1993)

The GLM $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ has order p and stage order $q = p$ iff:

$$\begin{aligned} e^{c^z} &= z\mathbf{A}e^{cz} + \mathbf{U}w(z) + O(z^{p+1}) \\ e^z w(z) &= z\mathbf{B}e^{cz} + \mathbf{V}w(z) + O(z^{p+1}), \end{aligned}$$

where $e^{c^z} = [e^{c_1 z}, \dots, e^{c_s z}]$ and $w(z) = \sum_{k=0}^p q_k z^k$.

Similar order conditions hold for $q = p - 1$.

order conditions \implies conditions on the coefficients matrices

Order conditions (*cont.*)

As a matter of fact, from the previous theorem follows that the GLM $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ has order p and stage order $q = p$ iff

$$\mathbf{U}\mathbf{q}_0 = \mathbf{e}, \quad \mathbf{V}\mathbf{q}_0 = \mathbf{q}_0, \quad (\text{preconsistency conditions}),$$

where $\mathbf{e} = (1, \dots, 1)^T$ and

$$\frac{\mathbf{c}^k}{k!} - \mathbf{A} \frac{\mathbf{c}^{k-1}}{(k-1)!} - \mathbf{U}\mathbf{q}_k = 0, \quad k = 1, \dots, p$$
$$\sum_{l=0}^k \frac{\mathbf{q}_{k-l}}{l!} - \mathbf{B} \frac{\mathbf{c}^{k-1}}{(k-1)!} - \mathbf{V}\mathbf{q}_k = 0 \quad k = 1, \dots, p$$

Linear stability

Linear test equation:

$$y'(t) = \xi y(t), \quad t \geq 0,$$

$\xi \in \mathbb{C}$, $\Re \xi < 0$. Applying the GLM method $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ we get:

$$y^{[n]} = \mathbf{M}(z)y^{[n-1]}, \quad n \geq 1, \quad z = h\xi.$$

Stability matrix $M(z)$ and stability function $p(w, z)$

$$\mathbf{M}(z) = \mathbf{V} + z\mathbf{B}(\mathbf{I} - z\mathbf{A})^{-1}\mathbf{U}$$

$$p(w, z) = \det(w\mathbf{I} - \mathbf{M}(z)).$$

The method is said *absolutely stable* for $z \in \mathbb{C}$ if $p(w, z)$ has roots of modulus < 1 . The *region of absolute stability* is the set of all the points $z \in \mathbb{C}$ such that the method is stable.

GLM:

Advantages

- furnish a general formulation of classical methods
- have more free parameters of classical methods, which allow to construct new classes of methods with desirable properties of accuracy, stability, efficiency

Present limitations

- only some classes of GLM have been completely analyzed (as DIMSIM, IRKS)
- not yet known *general* conditions to impose to a GLM in order to obtain optimal stability properties together with high accuracy

RESEARCH AIM

to construct optimal GLM methods, in particular

Aim

Explicit GLM with large stability regions



Methods with special stability functions

Methods with *special* stability functions

- GLM in Nordsieck form with
Inherent Runge-Kutta Stability
(Butcher('00,'06), B.& Wright('03), Wright ('02), B.& Jackiewicz('04))
- GLM of type two-step Runge-Kutta with
Inherent Quadratic Stability
(Conte, D'Ambrosio, Jackiewicz (2010))

Runge Kutta Stability (RKS)

Definition

A GLM in Nordsieck form with order $p = q$ and $r = s = p + 1$ possesses RK stability if the stability function $\rho(w, z)$ has the special form

$$\rho(w, z) = w^p(w - R(z))$$

Th1: $R(z) = \exp(z) + O(z^{p+1})$

Th2: A Runge-Kutta method of order p has stability matrix which reduces to the stability function $R_{RK}(z) = \exp(z) + O(z^{p+1})$.

stability behavior of GLM with RKS
similar to Runge-Kutta methods

Inherent Runge-Kutta stability (IRKS)

Definition

A GLM possesses the IRKS property if $\forall e_1 = e_1$ (preconsistency),
 $\rho(w, 0) = \det(wI - V) = w^p(w - 1)$ and there exist \mathbf{X} s.t.:

$$\mathbf{BA} \equiv \mathbf{XB} \quad \text{and} \quad \mathbf{BU} \equiv \mathbf{XV} - \mathbf{VX}$$

where $\mathbf{F} \equiv \mathbf{G}$ if $\mathbf{F} = \mathbf{G}$ except for the first row.

Inherent Runge-Kutta stability (2)

Theorem 1

IRKS \implies Runge-Kutta stability

In particular $p(w, z) = w^p(w - R(z))$, with

$$R(z) = 1 + z + \frac{z^2}{2} + \cdots + \frac{z^p}{p!} + \eta \frac{z^{p+1}}{(p+1)!}.$$

Also, the error constant of the method is $E = \frac{1 - \eta}{(p+1)!}$, so the **free parameter η** can be used to balance stability and accuracy.

Inherent Runge-Kutta stability (2)

Theorem 2

The matrix \mathbf{X} in the definition of IRKS is a doubly companion matrix and has this form:

$$\mathbf{X} = \begin{bmatrix} -\alpha_1 & \cdots & -\alpha_p & -\alpha_{p+1} - \beta_{p+1} \\ 1 & \cdots & 0 & -\beta_p \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & 1 & -\beta_1 \end{bmatrix}$$

where β_1, \dots, β_p are the error constants of the external stages:

$$y_i^{[n]} = h^{i-1} y^{(i-1)}(t_n) - \beta_{p+2-i} h^{p+1} y^{(p+1)}(t_n) + O(h^{p+2}),$$

$$i = 2, \dots, p + 1.$$

Construction of IRKS methods

- Butcher and Wright ('03) developed a theory on doubly companion matrices, which led to an *algorithm* to construct IRKS GLM based only on linear operation, with as input the collocation abscissa c_1, \dots, c_s and the free parameters $\eta, \beta_1, \dots, \beta_p$;
- B& Jackiewicz ('04) studied how to choose the free parameters in order to balance stability and accuracy;
- Explicit IRKS GLM: stability regions larger than corresponding Runge-Kutta methods;
- Diagonally Implicit IRKS GLM: found classes of A -stable and L -stable methods.

How to improve RKS results?

**Relax the conditions on the stability function
by Quadratic Stability**

Conte, D'Ambrosio and Jackiewicz ('10) introduced QS for the class of Two-step Runge-Kutta methods, giving the following

Definition

The GLM $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ with $\sigma(A) = \{\lambda\}$, possesses Quadratic stability (QS) if the stability function $\rho(w, z)$ has the special form

$$\rho(w, z) = w^{s-2}((1 - \lambda z)^s w^2 - p_1(z)w + p_0(z))$$

Inherent quadratic stability (IQS)

Definition

A two-step Runge-Kutta method possesses the IQS property if exists \mathbf{X} s.t.:

$$\mathbf{BA} \equiv \mathbf{XB} \quad \text{and} \quad \mathbf{BU} \equiv \mathbf{XV} - \mathbf{VX}$$

where $\mathbf{F} \equiv \mathbf{G}$ if $\mathbf{F} = \mathbf{G}$ except for the first **two** rows.

Theorem

For the class of two-step Runge-Kutta methods,

$$\mathbf{IQS} \implies \mathbf{quadratic\ stability}$$

Conte, D'Ambrosio and Jackiewicz ('10) found classes of A -stable and L -stable IQS two-step Runge-Kutta methods up to order 8.

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to construct optimal GLM methods, in particular

Aim

Explicit GLM in Nordsieck form with large stability regions



Methods with Quadratic Stability

Explicit GLM in Nordsieck form with QS

$$\begin{cases} Y_i^{[n]} &= h \sum_{j=1}^{i-1} a_{ij} f(Y_j^{[n]}) + \sum_{j=1}^r u_{ij} z_j^{[n-1]}, \quad i = 1, 2, \dots, s, \\ z_i^{[n]} &= h \sum_{j=1}^s b_{ij} f(Y_j^{[n]}) + \sum_{j=1}^r v_{ij} z_j^{[n-1]}, \quad i = 1, 2, \dots, r, \end{cases}$$

$Y_i^{[n]} \approx (t_n + c_i h)$ of order q

$z^{[n]} \approx z(t, h)$ of order p (Nordsieck vector), i.e.

$$z(t, h) = \begin{bmatrix} y(t) \\ hy'(t) \\ \vdots \\ h^{r-1} y^{(r-1)}(t) \end{bmatrix}.$$

Other assumptions:

$$r = s \quad p = q = s - 1$$

$$\mathbf{A} = \begin{bmatrix} 0 & & & & & \\ a_{21} & 0 & & & & \\ a_{31} & a_{32} & \ddots & & & \\ \vdots & \vdots & \ddots & \ddots & & \\ a_{s,1} & a_{s,2} & \cdots & a_{s,s-1} & 0 & \end{bmatrix}, \quad \mathbf{V} = \begin{bmatrix} 1 & v_{12} & \cdots & v_{1r} \\ 0 & 0 & \cdots & v_{2r} \\ \vdots & & \ddots & \vdots \\ 0 & & \ddots & v_{r-1,r} \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$

$$\sigma(V) = \{1\} \implies \text{the method is zero-stable}$$

Quadratic stability

Definition

An explicit GLM method possesses quadratic stability property if its stability polynomial $p(w, z)$ has the form:

$$p(w, z) = w^{s-2} (w^2 - p_{s-1}(z)w + p_{s-2}(z)),$$

where

$$p_{s-1}(z) = 1 + p_{s-1,1}z + p_{s-1,2}z^2 + \cdots + p_{s-1,s}z^s,$$

$$p_{s-2}(z) = p_{s-2,1}z + p_{s-2,2}z^2 + \cdots + p_{s-2,s}z^s.$$

Inherent Quadratic Stability

Definition

A GLM possesses the Inherent Quadratic stability (IQS) property if there exist a matrix \mathbf{X} s.t.:

$$\mathbf{BA} \equiv \mathbf{XB} \quad \text{and} \quad \mathbf{BU} \equiv \mathbf{XV} - \mathbf{VX}$$

where $\mathbf{F} \equiv \mathbf{G}$ if $\mathbf{F} = \mathbf{G}$ except for the first *two* rows.

Theorem 1

Inherent Quadratic stability \implies **Quadratic stability**

Inherent Quadratic Stability (*cont.*)

Theorem 2

Assume that the GLM possesses IQS. Then the matrix \mathbf{X} has this form:

$$\mathbf{X} = \left[\begin{array}{ccccc|c} x_{1,1} & x_{1,2} & x_{1,3} & \dots & x_{1,s-1} & x_{1,s} \\ x_{2,1} & x_{2,2} & x_{2,3} & \dots & x_{2,s-1} & x_{2,s} \\ \hline 0 & 1 & 0 & \dots & 0 & x_{3,s} \\ 0 & 0 & 1 & \dots & 0 & x_{4,s} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 1 & x_{s,s} \end{array} \right],$$

$x_{3,s}, \dots, x_{s,s}$ free parameters

Order conditions

Theorem 1 (Butcher '93; B&Jackiewicz '93)

The GLM $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ in Nordsieck form has order p and stage order $q = p$ iff:

$$\begin{aligned} e^{cz} &= zAe^{cz} + U\phi(z) + O(z^{p+1}) \\ e^z\phi(z) &= zBe^{cz} + V\phi(z) + O(z^{p+1}), \end{aligned}$$

where $e^{cz} = [e^{c_1z}, \dots, e^{c_sz}]$ and $\phi(z) = [1 \quad z \quad z^2 \dots z^p]$.

Order conditions

Theorem 2

If the GLM $(\mathbf{c}, \mathbf{A}, \mathbf{U}, \mathbf{B}, \mathbf{V})$ in Nordsieck form has order p and stage order $q = p$, then

$$\mathbf{U} = \mathbf{C} - \mathbf{A}\mathbf{C}\mathbf{K} \quad \mathbf{V} = \mathbf{F} - \mathbf{B}\mathbf{C}\mathbf{K}$$

$$\mathbf{C} = \begin{bmatrix} \mathbf{e} & \mathbf{c} & \frac{\mathbf{c}^2}{2!} & \cdots & \frac{\mathbf{c}^p}{p!} \end{bmatrix}$$

$$\mathbf{K} = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ 0 & 0 & 0 & \cdots & 0 \end{bmatrix}, \quad \mathbf{F} = \begin{bmatrix} 1 & 1 & \frac{1}{2!} & \cdots & \frac{1}{p!} \\ 0 & 1 & 1 & \cdots & \frac{1}{(p-1)!} \\ 0 & 0 & 1 & \cdots & \frac{1}{(p-2)!} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{bmatrix}$$

Representation formulae

From the theorem 2 we have \mathbf{U} and the first $s - 1$ columns of \mathbf{B} :

Theorem

$$\tilde{\mathbf{B}} = \left(\begin{bmatrix} f_2 & \cdots & f_s \end{bmatrix} - \begin{bmatrix} v_2 & \cdots & v_s \end{bmatrix} - b_s \tilde{\mathbf{c}} \right) \tilde{\mathbf{C}}^{-1}.$$

where

$$\mathbf{B} = \left[\tilde{\mathbf{B}} \mid b_s \right], \quad \tilde{\mathbf{B}} \in \mathbb{R}^{s \times (s-1)}, \quad b_s \in \mathbb{R}^s,$$
$$\mathbf{F} = \begin{bmatrix} f_1 & f_2 & \cdots & f_s \end{bmatrix}, \quad \mathbf{V} = \begin{bmatrix} v_1 & v_2 & \cdots & v_s \end{bmatrix}$$
$$\tilde{\mathbf{C}} = \begin{bmatrix} 1 & c_1 & \cdots & \frac{c_1^{p-1}}{(p-1)!} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & c_{s-1} & \cdots & \frac{c_{s-1}^{p-1}}{(p-1)!} \end{bmatrix}, \quad \tilde{\mathbf{c}} = \begin{bmatrix} 1 & c_s & \frac{c_s^2}{2} & \cdots & \frac{c_{s-1}^{p-1}}{(p-1)!} \end{bmatrix}$$

Search for GLM with IQS

Assumed collocation abscissae $c = (c_1, \dots, c_s)$ fixed, we have

order conditions \implies a class of GLM methods depending on
 $m = (\mathbf{A}, \mathbf{V}, b_s)$

b_s =last column of \mathbf{B}

Free parameters in the search

$$m = (\mathbf{A}, \mathbf{V}, b_s) \quad \text{and} \quad x = (x_{3s}, \dots, x_{ss})^T,$$

used to ask for:

- Inherent Quadratic Stability: $\mathbf{BA} \equiv \mathbf{XB}, \mathbf{BU} \equiv \mathbf{XV} - \mathbf{VX}$
- Maximizing the area of the stability region: $area(m)$

Search for GLM with IQS (*cont.*)

IQS: $\mathbf{BA} \equiv \mathbf{XB}$, $\mathbf{BU} \equiv \mathbf{XV} - \mathbf{VX}$, so we define:

$$Q_1(m, x) = \|(\mathbf{BA} - \mathbf{XB})(3 : s, :)\|_F,$$

$$Q_2(m, x) = \|(\mathbf{BU} - \mathbf{XV} + \mathbf{VX})(3 : s, :)\|_F$$

and search (m, x) which maximize

objective function

$$f(m, x) = -\text{area}(m) + K(Q_1(m, x) + Q_2(m, x)),$$

$K \approx 10^{12}$ penalty constant

This approach works well only for $s \leq 4$

Second approach: ask for Quadratic Stability

First step: find an *optimal* quadratic pol. $q_{op}(w, z)$

$$q_{op}(w, z) = w^{s-2}(w^2 + q_{s-1}(z)w + q_{s-2}(z)),$$

$$q_{s-1}(z) = 1 + q_{s-1,1}z + \cdots + q_{s-1,s}z^s,$$

$$q_{s-2}(z) = q_{s-2,1}z + \cdots + q_{s-2,s}z^s,$$

which

- satisfies the same conditions as the stability polynomial of a GLM of order $p = q = s - 1$
- has **maximum area of stability region**

Second approach: ask for Quadratic Stability (*cont.*)

Second step: find a GLM with stability polynomial

$$p(w, z) \approx q_{op}(w, z)$$

Find a GLM defined by $m = (\mathbf{A}, \mathbf{V}, b_s)$ with stability pol.

$$p(w, z) = w^s + p_{s-1}(z)w^{s-1} + \dots + p_1(z)w + p_0(z),$$

p_j pol. of degree s , with m minimizing the objective function:

$$f(m) = \text{distance between } p(w, z) \text{ and } q_{op}(w, z).$$

where the distance is the norm of the difference of the coefficients.

Low accuracy reached \implies stability regions notably smaller
than that of $q_{op}(w, z)$

Second approach: ask for Quadratic Stability (*cont.*)

Third step: find a GLM with stability pol. $p(w, z) \approx q_{op}(w, z)$

Using output m of the second step as a starting guess,
find a GLM defined by $m = (\mathbf{A}, \mathbf{V}, b_s)$ with stability pol.

$$p(w, z) = w^s + p_{s-1}(z)w^{s-1} + \dots + p_1(z)w + p_0(z),$$

with m solution of the nonlinear system:

$$\begin{cases} p_{s-1,k} = q_{s-1,k}, & p_{s-2,k} = q_{s-2,k}, & k = 1, \dots, s, \\ p_{j,k} = 0, & j = 0, \dots, s-3, & k = 1, \dots, s. \end{cases}$$

$$q := q_{op}, \quad p_j(z) = p_{js}z^s + \dots + p_{j1}z + p_{j0}.$$

High accuracy reached \implies same stability regions of $q_{op}(w, z)$

Efficient computation of stability polynomial coefficients

The computational kernel of our search is

- minimization problem of $f(m)$
- solution of a nonlinear system $g(m) = 0$

evaluation of $f(m)$ or $g(m) \implies$ *computation of the coefficients of $p(w, z)$*

so we need for:

- a representation formula for the coefficients of $p(w, z)$
- an efficient technique for the implementation of such formula

Representation formula for the coefficients of $p(w, z)$

$$p(w, z) = w^s + p_{s-1}(z)w^{s-1} + \dots + p_1(z)w + p_0(z), \quad p_j(z) = \sum_{k=0}^s p_{jk}z^k$$

By the Fourier series approach (Butcher and Jackiewicz '96):

$$p_{jk} = \frac{1}{N_1 N_2} \sum_{\mu=0}^{N_1-1} \sum_{\nu=0}^{N_2-1} w_{\mu}^{-j} z_{\nu}^{-k} p(w_{\mu}, z_{\nu}), \quad j, k = 0, 1, \dots, s-1.$$

$$w_{\mu} = \exp\left(-\frac{2\pi\mu i}{N_1}\right), \quad z_{\nu} = \exp\left(-\frac{2\pi\nu i}{N_2}\right)$$

Comp. cost $O(s^2 N_1 N_2)$ flops and $O(N_1 N_2)$ eval. of $p(w, z)$

fast Fourier transform technique

By the representation formula we have:

$$\forall j, \quad p_{jk} = \frac{1}{N_2} \sum_{\nu=0}^{N_2-1} G_{j,\nu} \exp\left(\frac{2\pi\nu k i}{N_2}\right), \quad k = 0, \dots, s,$$

$$\forall \nu, \quad G_{j,\nu} = \frac{1}{N_1} \sum_{\mu=0}^{N_1-1} p(w_\mu, z_\nu) \exp\left(\frac{2\pi\mu j i}{N_1}\right), \quad j = 0, \dots, s,$$

$(p_{j0}, \dots, p_{js}) =$ first $(s + 1)$ elements of **IDFT** of $(G_{j,0}, \dots, G_{j,N_2-1})$

$(G_{\nu,0}, \dots, G_{\nu,s-1}) =$ first s el. **IDFT** of $(p(w_0, z_\nu), \dots, p(w_{N_1-1}, z_\nu))$

IDFT = inverse discrete Fourier transform

Algorithm for the coefficients of $p(w, z)$ by FFT

for $\nu = 0 : N_2 - 1$

$(G_{\nu,0}, \dots, G_{\nu,s-1}) =$ first s components of IDFT of
 $(p(w_0, z_\nu), \dots, p(w_{N_1-1}, z_\nu))$

endfor

for $j = 0 : s - 1$

$(p_{j0}, \dots, p_{js}) =$ first $(s + 1)$ components of IDFT of
 $(G_{j,0}, \dots, G_{j,N_2-1})$

endfor

use of fast Fourier Transform algorithm for the IDFTs

fast Fourier transform technique: cost

Computing the IDFTs by the FFT algorithm we have:

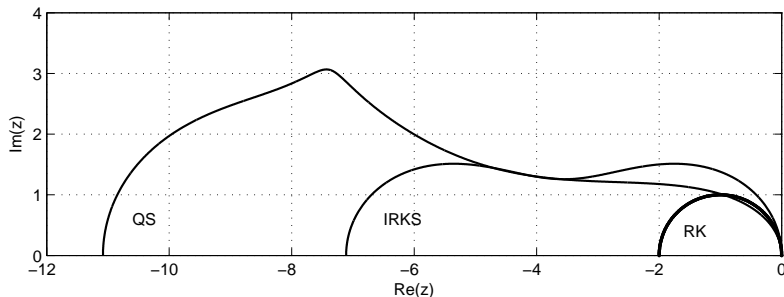
reduction of the computational cost

$$\frac{s^2 N}{(s + N) \log(N)}, \quad N = \max(N_1, N_2).$$

for example, if $s = 5$, $N = 16$, the reduction is about 5.

Examples of methods with IQS/QS

$$s = 2, p = 1$$



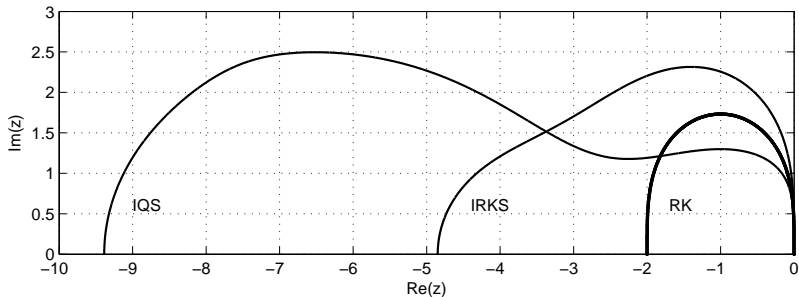
IQS: Area 35.06

Constant error: 0.6605

IRKS: Area ≈ 18

Constant error: 0.3594

$s=3, p=2$



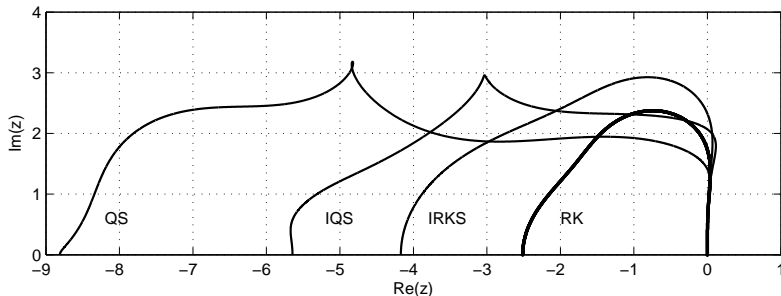
IQS: Area 32.28

Constant error: 0.2019

IRKS: Area ≈ 18

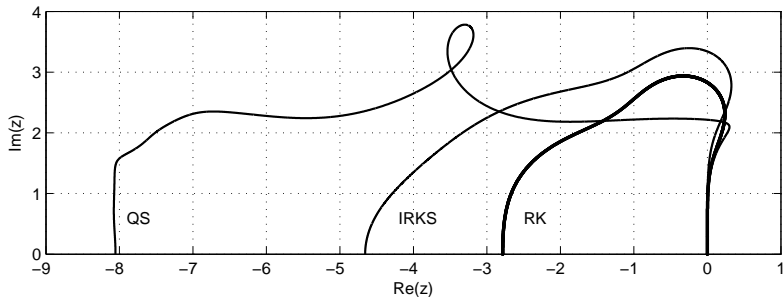
Constant error: 0.0885

$s=4, p=3$



QS:	Area: 35.09	Constant error: 0.0623
IQS:	Area: 22.85	Constant error: 0.0187
IRKS:	Area: 18.16	Constant error: 0.1667

$s=5, p=4$



QS: Area 36.53

Constant error: 0.0126

IRKS: Area 21.81

Constant error: 0.0033

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Conclusions

Explicit GLM in Nordsieck form with QS

- IQS sufficient condition for QS
- constructive procedure for IQS/QS GLM based on minimization and on the solution of nonlinear systems
- efficient computation of the stability polynomial coefficients
- stability regions considerably larger than corresponding IRKS methods for $s = 1, 2, 3, 4, 5$ and $p = s - 1$

Future developments

- higher order explicit GLM and implicit GLM with QS/IQS
- theoretical analysis to derive a constructive algorithm for IQS GLM based only on linear operations
- implementation in a variable-step and/or variable-order code