

On superconvergence features of methods based on the Crank-Nicolson scheme in the context of diffusion PDEs (deterministic and stochastic)

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Based on joint works with

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Contents

In this talk, we present two situations where the **Crank-Nicolson method** is surprisingly more **accurate** than one could expect:

- 1 In the context of splitting methods for parabolic PDEs
- 2 In the context of ergodic stochastic parabolic PDEs

Part 1

1 In the context of splitting methods for parabolic PDEs

2 In the context of ergodic stochastic parabolic PDEs

- G. Bertoli and V., Strang splitting method for semilinear parabolic problems with inhomogeneous boundary conditions: a correction based on the flow of the nonlinearity, *SIAM J. Sci. Comput.* 42 (2020), A1913-A1934.
- G. Bertoli, C. Besse, and V., Superconvergence of the Strang splitting when using the Crank-Nicolson scheme for parabolic PDEs with oblique boundary conditions, [arXiv:2011.05178](https://arxiv.org/abs/2011.05178), submitted.

The Strang splitting method

For parabolic semilinear problems of the form

$$\begin{aligned}\partial_t u(x, t) &= Du(x, t) + f(x, u(x, t)) \quad \text{in } \Omega \times (0, T], \\ Bu(x, t) &= b(x) \quad \text{on } \partial\Omega \times (0, T], \\ u(x, 0) &= u_0(x) \quad \text{in } \Omega,\end{aligned}$$

we consider time discretizations with **splitting methods**:

$$\begin{aligned}\phi_t^f : \quad & \partial_t u(x, t) = f(x, u(x, t)) \quad \text{in } \Omega \times (0, T], \\ \phi_t^D : \quad & \begin{aligned} \partial_t u(x, t) &= Du(x, t) \quad \text{in } \Omega \times (0, T], \\ Bu(x, t) &= b(x) \quad \text{on } \partial\Omega \times (0, T]. \end{aligned}\end{aligned}$$

For approximating $u(t_n, x) \simeq u_n(x)$, $t_n = n\tau$, the **Strang splitting** method writes

$$u_{n+1} = \phi_{\frac{\tau}{2}}^f \circ \phi_{\tau}^D \circ \phi_{\frac{\tau}{2}}^f(u_n),$$

Or alternatively, another version writes:

$$u_{n+1} = \phi_{\frac{\tau}{2}}^D \circ \phi_{\tau}^f \circ \phi_{\frac{\tau}{2}}^D(u_n).$$

The Crank-Nicolson method

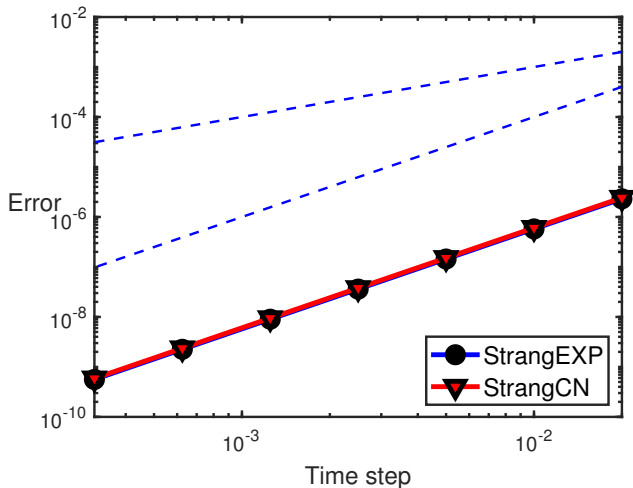
It is typical to approximate the exact flow of the diffusion part,

$$u(t) = \phi_t^D(u_0) : \quad \begin{aligned} \partial_t u &= Du && \text{in } \Omega \times (0, T], \\ Bu &= b && \text{on } \partial\Omega \times (0, T], \end{aligned}$$

either with a Krylov method, or with a Runge-Kutta type method, such as the **Crank-Nicolson method** (formally of order two),

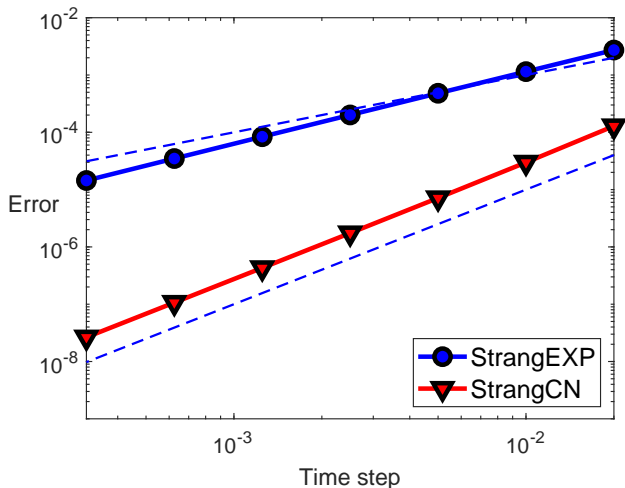
$$u_1 = \phi_\tau^{D,CN}(u_0) : \quad \begin{aligned} \frac{u_1(x) - u_0(x)}{2} &= D \frac{u_1(x) + u_0(x)}{2} && \text{in } \Omega, \\ B \frac{u_0(x) + u_1(x)}{2} &= b(x) && \text{on } \partial\Omega. \end{aligned}$$

A numerical example on $\Omega = (0, 1)$



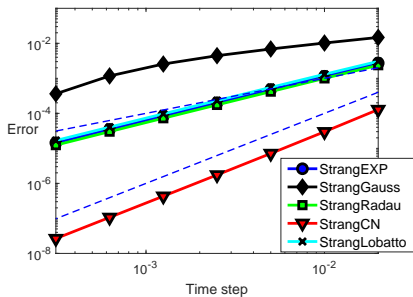
$$\partial_t u = \partial_{xx} u + e^{-x} \text{ on } \Omega, \quad \partial_x u + u = 1 \text{ on } \partial\Omega, \quad u(x, 0) = 1.$$

A numerical example: order reduction!

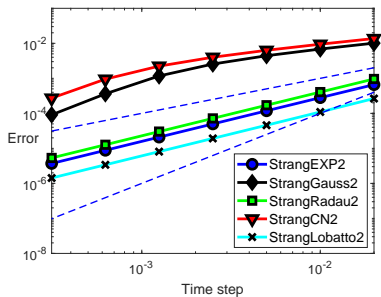


$$\partial_t u = \partial_{xx} u + 1 \text{ on } \Omega, \quad u = 1 \text{ on } \partial\Omega, \quad u(x, 0) = 1.$$

With other approximations of the diffusion part?



$$u_{n+1} = \phi_{\frac{\tau}{2}}^f \circ \phi_{\tau}^D \circ \phi_{\frac{\tau}{2}}^f(u_n)$$



$$u_{n+1} = \phi_{\frac{\tau}{2}}^D \circ \phi_{\tau}^f \circ \phi_{\frac{\tau}{2}}^D(u_n)$$

$$\partial_t u = \partial_{xx} u + 1 \text{ on } \Omega, \quad u = 1 \text{ on } \partial\Omega, \quad u(x, 0) = 1.$$

What happened?

An observation from (W. Hundsdorfer and J. Verwer. 1994): the reason of the order reduction is that $w = \phi_\tau^f(u_n)$ does not preserve the **boundary conditions**, i.e. $Bu_n = b \not\Rightarrow Bw = b$ for $Bf(u) \neq 0$.

Possible remedies to order reduction in splitting methods:

- (L. Einkemmer and A. Ostermann, 2016) Modify the splitting as:

$$u_{n+1} = \phi_{\frac{\tau}{2}}^{D+q_n} \circ \phi_\tau^{f-q_n} \circ \phi_{\frac{\tau}{2}}^{D+q_n}(u_n),$$

where q_n is chosen such that $Bq_n = Bf(u_n) + \mathcal{O}(\tau)$ on $\partial\Omega$.

- (G. Bertoli and V., 2020) Consider a **five step splitting method** as

$$u_{n+1} = \phi_{\frac{\tau}{2}}^f \circ \phi_{\frac{\tau}{2}}^{-q_n} \circ \phi_\tau^{D+q_n} \circ \phi_{\frac{\tau}{2}}^{-q_n} \circ \phi_{\frac{\tau}{2}}^f(u_n)$$

where q_n is computed only using the source flow output $\phi_{\frac{\tau}{2}}^f(u_n)$ such that $w = \phi_{\frac{\tau}{2}}^{-q_n} \circ \phi_{\frac{\tau}{2}}^f(u_n)$ satisfies the boundary conditions $Bw = b$ up to $\mathcal{O}(\tau^2)$.

This halves the number of evaluations of the diffusion part ϕ_τ^D .

Note: In this talk, we shall not use the above “repair” techniques.

Superconvergence of splitting with Crank-Nicolson

For the diffusion-reaction problem, consider the **splitting method**

$$u_{n+1} = \phi_{\frac{\tau}{2}}^f \circ \phi_{\tau}^{D, CN} \circ \phi_{\frac{\tau}{2}}^f(u_n).$$

Theorem (Bertoli, Besse, V., 2020) (order two for $f = f(x)$)

Let $u_0 \in W^{2,p}(\Omega)$ with $Bu_0 = b$ on $\partial\Omega$ and $f = f(x) \in L^p(\Omega)$. Then the global error $e_n = u_n - u(t_n)$ satisfies

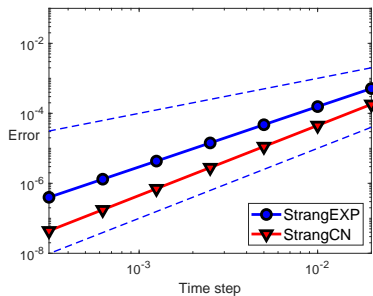
$$\|e_n\|_{L^p(\Omega)} \leq \frac{C\tau^2}{t_n}, \quad e_n = (r(\tau A)^n - e^{n\tau A}) A^{-1}(Du_0 + f),$$

where A is the restriction of the operator D to the set of functions satisfying the **homogeneous boundary condition** $Bu(x) = 0$ on $\partial\Omega$, and $r(z) = \frac{1+\frac{z}{2}}{1-\frac{z}{2}}$ is the stability function of the CN scheme.

Remark. This result **does not** persist for $\phi_{\frac{\tau}{2}}^{D, CN} \circ \phi_{\tau}^f \circ \phi_{\frac{\tau}{2}}^{D, CN}$.

Homogeneous case: $A = D, f = 0, u_0 \in D(A)$, see A. Hansbo, 1999.

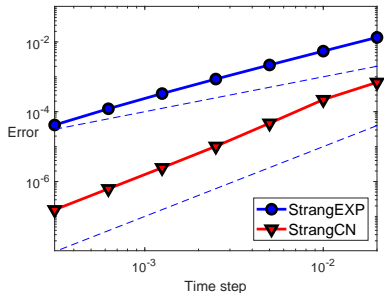
Numerical experiments on $\Omega = (0, 1)^2$: nonlinear case



$$f(u) = u$$

Robin boundary conditions

$$u + \partial_n u = u_0 \text{ on } \partial\Omega$$



$$f(u) = u^2$$

Dirichlet + Neumann
boundary conditions

$$\partial_t u = \Delta u + f(u) \text{ on } \Omega, \quad u(x, 0) = u_0(x).$$

Part 2

1 In the context of splitting methods for parabolic PDEs

2 In the context of ergodic stochastic parabolic PDEs

- C.-E. Bréhier and V., High-order integrator for sampling the invariant distribution of a class of parabolic stochastic PDEs with additive space-time noise, *SIAM J. Sci. Comput* 38 (2016) A2283-A2306.
- A. Abdulle, I. Almuslimani, and V., Optimal explicit stabilized integrator of weak order one for stiff and ergodic stochastic differential equations, *SIAM/ASA J. Uncertain. Quantif.* 6 (2018), 937-964.
- A. Abdulle, C.-E. Bréhier, and V., Convergence analysis of explicit stabilized integrators for parabolic semilinear stochastic PDEs, arXiv:2102.03209, Submitted.

Parabolic stochastic PDE case

Consider e.g. the semilinear stochastic heat equation:

$$\begin{aligned}\partial_t u(t, x) &= \partial_{xx} u(t, x) + f(u(t, x)) + \dot{W}(t, x), \quad t > 0, x \in (0, 1), \\ u(0, x) &= u_0(x), \quad x \in (0, 1), \\ u(t, x) &= 0, \quad x \in \partial(0, 1),\end{aligned}$$

or its abstract formulation on the Hilbert space H :

$$\begin{aligned}du(t) &= Au(t)dt + f(u(t))dt + dW(t), \quad t > 0 \\ u(0) &= u_0.\end{aligned}$$

Ergodicity convergence properties,

$$\begin{aligned}\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \phi(u(t)) dt &= \int_H \phi(y) d\mu_\infty(y) \quad \text{a.s.} \\ \left| \mathbb{E}(\phi(u(t))) - \int_H \phi(y) d\mu_\infty(y) \right| &\leq K(x, \phi) e^{-ct}, \quad \text{for all } t \geq 0.\end{aligned}$$

The stochastic PDE case: the θ -method

$$du(t) = Au(t)dt + f(u(t))dt + dW^Q(t), \quad u(0) = u_0 \in H.$$

- θ -method (linearized):

$$v_{n+1} = v_n + \theta\tau v_n + (1 - \theta)\tau Av_{n+1} + \tau f(v_n) + \sqrt{\tau}\xi_n^Q$$

- $\theta = 1/2$ method (“Crank-Nicolson”)
- $\theta = 1$ method (“Euler method”):

$$v_{n+1} = J_1 v_n + \tau J_1 f(v_n) + \sqrt{\tau} J_1 \xi_n^Q,$$

where $J_1 = (I - \tau A)^{-1}$ and $\sqrt{\tau}\xi_n^Q = W^Q((n+1)\tau) - W^Q(\tau h)$.
order of convergence is $\bar{s} - \varepsilon$ for all $\varepsilon > 0$ (see Bréhier 2014):

$$\bar{s} = \sup \left\{ s \in (0, 1) ; \text{Trace} \left((-A)^{-1+s} Q \right) < +\infty \right\} > 0.$$

Example: for $A = \frac{\partial^2}{\partial x^2}$, $Q = I$ in dimension 1, we have $\bar{s} = 1/2$.

Numerical experiments (stochastic heat equation)

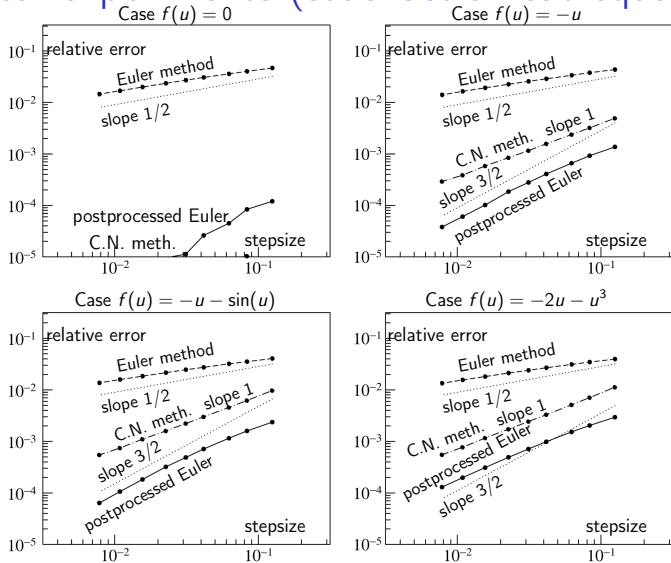


Figure: Orders of convergence, test function $\varphi(u) = \exp(-\|u\|^2)$.

Quadratic one-dimensional case (Ornstein-Uhlenbeck)

Consider the scalar OU process, with unique inv. measure $\mathcal{N}(0, \frac{\sigma^2}{2|\lambda|})$,

$$dX(t) = \lambda X(t)dt + \sigma dW(t), \quad (\lambda < 0).$$

Proposition: invariant measure preservation

For a one-step integrator where $z = \lambda\tau$, $\xi_n \sim \mathcal{N}(0, 1)$,

$$X_{n+1} = A(z)X_n + B(z)\sqrt{h}\sigma\xi_n,$$

\bar{X}_n converges to the exact invariant measure iff $|A(z)| < 1$ and

$$\frac{-2zB(z)^2}{1 - A(z)^2} = 1.$$

For rational $A(z), B(z)$, this condition for all z implies $|A(\infty)| = 1$.

Example: $\theta = 1/2$ method (Crank-Nicolson):

$$A(z) = \frac{1+z/2}{1-z/2}, \quad B(z) = \frac{1}{1-z/2} \quad (\text{see Chong, Walsh, 2012}).$$

Remark: Impossible to be exact (without postprocessing) for an explicit Runge-Kutta method or an L -stable method ($A(\infty) = 0$).

Postprocessed integrators for ergodic SDEs

Idea: extend to the context of ergodic SDEs the popular idea of effective order for ODEs from Butcher 69',

$$y_{n+1} = \chi_h \circ K_h \circ \chi_h^{-1}(y_n), \quad y_n = \chi_h \circ K_h^n \circ \chi_h^{-1}(y_0).$$

Example based on the Euler-Maruyama method

for Brownian dynamics: $dX(t) = -\nabla V(X(t))dt + \sigma dW(t)$.

$$X_{n+1} = X_n - \tau \nabla V \left(X_n + \frac{1}{2} \sigma \sqrt{\tau} \xi_n \right) + \sigma \sqrt{\tau} \xi_n, \quad \bar{X}_n = X_n + \frac{1}{2} \sigma \sqrt{\tau} \xi_n.$$

X_n has order 1 of accuracy for the invariant measure.

\bar{X}_n has order 2 of accuracy for the invariant measure (postprocessor).

This method was first derived as a [non-Markovian method](#) by Leimkhuler, Matthews, 2013, see Leimkhuler, Matthews, Tretyakov, 2014,

$$\bar{X}_{n+1} = \bar{X}_n - \tau \nabla V(\bar{X}_n) + \frac{1}{2} \sigma \sqrt{\tau} (\xi_n + \xi_{n+1}).$$

Quadratic one-dimensional case (Ornstein-Uhlenbeck)

Proposition: invariant measure preservation

For a postprocessed integrator where $z = \lambda\tau$, $\xi_n \sim \mathcal{N}(0, 1)$,

$$X_{n+1} = A(z)X_n + B(z)\sqrt{h}\sigma\xi_n, \quad \bar{X}_n = C(z)X_n + D(z)\sqrt{h}\sigma\xi_n,$$

\bar{X}_n converges to the exact invariant measure if $|A(z)| < 1$ and

$$\frac{-2zB(z)^2C(z)^2}{1 - A(z)^2} - 2zD(z)^2 = 1.$$

- $\theta = 0$ meth., $A(z) = 1 + z$, $B(z) = 1 + \frac{z}{2}$, $C(z) = 1$, $D(z) = \frac{1}{2}$,
(see the order two method from Leimkhuler, Matthews, 2013).
- $\theta = 1$ method, $A(z) = B(z) = \frac{1}{1-z}$, $C(z) = 1$, $D(z) = \frac{1}{\sqrt{1-z/2}}$,
(see Bréhier, V., 2016).

Quadratic one-dimensional case (Ornstein-Uhlenbeck)

Consider the scalar OU process, with unique inv. measure $\mathcal{N}(0, \frac{\sigma^2}{2|\lambda|})$,

$$dX(t) = \lambda X(t)dt + \sigma dW(t), \quad (\lambda < 0).$$

Proposition: invariant measure preservation by explicit stabilized methods (Abdulle, Almuslimani, V., 2018)

For all $s \geq 1$, the family of polynomials

$$A(z) = T_s(1 + \frac{z}{s^2}), \quad B(z) = s^{-1} U_{s-1}(1 + \frac{z}{s^2})(1 + \frac{z}{2s}),$$
$$C(z) = 1, \quad D(z) = \frac{1}{2s},$$

satisfy the invariant measure preservation condition

$$\frac{-2zB(z)^2 C(z)^2}{1 - A(z)^2} - 2zD(z)^2 = 1.$$

For $s = 1$, we recover the scheme of Leimkhuler, Matthews, 2013.

Related work for MCMC: Vargas, Pereyra, Zygalakis (2020).

Analysis of the postprocessed Euler method

Theorem (Bréhier, V., 2016)

- The Markov chain $(u_n, \bar{u}_{n-1})_{n \in \mathbb{N}}$ is ergodic, with unique invariant distribution, and for any test function $\varphi : H \rightarrow \mathbb{R}$ of class \mathcal{C}^2 , with bounded derivatives,

$$\left| \mathbb{E}(\varphi(\bar{u}_n)) - \int_H \varphi(y) d\bar{\mu}_\infty^h(y) \right| = \mathcal{O} \left(\exp\left(-\frac{(\lambda_1 - L)}{1 + \lambda_1 h} n\tau\right) \right).$$

Remark: No such estimate $\mathcal{O}(e^{-ct_n})$ for Crank-Nicolson.

- Moreover, for the case of a linear F , for any $s \in (0, \bar{s})$,

$$\int_H \varphi(y) d\bar{\mu}_\infty^h(y) - \int_H \varphi(y) d\mu_\infty(y) = \mathcal{O}(\tau^{s+1}).$$

Remark: error for the standard Euler: $\mathcal{O}(\tau^s)$, $s \in (0, \bar{s})$.

Strong convergence of explicit stabilized methods

$$u_{n+1}^h = A_s(\tau\Lambda_h)u_n^h + B_s(\tau\Lambda_h)P_h(\tau F(u_n^h) + \Delta W_n^Q).$$

Remark: the stage s is chosen s.t. $\tau\lambda_{\max,h} \leq L_s$, i.e. $\tau Ch^{-2} \simeq s^2$.

Theorem (Abdulle, Bréhier, V., 2021)

Assume $A_s(0) = A'_s(0) = B_s(0) = 1, s \geq 1$,

$$\sup_{s \geq 1, z \in [-L_s, 0]} (1 + |z|)|B_s(z)|^2 < \infty,$$

$$\sup_{s \geq 1, z \in [-L_s, -\delta]} |A_s(z)| < 1, \quad \text{for all } \delta \in (0, 1],$$

$$\sup_{s \geq 1, |z| \leq \delta} (|A_s(z)| + |B_s(z)|) < \infty \quad \text{for some } \delta \in (0, 1].$$

Then, for all $T > 0$, and all $\alpha \in (0, \bar{\alpha})$,

$$|\mathbb{E}\|u^h(n\tau) - u_n^h\|^2|^{1/2} \leq C_{T,\alpha}(1 + |u_0|t_n^{-\alpha/2})\tau^{\alpha/2},$$

where $C_{T,\alpha}$ is independent of h, τ, α .

Stability functions for the SK-ROCK method

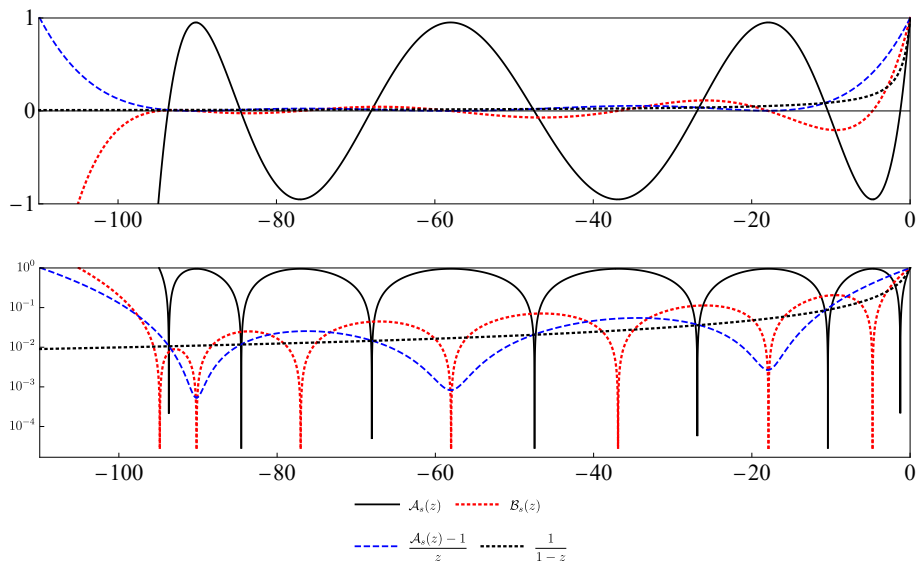


Figure: Stability functions, degree $s = 7$, damping parameter $\eta = 0.05$.

Concluding remarks

- Surprisingly, the Crank-Nicolson scheme (not L-stable) performs better than the exact solution in the Strang splitting method, avoiding order reduction phenomena.
- This seems specific to the Crank-Nicolson scheme, and it does not persist for dispersive problems, e.g.

$$i\partial_t u = \partial_{xx} u + 1 \text{ on } \Omega, \quad u = 1 \text{ on } \partial\Omega, \quad u(x, 0) = 1.$$

- In the context of ergodic stochastic dynamics, L-stability is a desirable property and high order for the invariant measure can be achieved in spite of the low regularity of the solution.

Current related works:

- deterministic: Analysis for nonlinear $f(u)$? Case of absorbing boundary conditions?
- stochastic: study of algebraic structures with exotic aromatic Butcher trees, accelerating the convergence to equilibrium using reversible perturbations (Stratonovitch noise).