

EFFECT OF STATISTICAL NOISE ON COUPLED PLASMA FLUID – MONTE CARLO KINETIC NEUTRALS SIMULATIONS

**Y. Marandet¹, H. Bufferand¹, M. Valentinuzzi², G. Ciruolo²,
P. Meliga³, J. Rosato¹, E. Serre³, P. Tamain²**

¹ PIIM, CNRS, Aix-Marseille Université, Marseille

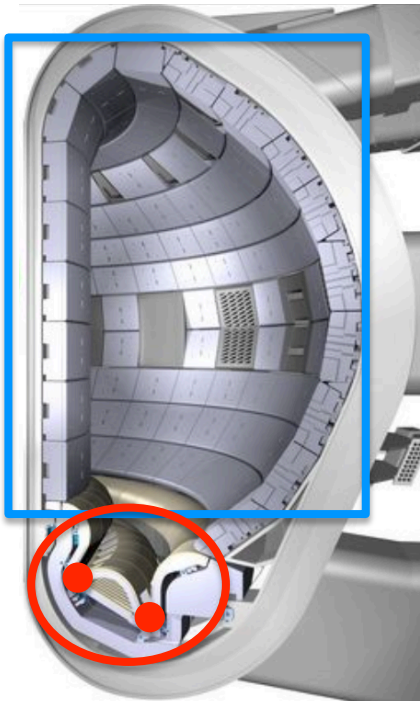
² IRFM-CEA, Cadarache

³ M2P2, CNRS, Aix-Marseille Université, Marseille

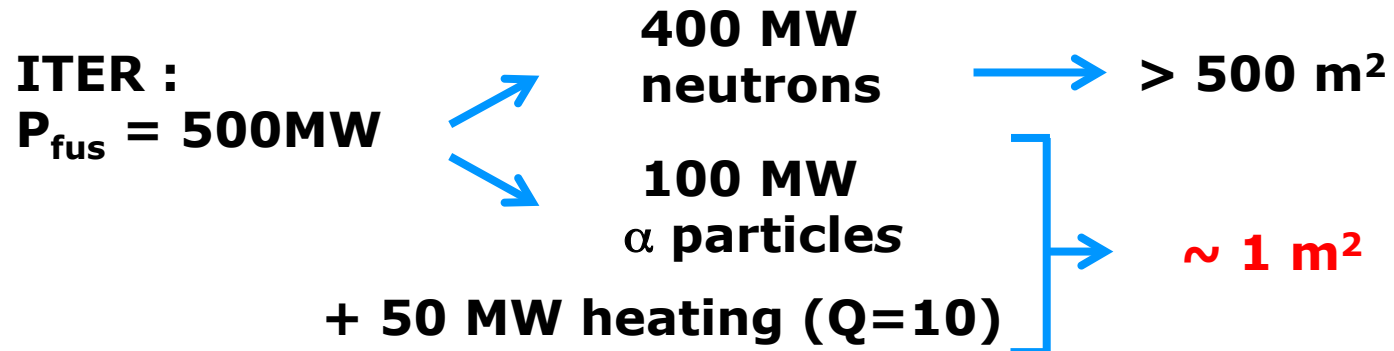


Introduction – Power exhaust in tokamaks

- ✓ Plasma wall interactions play an essential role in magnetized fusion devices
- ✓ In steady state, the fusion energy has to be extracted



Divertor



Spreading the energy on a larger surface area is *mandatory* (B geometry, radiation, ...)

- ✓ Avoid sputtering : lower temperature to a few eV

$$\text{Heat flux} = 13.6 \text{ eV} \times \Gamma (\text{particle flux})$$

- ✓ On top of that: ELMs (See Marina's talk)

Introduction

- ✓ **No identified scaling parameters, numerical simulations req.**
- ✓ **2D Transport codes (SOLPS, ..., Soledge2D-EIRENE)**

Often: plasma fluid solver + Monte Carlo kinetic for neutrals

- ✓ **Simulations in relevant regimes are *challenging* (\sim months)**

**$R_{\text{eff}} \sim 1$, neutrals reach fluid limit at some places
(= large number of collisions/particle), ...**

- ✓ **Role of MC noise ? (making things worse, but **how exactly** ?)**

Introduction

What was the common practice so far ?

- ✓ Use **as many particle as « needed »** (decision based on each user's judgement)
- ✓ Use **last time step** as solution
- ✓ Monitor **global balances** (according to a more or less well defined metric) to assess convergence

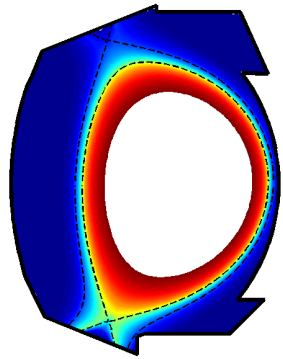
No « theoretical » look at this issue (this talk)

and no error assessment (KUL group, Baelmans et al.)

general architecture of Soledge2d-EIRENE

Particle fluxes on the boundary - Recycling

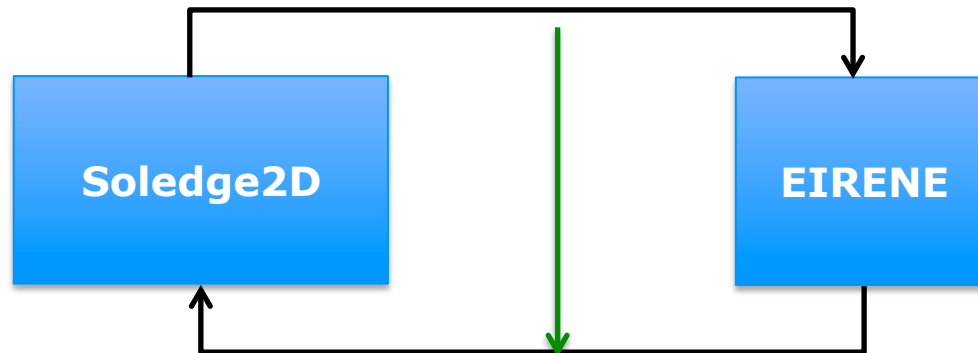
Plasma



2D fluid
URANS

plasma parameters

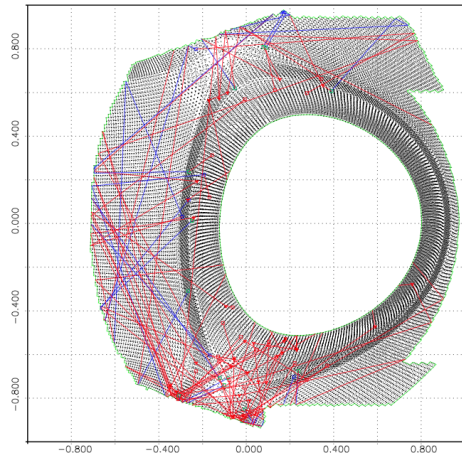
$$\Gamma_{\alpha}, n_{\alpha}, u_{\alpha}, T_{\alpha}$$



$$S_{n\alpha}, S_{m\alpha}, S_{Ei\alpha}, S_{E\epsilon}$$

Volumetric sources

Neutral Gas



3D MC
Linear Boltzmann

- ✓ **Short cycling** scheme essential for code speed-up
(Soledge2D relies on a mixed implicit/explicit scheme)

Outline

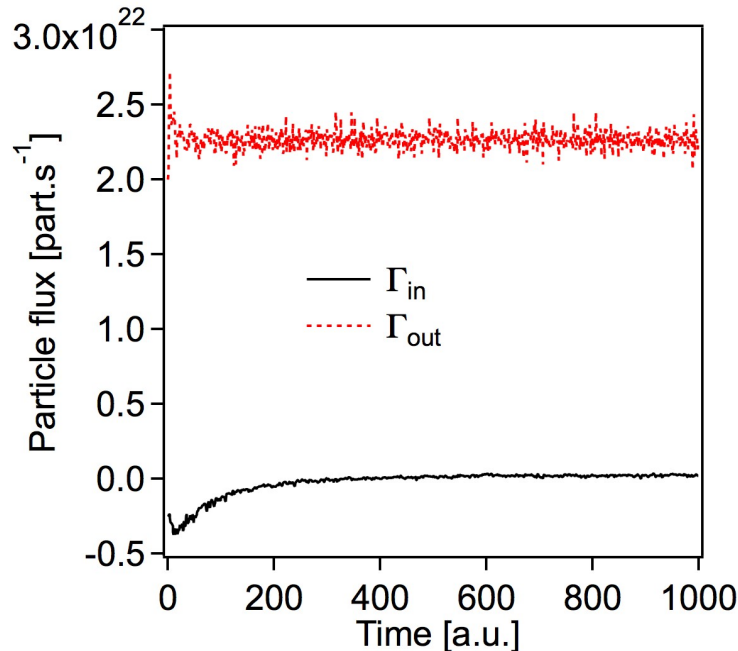
- 1) General considerations on transport codes convergence**
- 2) Simplified model with synthetic noise**
- 3) Effect of noise on the simulations**
- 4) Conclusions and Perspectives**

Outline

- 1) General considerations on transport codes convergence**
- 2) Simplified model with synthetic noise
- 3) Effect of noise on the simulations
- 4) Conclusions and Perspectives

Convergence of fluid/kinetic MC transport codes

Ex. of convergence of particle balance



✓ **The plasma is fluctuating**

how to define a *stationary solution* ?

✓ **“Looks like my favourite stochastic process ...”**

✓ **“Looks like a turbulence code output ...”**

Is there a (useful) connection to be made ?

Take the problem for what it is ...

With MC statistical noise:

- ✓ ***system of stochastic differential equations forced by multiplicative noise***

$$S_n = n(n_0 + \delta n_0) \overline{\sigma v}$$

- ✓ **Probability average $\langle \dots \rangle$, quantities of interest = moments**
- ✓ **1 run = 1 random seed = 1 realization of the stochastic process**
- ✓ **Estimation by ensemble averaging (N runs, can be painful)**

Estimation of moments by time averaging

- ✓ **Similar situation as in turbulence theory (e.g. Monin&Yaglom)**
- ✓ **ensemble average impractical, so we rely on the ergodic theorem**

$$\left\langle \left(\frac{1}{T} \int_{-T/2}^{T/2} n dt - \langle n \rangle \right)^2 \right\rangle \propto \frac{\tau_c}{T}$$

- ✓ **In practice, run the code in the “converged” statistically stationary Steady State (SS) and compute the **mean solution****
- ✓ **Compute **standard deviations** too: measure the dispersion of the solution from time step to time step**

Indication on the distance between the *last time step* and the *mean*

Noise-induced terms in the equation

✓ **Key question:**

how much does the **mean solution** differ from
the **noise free solution** ?

✓ **Equation for the mean density**

$$\nabla \cdot (\langle n \rangle \langle u_{\parallel} \rangle + \Gamma_{\parallel\eta}) \mathbf{b} = \nabla (D \nabla_{\perp} \langle n \rangle) + \langle n \rangle \langle n_0 \rangle [\overline{\sigma \mathbf{v}}]_{\eta}$$

✓ **Extra terms induced by noise – “turbulent fluxes”**

$$\Gamma_{\parallel\eta} = \langle \delta n \delta u_{\parallel} \rangle \quad \textit{Spurious parallel transport}$$

$$[\overline{\sigma \mathbf{v}}]_{\eta} = \langle \overline{\sigma \mathbf{v}}(n, T_e) \rangle + \frac{\langle \delta n \overline{\sigma \mathbf{v}} \rangle}{\langle n \rangle} + \frac{\langle \delta n_0 \overline{\sigma \mathbf{v}} \rangle}{\langle n_0 \rangle} + \frac{\langle \delta n \delta n_0 \overline{\sigma \mathbf{v}} \rangle}{\langle n \rangle \langle n_0 \rangle}$$

Executive Summary

- ✓ The **mean solution** is the **proper solution** to the problem
- ✓ It can be **estimated** by **time averaging** in the **SS**
- ✓ need to run the code for $T \gg \tau_c$ in that phase = **price to pay**
- ✓ It is the solution of an equation with **spurious noise-induced terms**, similar to **turbulent fluxes**
- ✓ These terms can be estimated from the **SS too**:
- ✓ Could ultimately lead to a criteria useful to make sure that noise “does not perturb the solution too much”.

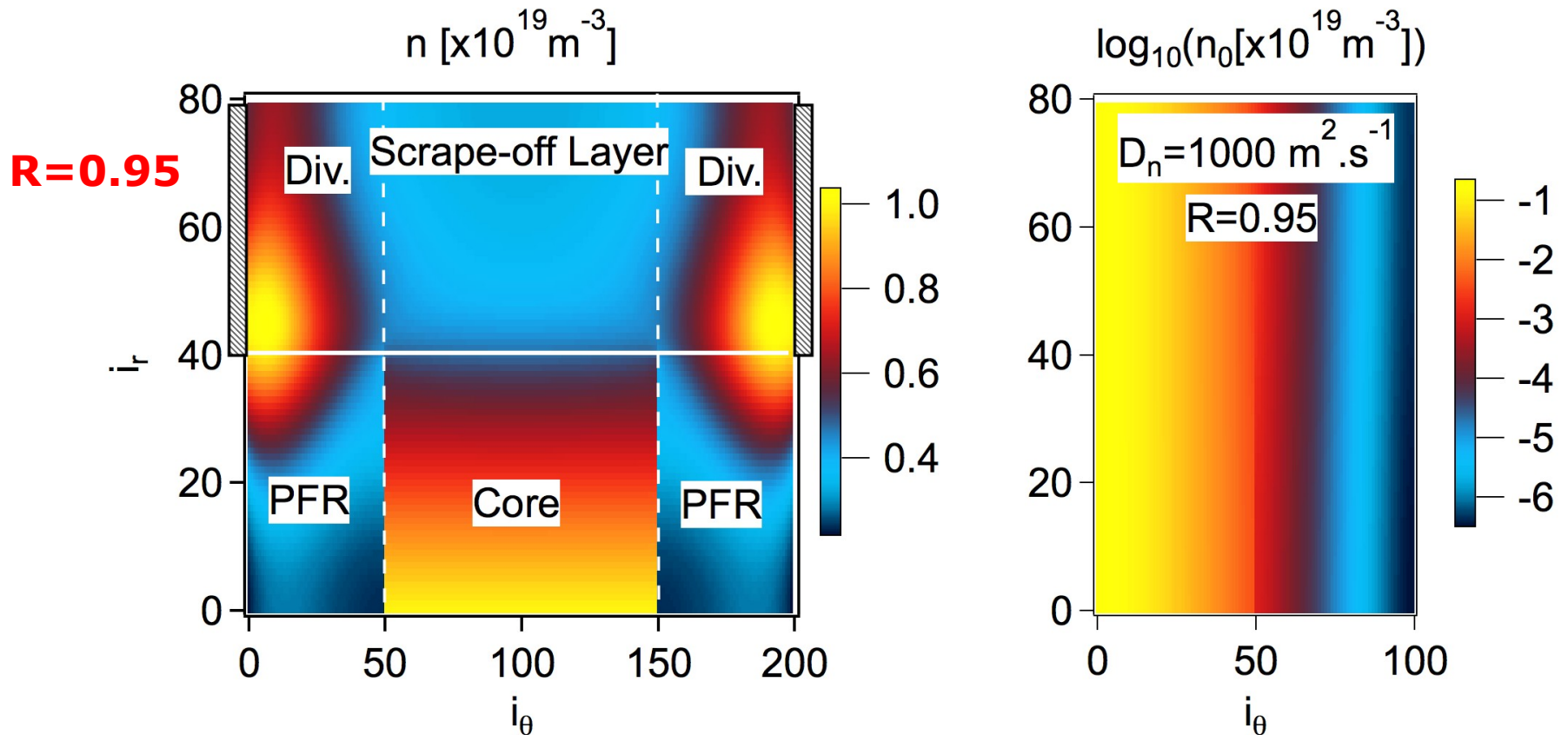
Outline

- 1) General considerations on transport codes convergence
- 2) Simplified model with synthetic noise**
- 3) Effect of noise on the simulations
- 4) Conclusions and Perspectives

Slab case with neutral fluids in SolEdge2D

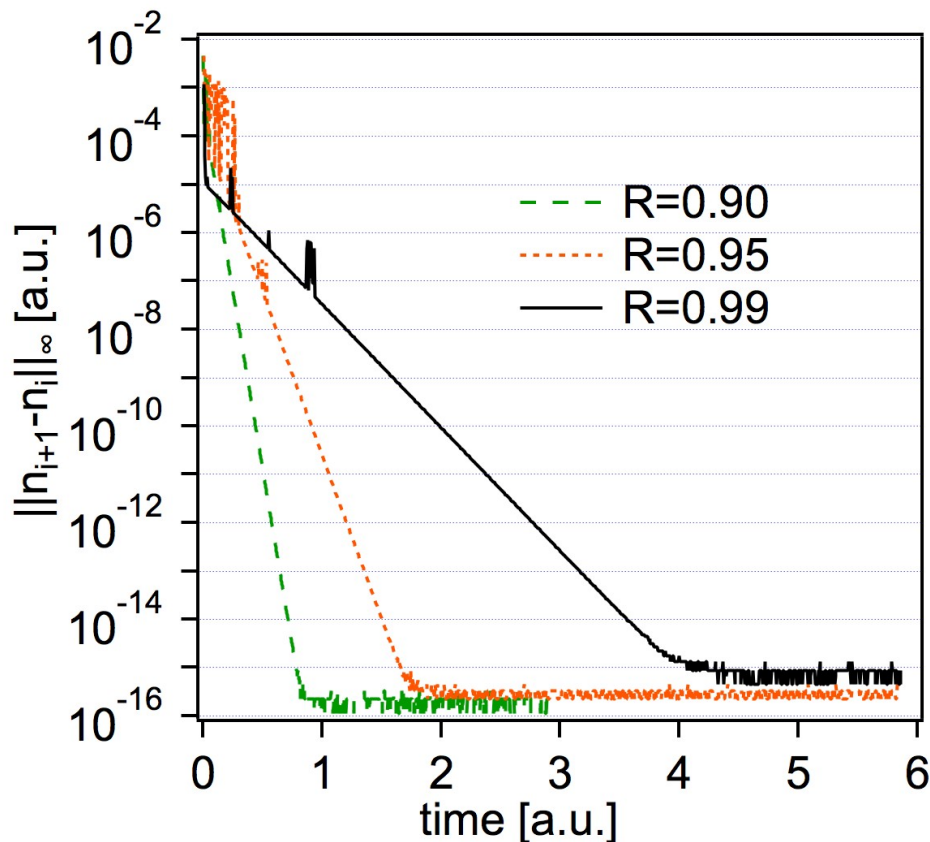
✓ **Soledge2D neutral fluid model:**

$$\partial_t n_0 - D_N \nabla^2 n_0 = -S_n$$



Convergence of the simulations

✓ **No noise: residuals go to machine precision**



✓ **Time scale depends on the recycling coefficient R:**

$$\frac{d\mathcal{N}}{dt} = \Gamma_{in} + S - \Gamma_{out}$$

$$S = R \Gamma_{out}$$

Confinement time τ_0

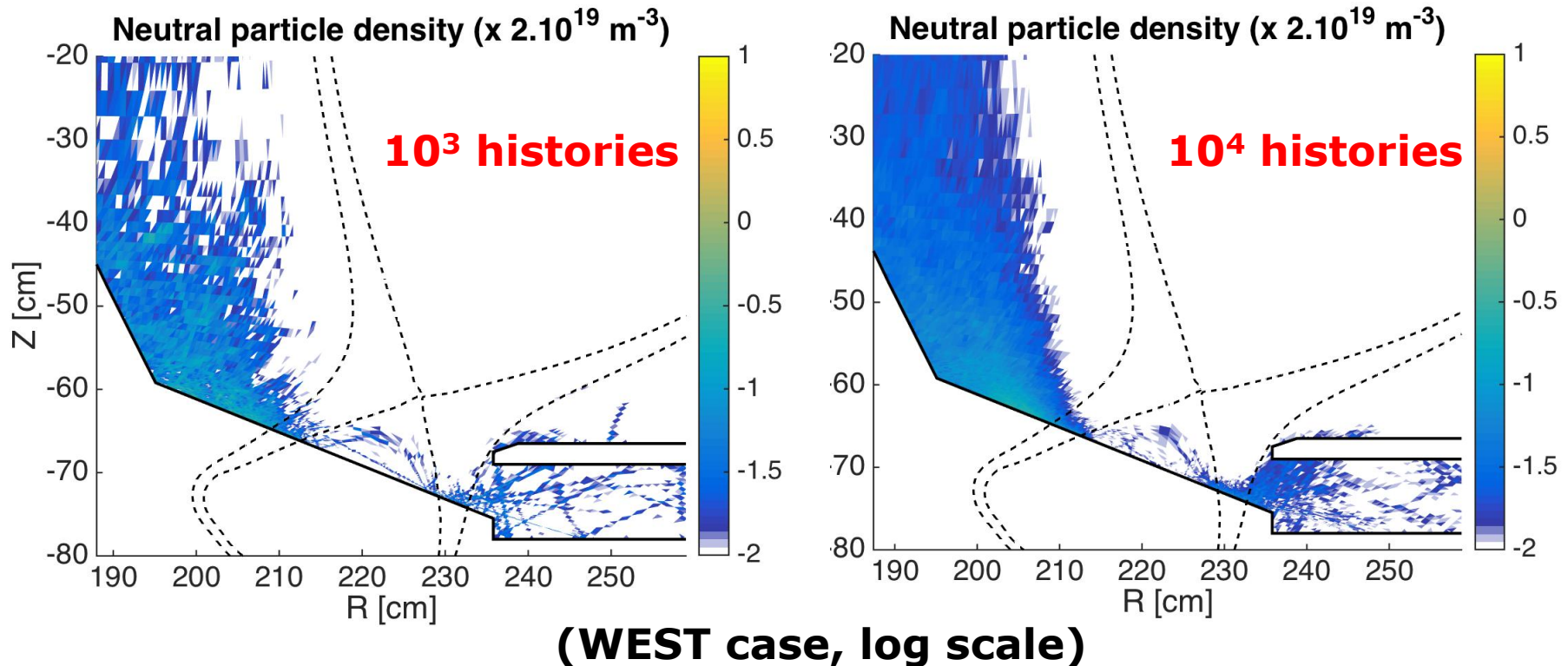
$$\Gamma_{out} = \mathcal{N} / \tau_0$$

$$\frac{d\mathcal{N}}{dt} = -\frac{\mathcal{N}}{\tau^*}, \quad \tau^* = \frac{\tau_0(R)}{1-R}$$

Synthetic noise model

Wish list:

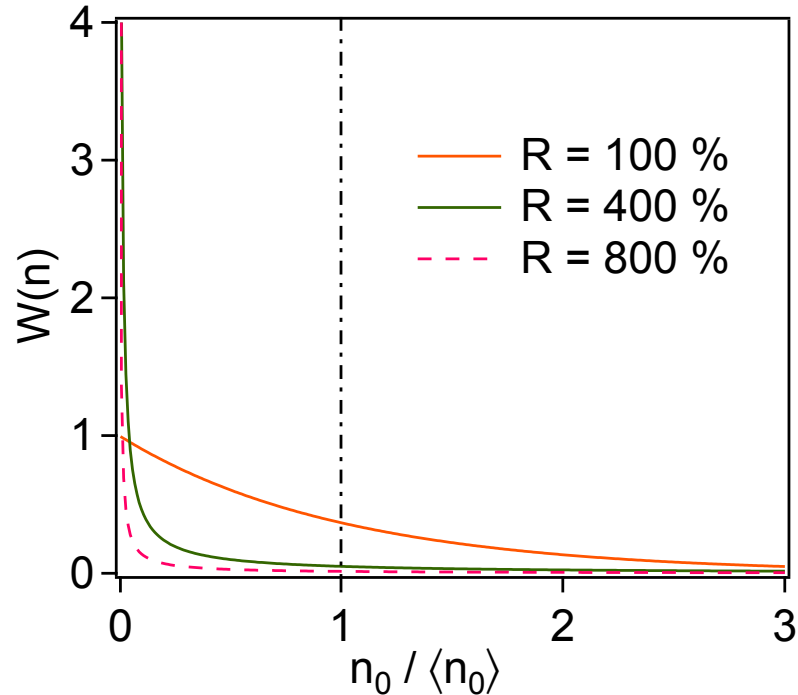
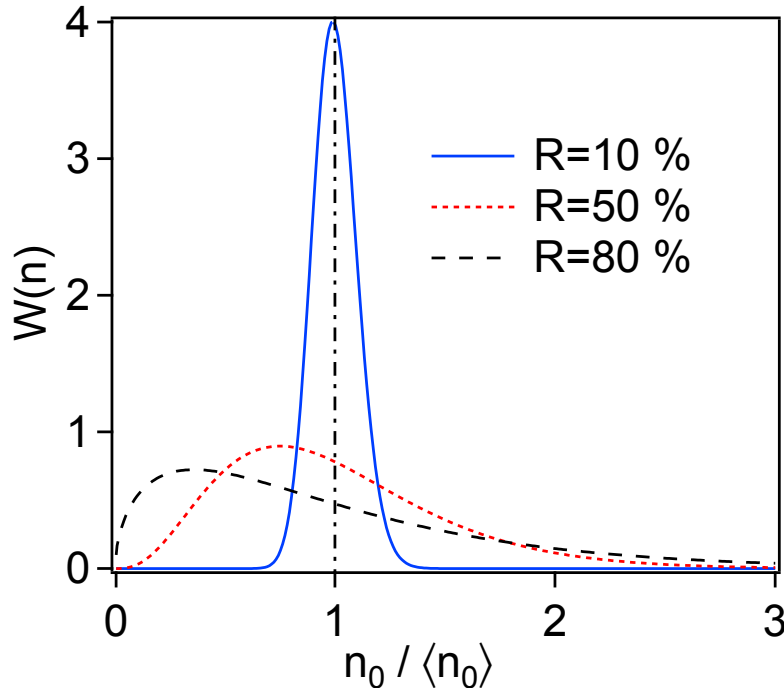
- ✓ **Gaussian** at low noise level (C.L.T.)
- ✓ Providing **positive densities** only at high noise levels
- ✓ Substantial probability for **zero densities**



A good candidate : the gamma distribution

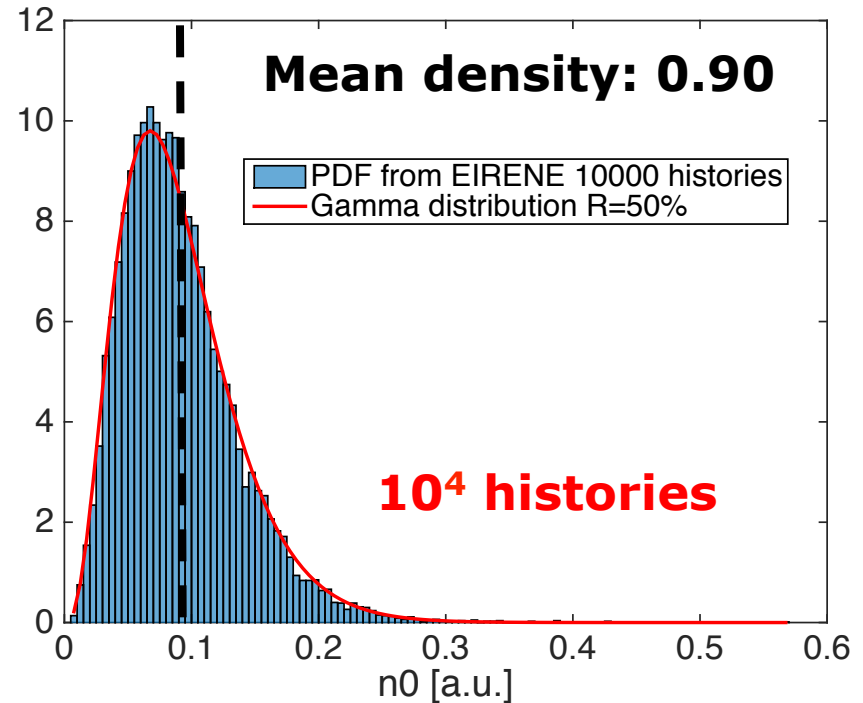
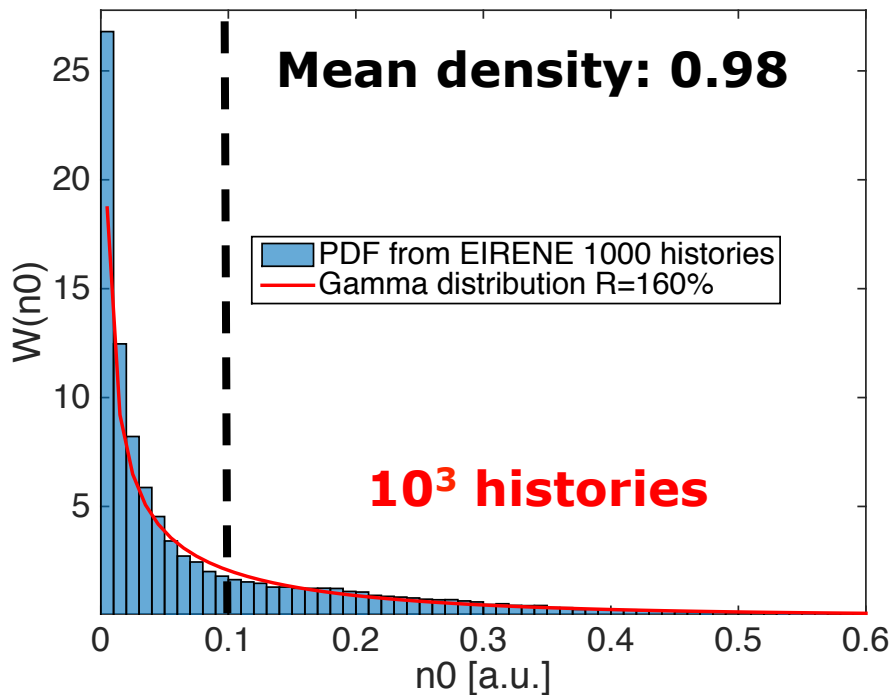
$$W(n_0) = \frac{1}{\Gamma(\beta)\alpha^\beta} n_0^{\beta-1} \exp\left(-\frac{n_0}{\alpha}\right)$$

$$\langle n_0 \rangle = \alpha\beta \quad \sigma_{n_0}^2 = \langle n_0^2 \rangle - \langle n_0 \rangle^2 = \alpha^2\beta \quad \mathcal{R} = \frac{\sigma_{n_0}}{\langle n_0 \rangle} = \beta^{-1/2}$$



How realistic is the gamma model ?

- ✓ **PDF of the neutral particle density in the outer divertor leg (WEST simulations), up to 7×10^4 calls to EIRENE in SS**



Too good to be coincidental ? Erlang distribution ?
= *Sum of exponentially (Poisson) distributed events*

Implementation and examples

- ✓ **Assumption: uniform** fluctuation level \mathcal{R}
- ✓ **Freeze the noise for k iterations : introduce *time correlations***

$$\tau_c = k\Delta t$$

Real life situations :

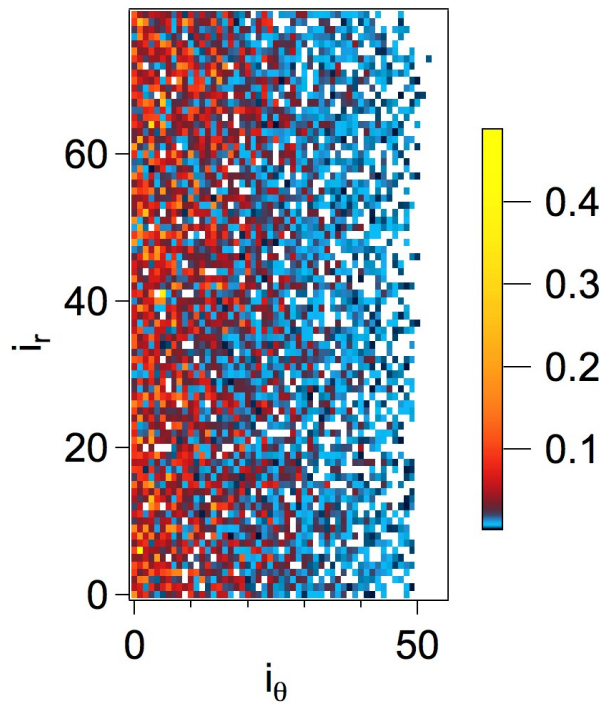
- Not calling MC solver at each time step (“short cycling”)**
- Correlated sampling : freezing noise**

NB: Soledge2D relies on a mixed implicit/explicit scheme

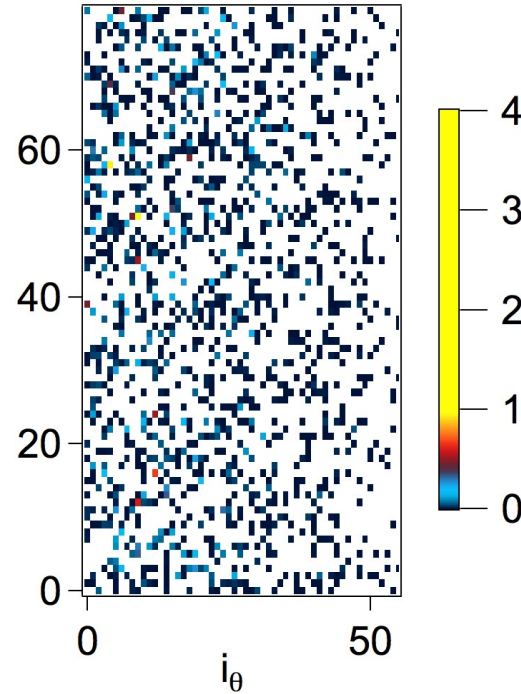
$$\Delta t \simeq 10^{-8} s$$

Examples of neutral particle density maps

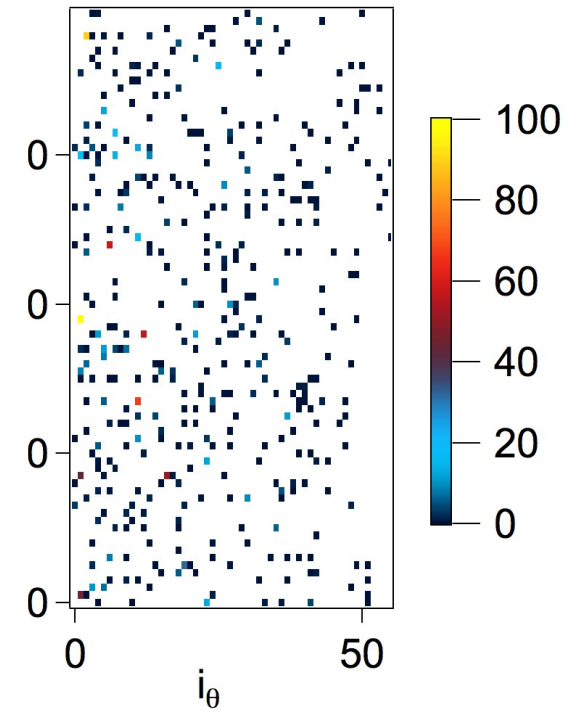
$\mathcal{R}=90\%$



$\mathcal{R}=400\%$



$\mathcal{R}=800\%$

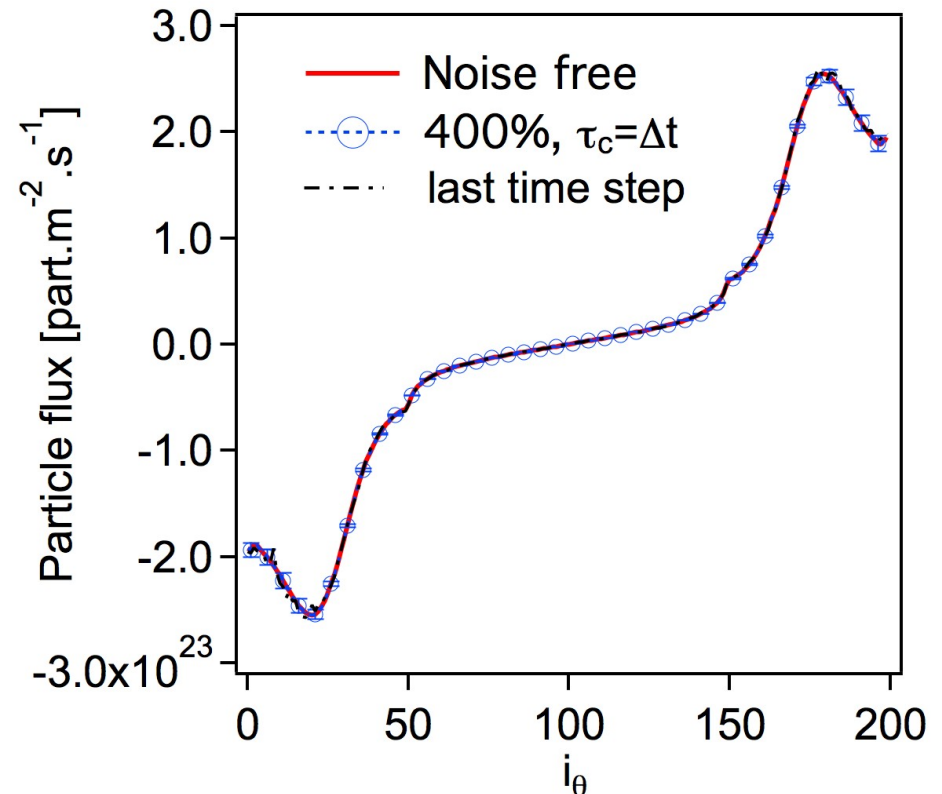
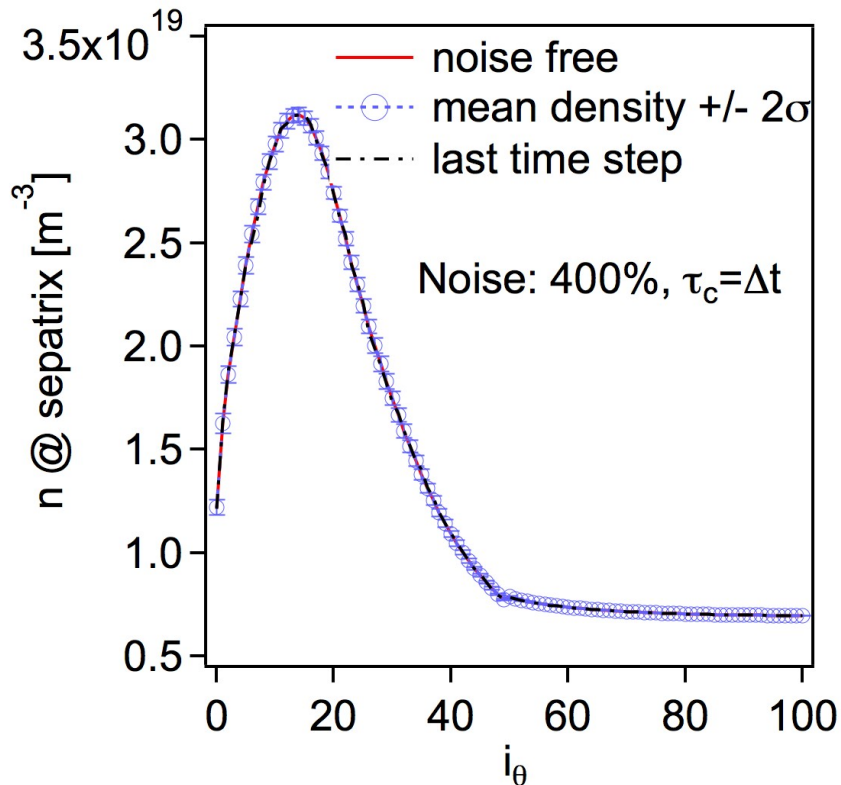


Outline

- 1) General considerations on transport codes convergence
- 2) Simplified model with synthetic noise
- 3) Effect of noise on the simulations**
- 4) Conclusions and Perspectives

The (toy) system is robust to noise

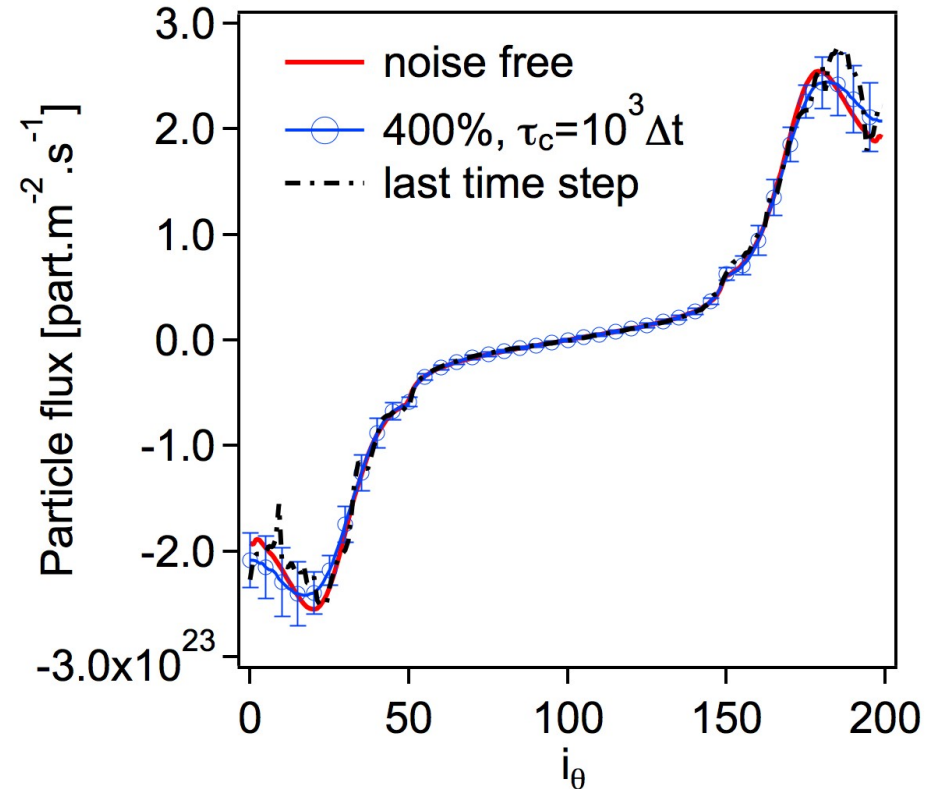
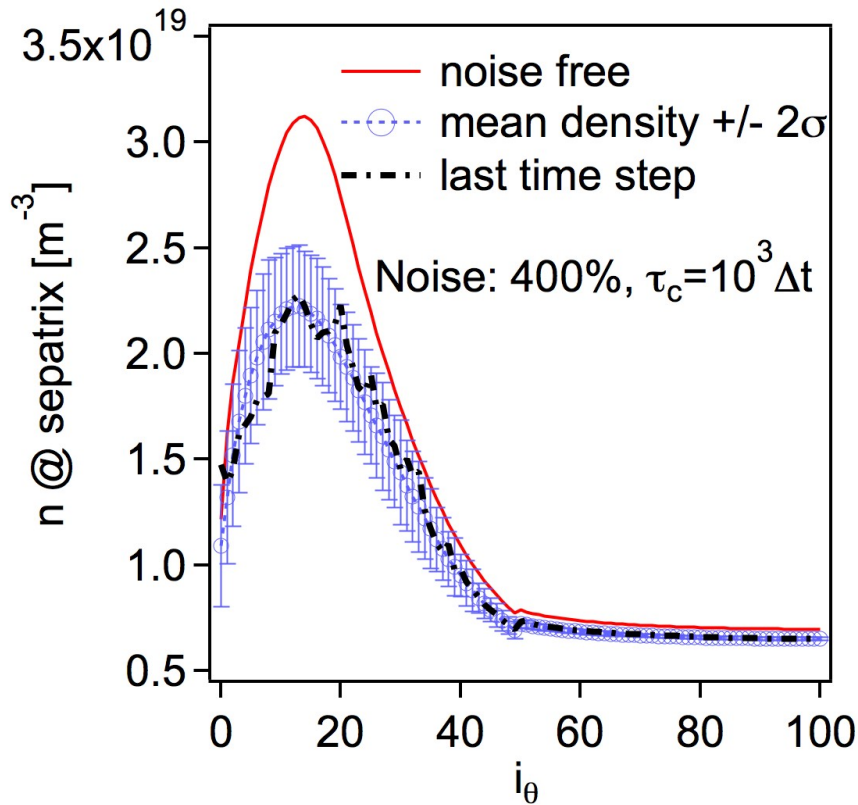
- ✓ **Push the model close to the brink: $R=0.99$, “detached”**
- ✓ **Start with $\tau_c = \Delta t$, ramping up the noise level up to 400%**



Effects of noise *immaterial* even for such strong amplitudes

Role of the noise correlation time

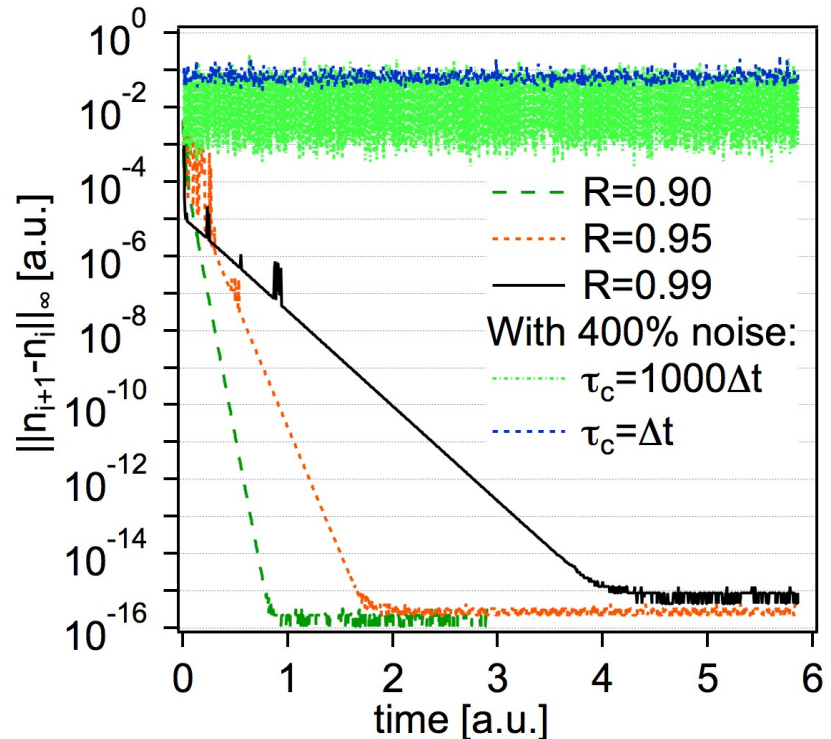
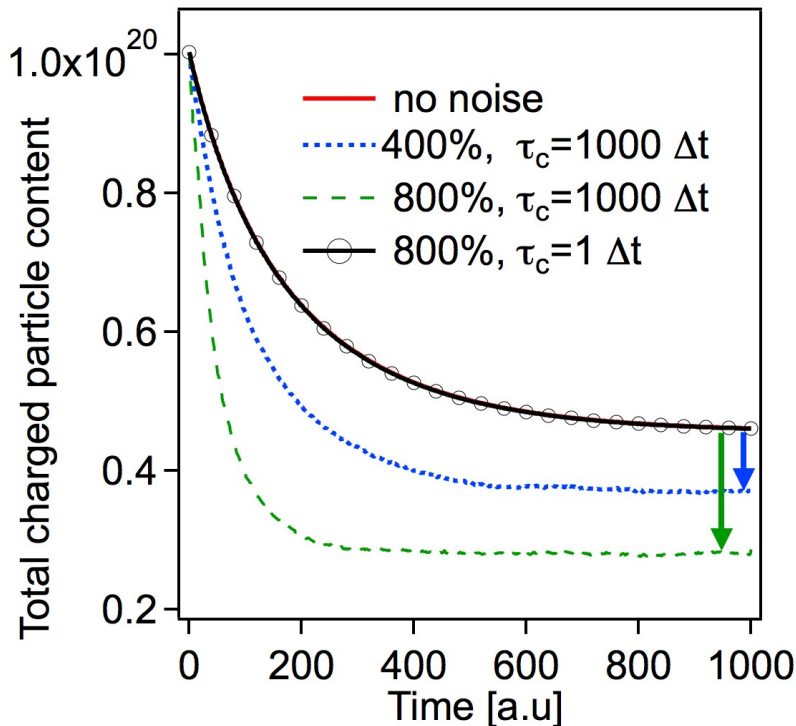
- ✓ Same noise level, but $\tau_c = 10^3 \Delta t$: things start to go wrong ...



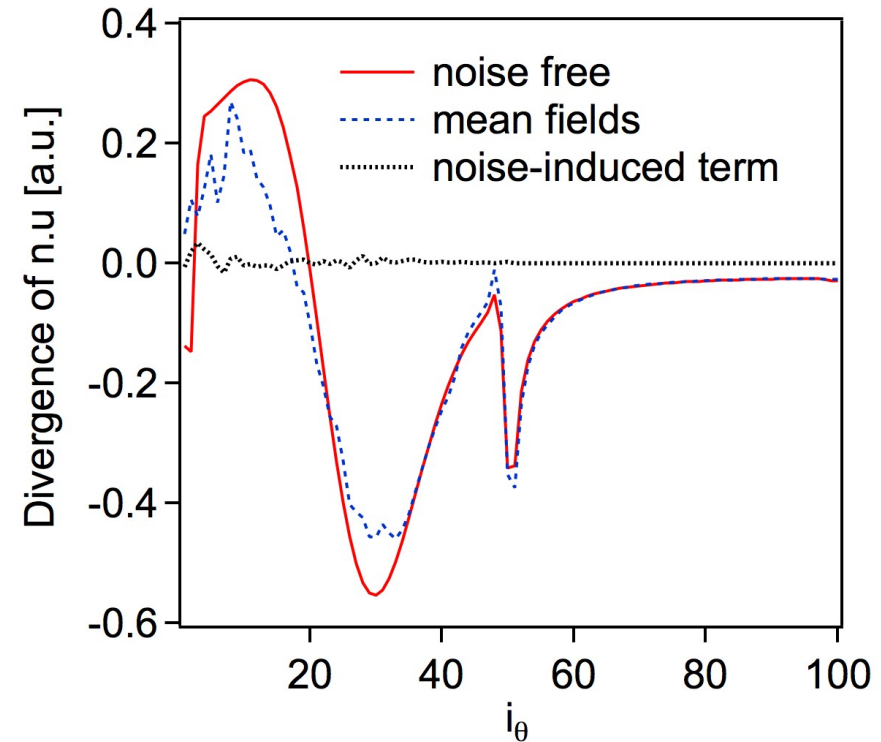
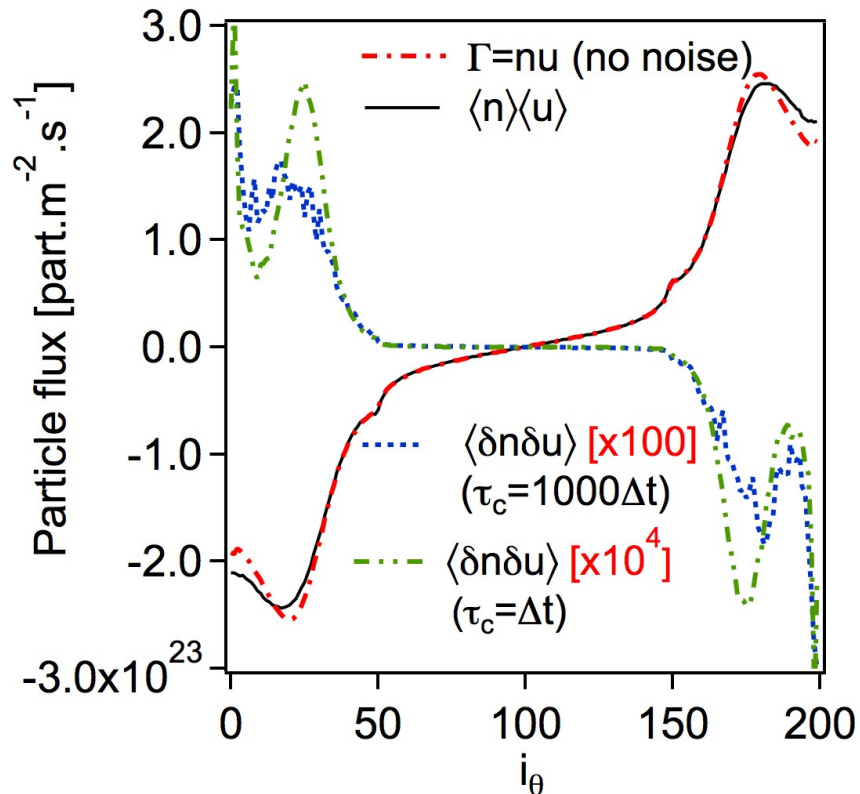
- ✓ Noise with short correlation is **filtered out** by the system
- ✓ Here **frozen long enough** to build strong gradients

Particle balance, total content and residuals

- ✓ **Mean particle balance** $(\langle \Gamma_{in} \rangle + \langle S \rangle - \langle \Gamma_{out} \rangle) / \langle \Gamma_{in} \rangle = 4 \times 10^{-3}$
- ✓ **Total content: τ_0 goes down with noise**
- ✓ **Residuals reflect non-stationarity, and “frustrated” relaxation**



First analysis of noise-induced terms



Another possible culprit :

numerical diffusion in the advection scheme ?

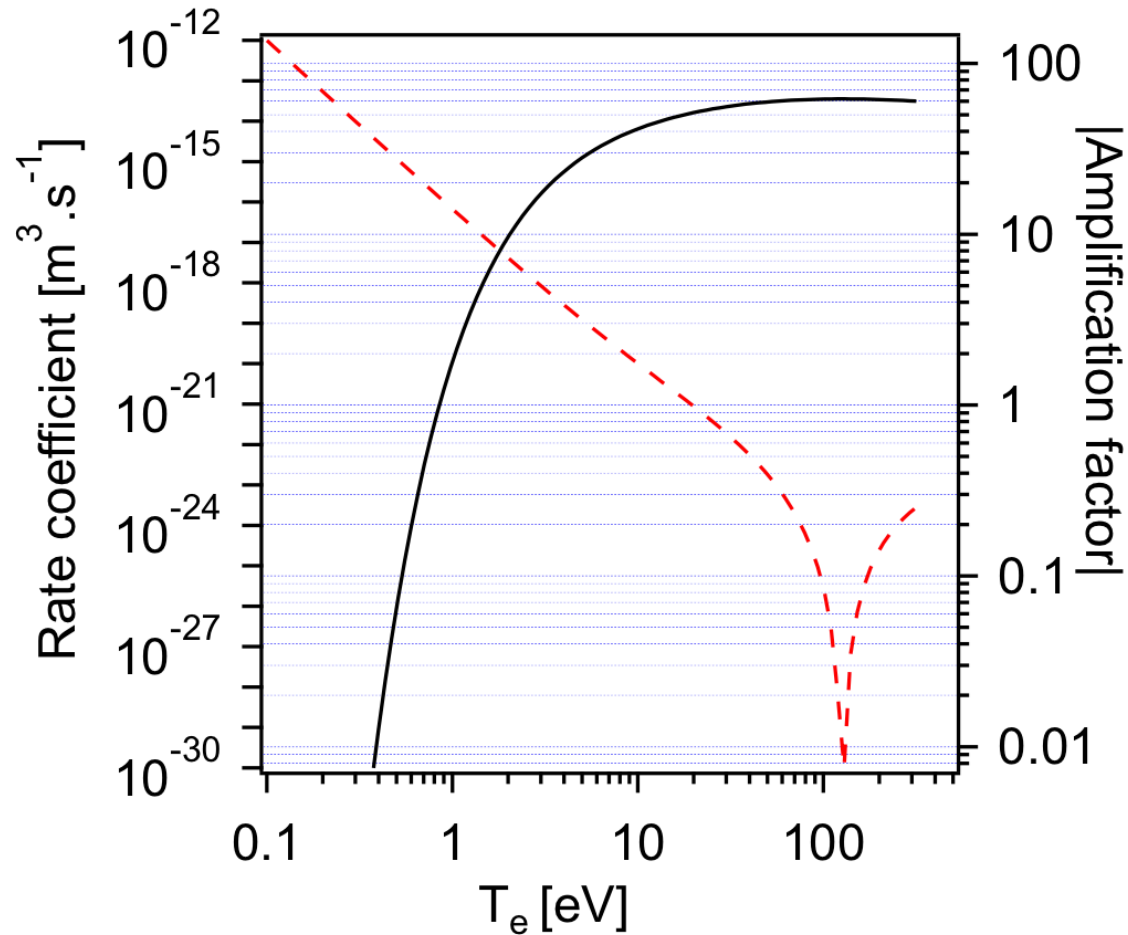
Outline

- 1) **General considerations on transport codes convergence**
- 2) **Simplified model with synthetic noise**
- 3) **Effect of noise on the simulations**
- 4) **Conclusions and Perspectives**

Conclusions perspectives

- ✓ The « converged » SS allows one to define a **proper stationary solution** and evaluate its **distance** to the noise free solution
- ✓ First results with our synthetic noise model suggest **that there is no strong bias unless** relative fluctuation level is **very large**
AND the noise is strongly **time correlated**
(see how this transfers *quantitatively* to real life cases ...)
- ✓ **With this procedure the KUL group has been able to converge SOLPS ITER simulations ~ 50x faster than what was previously done (Baelmans et al., PSI 2016)**
- ✓ **Effects of numerical scheme** need to be assessed too.
- ✓ **More work needed to see whether practical criteria ruling out strong biases from noise** can be established

Noise amplification by non-linearities



Details on the sampling procedure

✓ **After evolution of neutral fluid model $n_0(r, t_{i+1})$**

At each point r_j in space:

sample with mean $n_0(r_j, t_{i+1})$ and s.d. $R n_0(r_j, t_{i+1})$

calculate S_n , S_m and $S_{Ee,i}$ including noise