

Reduction methods for astrochemical networks

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Outline

- Introduction
- Computational problem
- Methods overview
- Flux-based method
- Large network example
- Conclusions

Why chemistry is important (in ISM numerical simulations)

- Needed to compute metal/molecular cooling \rightarrow SF
- Opacity
- Comparison with observations
- ...

Why chemistry is troublesome (in ISM numerical simulations)

- very CPU demanding
- has a non-linear behavior
- chemical networks are complex
- needs accurate rates

Dealing with complexity - Numerical framework

Typical framework:

e.g. Hydrodynamical simulations (1D, 3D, shock, PDR, ...)

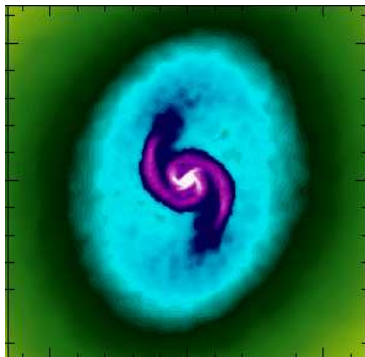
particle \equiv unit of gas ($\approx 10^4 M_{\odot}$)

Each gas particle computes:

- hydrodynamics (e.g. SPH)
- gravity (e.g. tree)

Each gas particle updates:

- chemistry (H , H_2 , e^- , ...)
- gas temperature (Λ , Γ)
- dust (ρ , T , dn/da , ...)

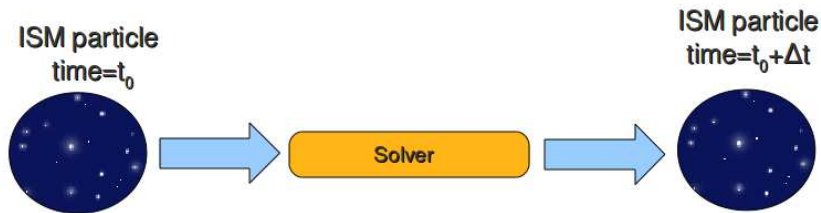


Merlin, Grassi, et al. 2009 A&A



Dealing with complexity - Numerical framework

How to update a gas particle at a given time-step
($\gtrsim 10^6$ particles (gas units) $> 10^4$ time-steps)



- e.g. 1D-hydro: chemistry (large network)
- $\gtrsim 90\%$ of the CPU time

Dealing with complexity - The chemical network

Chemical network = Cauchy's problem:

$$\frac{dn_i}{dt} = \overbrace{\sum_{lm} k_{lm}(T) n_l(t) n_m(t)}^{\text{formation}} - \overbrace{\sum_j k_{ij} n_i(t) n_j(t)}^{\text{destruction}}$$

- $n_i(t=0) = \hat{n}_i$
- $\sum_i n_i(t) \mu_i = \text{const}$

e.g. $A+B \rightleftharpoons C+D$

$$\dot{n}_A = -k_{AB} n_A n_B + k_{CD} n_C n_D$$

$$\dot{n}_C = +k_{AB} n_A n_B - k_{CD} n_C n_D$$

Problem:

Solve an Ordinary Differential Equations System (ODE)

“Pure” computational strategies

- Efficient solver: DVODE/DLSODES (up to $\times 100$)
- Interpolate rate tables (up to $\times 5$)
- Particle Buffering (up to $\times 10$)
- Custom compiler optimization
- Know your code: profiling
- Good programming practices (save divisions, avoid casting, . . .)

Buffering method or “The solver always* rings twice”

- calls to the solver are cpu demanding
- same initial particle conditions lead to same particle evolution
- store already computed chemical evolutions in a buffer
- criterion of similarity: $s_j = \sqrt{\sum_i (B_{ij} - x_i)^2} / N < \xi_b$

$x \leftarrow$ initial conditions

for $j =$ particle in buffer B **do**

if ($s_j < \xi_b$) **then** $\hat{x} = (\hat{x}_j - x_j) dt/dt_j + x$

endfor

if (not found) **then**

$\hat{x} \leftarrow$ solver(x)

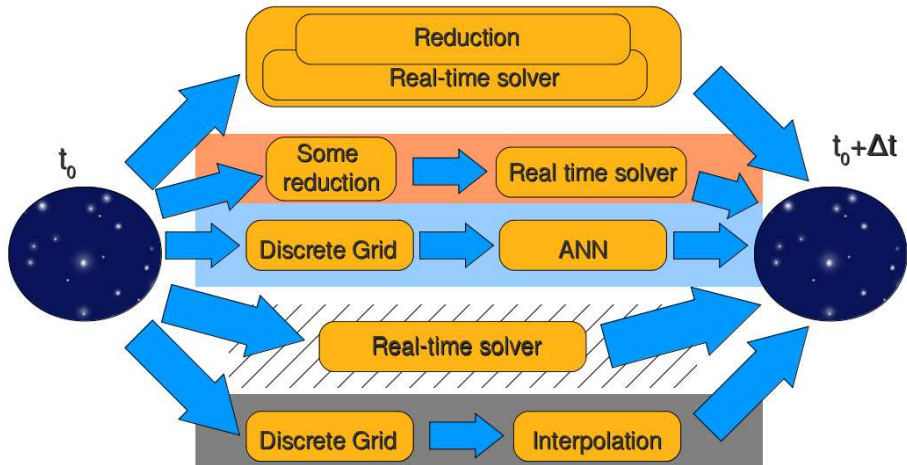
 add $\{x, \hat{x}\}$ to buffer B

endif

*often

Dealing with complexity - Methods/3

Problem: too many reactants (>400) and too many reactions (>4000)!
Methods to solve ODE System:



Dealing with complexity - Flux-based reduction /1

Flux-based reduction

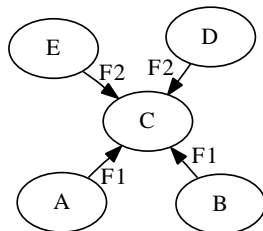
- Flux: $F_i = k_i n_q n_p$
- RHS term: $\dot{n}_i = \sum_{j=1}^N s_j F_j$
- Determining less “active” reactions @ given sub-steps
- i.e. neglect $F_i | F_i < \zeta F_{\max}$
- reduced RHS term: $\dot{n}_i = \sum_{j=1}^M s_j F_j$ where $M < N$

Two-reactions example



- $\dot{n}_C = F_1 + F_2$ $N = 2$

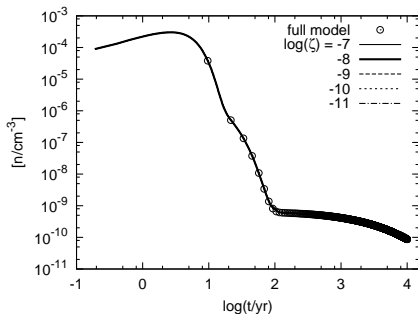
- $\dot{n}_C = F_2$ $M = 1$ ($n_A = 0$)



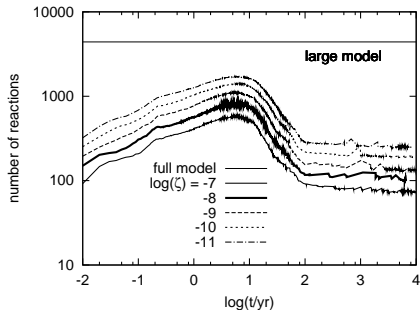
Large model example

- \approx Wakelam&Herbst 2008
- species: 451
- reactions: 4399
- one-zone
- no cooling
- $T = 10$ K
- $\zeta_{\text{CR}} = 1.3 \times 10^{-17} \text{ s}^{-1}$
- $A_V = 10$

Results $x_C(t)$:



speed-up: $\times 10$



(details in Grassi, Bovino et al. 2012 MNRAS)

Dealing with complexity

- Computational astrochemistry needs reduction methods
- “Pure” computational methods are required
- Other strategies allow large simulations
- Flux-based method is the best choice*

Next talk by Stefano Bovino: *a priori* methods!

*for the moment

Thank you for your attention!

Also thanks to: F.A.Gianturco, D. Schleicher, S. Bovino