

Chemodynamics in Dense Gas

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Gothenburg, Sweden 2021

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to my family

Abstract

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Keywords: Product configuration, constraint satisfaction, Boolean satisfiability, knowledge compilation, supervisory control theory.

Acknowledgments

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Chia-Jung Hsu
Göteborg, December 2012

List of Publications

This thesis is based on the following appended papers:

Paper 1. Chia-Jung Hsu, Jonathan C. Tan, Matthew D. Goodson, Paola Caselli, Bastian Körtgen, Yu Cheng *Deuterium Chemodynamics of Massive Pre-stellar Cores* Accepted by MNRAS (2020).

List of Acronyms

ISM – Interstellar Medium
PSC – Prestellar Core

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Part I

Introductory chapters

Chapter 1

Introduction

Star is the unit of bright objects and the major sources of energy in the universe. However, it is still hard to know how/where/how often the star formation occurs. As the collection of materials pervading between stars in galaxies, interstellar medium (ISM) has a directly connection and contribution to star formation. In the lives of galaxies, the interstellar medium can gradually be converted to stars. When the stars reach the ending of their lifetimes, they return some materials by stellar winds, or supernova explosions. The returned materials could contain dusts and heavy metals and increase the complexity in the galaxies. Stars also change the local properties of ISM, such as the temperature and radiation field. Besides the stellar feedbacks, the interstellar medium could also accretes mass from or lose mass into intergalactic medium (IGM). Overall, ISM is complex and it is important to study its structure and its influences to star formation.

1.1 Interstellar Medium and Star Formation

Because of the diversity of local properties, ISM can be roughly categorized into following several phases.

- Coronal gas: also named as hot ionized medium (HIM). In this phase, gas has been shock-heated to $T \gtrsim 10^{5.5}$ K by the blastwave of supernova explosion.
- HII gas: In this phase, hydrogen gas is photoionized by ultraviolet photons from hot stars. Depending on its density, it could be called as "HII region" (high density, $n_H \sim 0.03\text{cm}^{-3}$) or "diffuse HII" (low density, $n_H > 100\text{cm}^{-3}$, also named as "warm ionized medium" or WIM)
- Warm HI: also called as warm neutral medium (WNM). The temperature is heated to $T \sim 10^{3.7}$ K, and the density is about $n_H = 0.6\text{cm}^{-3}$
- Cool HI: Predominantly atomic gas temperatures $T \sim 10^2$ K. Densities $n_H \sim 30\text{cm}^{-3}$. also called cold neutral medium or CNM.

- Diffuse molecular gas: Similar to the cool HI clouds, but with higher densities and column densities so that H_2 self-shielding strong enough to keep H_2 in the cloud interior
- Dense molecular gas: gravitationally bound clouds having $n_H \gtrsim 10^3$

The huge molecular gas structure, which is likely gravitationally bound (e.g., Roman-Duval et al., 2010; Tan et al., 2013a), can also be identified as Giant Molecular clouds (GMCs). Typically, GMCs have mass $\gtrsim 10^4 M_{odot}$ with mean mass surface density $\Sigma = M/(\pi R^2) \sim 0.02 \text{ g cm}^{-2}$ or $(100 M_\odot \text{ pc}^{-2})$, and mean volume density $n_H \sim 100 \text{ cm}^{-3}$ (McKee and Ostriker 2007; Jonathan C. Tan, Kong, et al. 2013). The temperature of GMCs could range between 10-50 K. Due to the cold and dense properties, GMCs are responsible to star formation (McKee and Ostriker 2007; Kennicutt and Evans 2012). Substructures also exist in GMCs because of the gravitational instability. Clumps are self-gravitating structures which star clusters form (**Williams2000**) and cores are the smallest units where star formation occurs within clumps (Bergin and Tafalla 2007; Jonathan C. Tan, Beltran, et al. 2014). A prestellar core could collapse to one or a few stars. To further study star formation, we need more ways to trace and resolve these dense structures.

1.1.1 Massive Star Formation

Massive stars play an important role in the universe because of the strong radiation feedback and the heavy metals forged in the cores. However, the formation mechanism of massive stars is still unclear. Their short lifetimes and relative rarity make observational studies challenging. Candidate theories to explain massive star formation include Turbulent Core Accretion and Competitive Accretion. Turbulent Core Accretion (or Turbulent Core Model) (McKee and Jonathan C Tan 2003) is a scaled-up version of low-mass star formation theory (Shu et al. 1987). The model proposes that a combination of supersonic turbulence and magnetic fields supports massive pre-stellar cores (PSCs) against fragmentation and that these then collapse to a central star-disk system. However, the collapse is not necessarily as ordered as in the case of low-mass star formation, especially if there is significant turbulence in the PSC. The Competitive Accretion model of (Bonnell et al. 2001) proposes that large numbers of low-mass stars form in a protocluster clump, with a few of them later accreting chaotically, by Bondi-Hoyle accretion of gas supplied by the collapsing clump, to become massive stars. To distinguish these two theories, one of the decisive differences is in the existence of massive, coherent PSCs.

1.2 Virial Theorem

Virial Theorem is a way to identify whether a system is gravitationally bound. From classical mechanics, a system of particles is in virial equilibrium if the kinetic energy T and the potential energy U satisfy the relationship:

$$2T + U = 0 \quad (1.1)$$

In the case of clouds/clumps/cores, the kinetic energy is approximated by velocity dispersion. The virial parameter α is then defined by

$$\alpha = 5\sigma^2 R / (GM) = 2aE_k/E_G \quad (1.2)$$

where M , R are the mass and radius of the cloud/clump/core, and σ is the one-dimensional velocity dispersion. a is the ratio of gravitational energy, E_G , to the gravitational energy of a uniform sphere, $3GM^2/5R$ (Bertoldi and McKee 1992). The meaning of the virial parameter is the ratio of twice of kinetic energy (estimated by one-dimensional velocity dispersion) to the gravitational energy of uniform sphere. Although astronomical objects are not always uniform density and contain thermal and magnetic field energy, the virial parameter provides a simple way to know whether the system is bounded. A system is said subvirial if $\alpha < 1$ and super virial if $\alpha > 1$.

1.3 Mass-to-Flux Ratio

Unlike the turbulence motion, magnetic field contribute the pressure to gas in another way and cannot be seen from the motion. To know whether the cloud/clump/core can collapse in a magnetic field, a critial mass associated with magnetic field is defined(Mouschovias and Spitzer, L. 1976):

$$M_\Phi = \Phi / (2\pi G^{1/2}) \quad (1.3)$$

The mass-to-flux ratio then is given by:

$$\mu_\Phi = \frac{M}{M_\Phi} = \frac{2\pi G^{1/2}}{\Phi} \quad (1.4)$$

If $\mu_\Phi > 1$, then the gas is supercritical and cannot prevent collapse.

Chapter 2

Astrochemical Modelling

Astrochemistry studies the formation and destruction of species in the universe. Comparing the observational data with modeling results can help us study the physical environments around our interested objects. It is an important tool to probe the universe.

Currently, several public databases are provided with a variety of modelling codes to study astrochemistry. Databases include KIDA and UMIST. Associating with this databases and even appending with some additional reactions, several astrochemical codes like AstroChem, Nahoon, Nautilus, UCLCHEM are also public on the internet. These codes may base on different models, especially in the aspect of grain-surface reactions, but all of them solve the chemical reaction in the approach of rate equations. The main difference among them is the diversity of reaction rate formalism.

There are mainly two different chemical models in our study. One is the deuterium fractionation network, which mainly comes from KIDA database. The other one is a reduced network extracted from UCLCHEM code, which is a mixture of UMIST database and their own grain model.

2.1 Deuterium Fractionation

According to Big Bang Nucleosynthesis theory, the abundance ratio of D/H is about 1.6×10^{-5} . If there is no deuterium fractionation, the D/H fraction in every H-bearing species are expected to be about 1.6×10^{-5} , too. However, on the earth, comets or in some observed prestellar cores, the D/H fractions of certain molecules are much higher than the theoretical value. These evidences support that deuterium fractionation occurred. Several observations have revealed that deuterium fractionation occurs in prestellar cores, where provides a cold and dense environment. The formed deuterated species could be inherited by the following protoplanetary disks and planets, comets. However, the detailed procedure is still under debate.

The deuterium fractionation process can be roughly divided into three steps. Figure 2.1 shows the steps in graph. The first step is the formation of H_3^+ , which

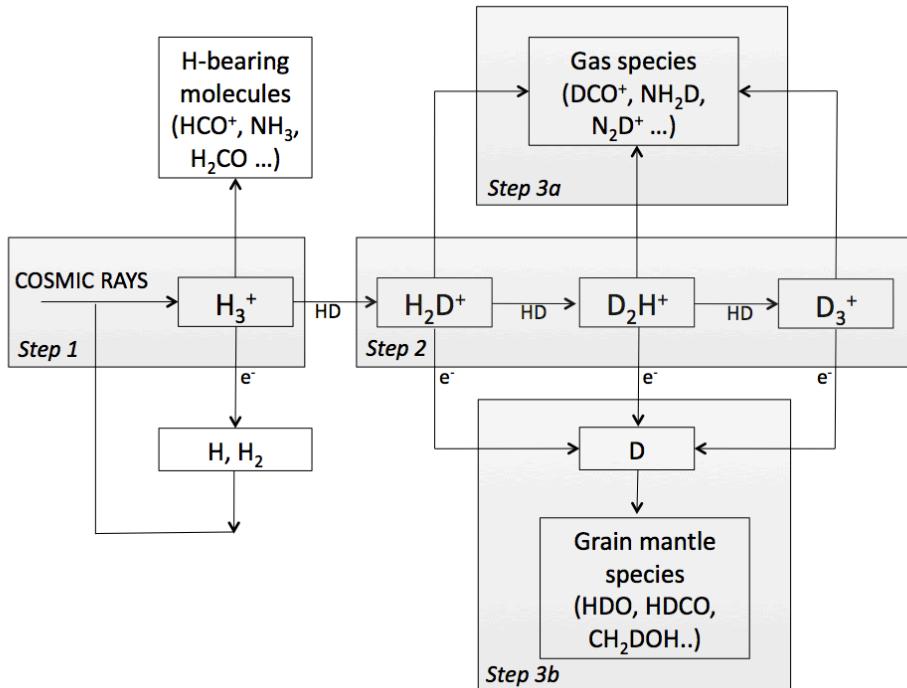


Figure 2.1: Deuterium Fractionation

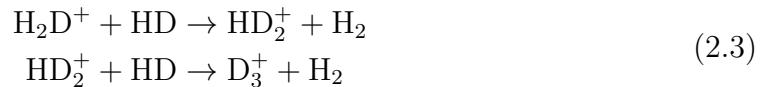
comes from the reaction:



Because of the charged property, H_3^+ is more active than neutral species. The formed H_3^+ then reacts with HD , which is the main reservoir of D-atoms, and form H_2D^+ .



The inverse reaction could also happen to destroy H_2D^+ to form HD . However, there is a small activation barrier for this inverse reaction. If the temperature is low enough, the environment has a tendency to form H_2D^+ . H_2D^+ then could continue react with HD to form HD_2^+ and D_3^+ .



The formed H_2D^+ , HD_2^+ and D_3^+ then could react with species in gas phase, or release D-atom on grain mantles by dissociative recombination to transfer D-atoms to other species.

Because H_2D^+ is the central role to transfer the D-atoms into species and its easier to exist in cold gas, deuterium fractionation is easier to be observed in cold gas. Since the formation of H_2^+ depends on H_3^+ , it means higher cosmic-ray ionization rate can accelerate the reactions. Besides, because H_3^+ could be destruct by O

and CO, removing these species from the gas-phase can also enhance the deuterium fractionation. That is to say, the cold and dense region like prestellar cores, where the O and CO tend to stick on grain mantles, is an ideal place for deuterium fractionation.

Another important factor influence the deuterium fractionation is the ortho-and-para ratio of molecular hydrogen (OPR^{H_2}). The internal energy of ortho- H_2 could overcome the barrier in the inverse reaction of Equation 2.2 (i.e. $\text{H}_2\text{D}^+ + \text{H}_2 \rightarrow \text{H}_3^+ + \text{HD}$). As a consequence, the ratio of $\text{H}_2\text{D}^+/\text{H}_3^+$ is suppressed and stop the third step to occur. Theoretically, the H_2 formed on dust grains has 75% probability to form ortho- H_2 and 25% probability to for para- H_2 , but the ratio decreases in gas phase as time evolves because of proton-exchange reaction. The OPR^{H_2} is believed to be low in prestellar cores (Sipilä, Caselli, et al. 2013; Brünken et al. 2014).

Observationally, N_2H^+ and its deuterated form N_2D^+ have been recognized as good tracers for prestellar cores. They have been oberved in L1544 and several cores in Infrared Dark Clouds.

A variety of astrochemical studies have been carried out to model deuterium fractionation. (Walmsley et al. 2004) considered a reduced chemical network, including the spin states of H_2 , H_2^+ , H_3^+ and H_2D^+ . This network assumed heavy elements, like C, N, O, etc, are fully depleted. Extending this work, (Flower et al. 2006), (Edouard Hugo et al. 2009), (Pagani2009) and (Sipilä, E. Hugo, et al. 2010) included updated reaction rates for spin states and deuterated forms of H_2 and H_3^+ . (Vastel et al. 2012) presented networks including molecular species with up to three atoms. (Kong et al. 2015) extended these works to include H_3O^+ to acquire more precise results. As a consequence, the abundances of electrons, water, HCO^+ , DCO^+ , N_2H^+ , N_2D^+ are improved and have a good agreement with the even more extensive network of (Sipilä, Caselli, et al. 2013). More recently, (Majumdar et al. 2016), based on the work of (Wakelam et al. 2015), presented a complete network including spin state chemistry with 13 elements (H, He, C, N, O, Si, Fe, Na, Mg, Cl, P, F).

In our work, we based on the network from (Kong et al. 2015), which approximated the surface chemistry by gas-phase reactions.

2.2 UCLCHEM

UCLCHEM (Holdship et al. 2017) is a time-dependent gas-grain chemical model written in Fortran. It extends the pure gas-phase reactions from UMIST database with gas-grain interactions including freeze-out, thermal desorption, photodesorption and cosmic-ray-induced thermal desorption based on the rate functions in (Roberts et al. 2007). To simulate the surface chemistry, it adopts an approximated way to allow users to set the branch ratio of freeze-out reactants. For example, when CO stick on grain mantles, it could be hydrogenated and form CH_3OH . UCLCHEM allows users to set, for example, 90% CO to become grain-phase CO and the rest 10% CO to become CH_3OH in their defined grain file. The whole package also includes a routine to generate the chemical modeling code and other hydrodynamics routine like C-shock and molecular clouds. However, the hydrodynamics modules are limited in single grid or one dimension.

The standard chemical network defined in UCLCHEM includes hundreds of gas-grain reactions besides the 6173 gas-phase reactions. However, if we only focus on simple species like CO, H₂O, etc, a reduced network can provide similar results as the complete network. In the simulations of ISM scale, we mainly use the reduced network generated from UCLCHEM.

Chapter 3

Chemodynamics Simulation

Although there are more and more sophisticated astrochemical models, these models are usually implemented in one zone or one dimension. However, it means that these models ignore the kinematics influences, especially the advection diffusion induced by turbulence. To further understand the kinematics influence of the turbulence motion and even have the synthetic observations, three-dimensional chemodynamics simulations are necessary.

3.1 Coupling Chemistry with Magnetohydrodynamics

In a typical ideal Magnetohydrodynamics (MHD) simulation, the following equations are solved:

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{v}) = 0, \quad (3.1)$$

$$\frac{\partial \rho \mathbf{v}}{\partial t} + \nabla \cdot (\rho \mathbf{v} \mathbf{v} + \mathbf{I} P - \mathbf{B} \mathbf{B}) = -\rho \nabla \phi, \quad (3.2)$$

$$\frac{\partial E}{\partial t} + \nabla \cdot [(E + P) \mathbf{v} - \mathbf{B}(\mathbf{B} \cdot \mathbf{v})] = -\rho \mathbf{v} \cdot \nabla \phi - \Lambda + \Gamma, \quad (3.3)$$

$$\frac{\partial \mathbf{B}}{\partial t} - \nabla \times (\mathbf{v} \times \mathbf{B}) = 0 \quad (3.4)$$

where $\rho, \mathbf{v}, P, E, \mathbf{B}, \phi$ are the density, velocity vector, pressure, energy density and magnetic field vector and potential. \mathbf{I} is a identity tensor. Λ and Γ represent the radiative cooling and heating rates. To include the contribution of magnetic field, the pressure and energy are given by:

$$P = p + \frac{B^2}{2}, \quad (3.5)$$

$$E = e + \frac{\rho v^2}{2} + \frac{B^2}{2}, \quad (3.6)$$

where p and e are thermal pressure and thermal energy density. The magnetic permeability (μ_0) is unity in these equations. In the case of ideal gas with a ratio of specific heats γ , the thermal energy can be given by:

$$e = \frac{p}{(\gamma - 1)} \quad (3.7)$$

If the ideal gas is isothermal, the thermal energy can also be given by sound speed (c_s):

$$c_s = \sqrt{\frac{\gamma k T}{\mu m_H}} \quad (3.8)$$

$$e = \frac{\rho c_s^2}{\gamma(\gamma - 1)} \quad (3.9)$$

Equations 3.1 ~ 3.3 represent the conservation of mass, momentum, and energy and Equation 3.4 is the magnetic induction equation. However, to extend a fluid to a reactive flow, the equation of specie evolution must to be included:

$$\frac{\partial n_i}{\partial t} + \nabla \cdot (n_i \mathbf{v}) - \nabla \cdot (\mathcal{D} \nabla n_i) = C(n_i, n_j, T) - D(n_j, T) n_i \quad (3.10)$$

where n_i represent the number density of each species and the summation of number density weighted by mass must equal to mass density ($\sum_i n_i m_i = \rho$). This equation includes three processes: advection, diffusion and chemical reactions. The second and the third term in the left-hand side represent the advection and the diffusion process. In astrophysical flows, the diffusion term is usually ignored since its influence is much smaller than advection term. The right-hand side means the construction (C) and destruction (D) of i th species.

Equation 3.10 can be splitted into two parts:

$$\frac{\partial n_i}{\partial t} + \nabla \cdot (n_i \mathbf{v}) = 0 \quad (3.11)$$

$$\frac{dn_i}{dt} = C(n_i, n_j, T) - D(n_j, T) n_i \quad (3.12)$$

The first equation only consider the advection term and the second equation solves the reactions, which is the same set of differential equations handled by those one-zone astrochemical model (See Chapter 2). That is to say, to implement a chemodynamics simulation, we can make the hydrodynamics code calculate the advection and then insert the astrochemical model to solve the reactions.

3.2 Simulation Code: ENZO+KROME

In our projects, we utilize **ENZO**(Bryan et al. 2014) and **KROME**(T. Grassi, Bovino, Schleicher, et al. 2014) to run the chemodynamics simulations. **ENZO** is a adaptive mesh refinement (AMR) hydrodynamics code developed for cosmological simulations.

It has built-in functions using **GRACKLE** chemistry library to calculate the evolution of primordial gas. **GRACKLE** also supports calculating heating/cooling rates of primordial gas and metals and some UV background radiation effects. However, the main purpose of **GRACKLE** is the primordial chemistry. Although it can be extended to include some user-defined **Cloudy** heating/cooling table, it is not practical to make it include a general chemical network of ISM.

Instead, **KROME** is a chemistry package working as a Python-based parser to convert a user-provided chemical network to Fortran codes. Users have the freedom to define their own reaction format by using its pre-defined tokens. It provides a unique interface function so that it can work with other hydrodynamics codes. However, it does not guarantee the consistency with hydrodynamics codes. Users have to ensure that all species in the network exists in the hydrodynamics codes and that the advection is handled by the hydrodynamics codes (see Section 3.1). Besides, because it generates source codes, these codes have to be compiled with the hydrodynamics codes to make it work. It also provides some official patches for several hydrodynamics codes, including **FLASH**, **ENZO**, **GIZMO** and **RAMSES**. For **ENZO**, the patch does not handle the advection and the existence of species. Users have to define the additional species and modify some routines for advection and renormalization (e.g. `Grid_SolveHydroEquations.C` for PPM, Zeus, MHD_Li/CT solvers and `Grid_UpdatePrim.C` for Runge Kutta solvers).

3.3 Performance

Although **KROME** provide a way to set up the codes for chemodynamics simulations, a significant problem is the size of the network. To model the chemistry happens in ISM, the networks usually contain thousands of reactions. Even if it may only take a few milliseconds to finish one step in one grid, the cost becomes significant when there are 10^6 (the order of 128^3 resolution) grids. Figure 3.1 shows the mean time of one step of chemodynamics simulations in a 64^3 cube. The simulations use the same initial condition but couple with different sizes of networks. Obviously, the computational time is almost proportional to the number of reactions. Note that the performance could also be influence by the initial conditions. For example, if the temperature of some grids is out of the valid range reactions, the reaction rates are zeros and could make the calculation faster. However, in general, the performance of simulations could be thousand times slower if a network containing thousands of reactions is used.

Another problem of chemodynamics simulation is the number of species. To track the evolution and calculate the advection, the density of each species must be stored in each grid. As the number of species increases, the simulation also occupies more disk storages and memories. For example, a magnetohydrodynamics simulation usually stores about 10 fields (typically, ρ , \mathbf{v} , E , \mathbf{B} and some optional fields like internal energy or potential). However, a network of ISM could contain hundreds of species. For example, if the network contains about 100 species, the simulation with the resolution of 128^3 will use about 2GB memory. If the resolution is enhanced to 512^2 , the simulation data will exceed 100GB. The number of species

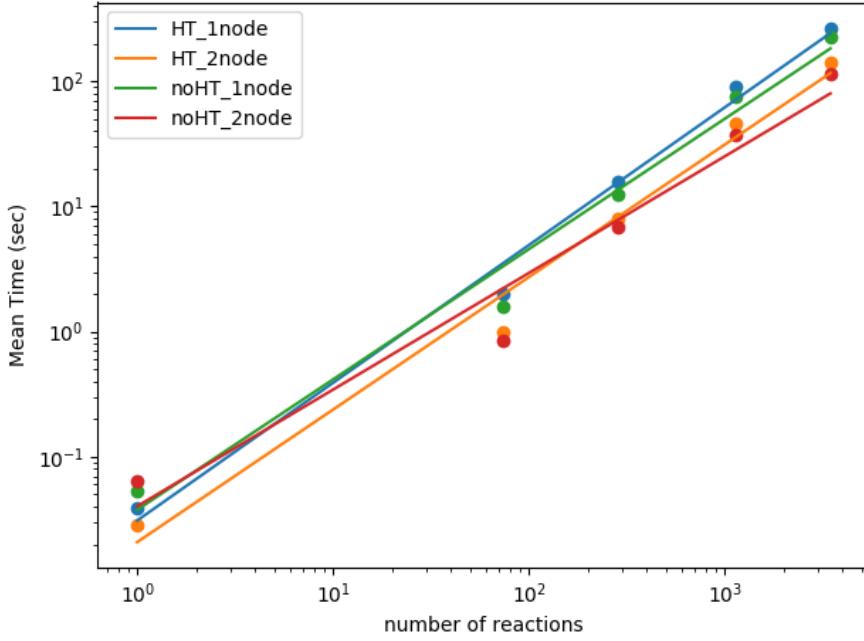


Figure 3.1: Time consumption

limits the resolution and increase the loading of communication when the simulation running on clusters.

A simple way to save memory and disk storage is using the AMR structure to limit the finer grids focusing on subregions in the simulation. The way is effective if the interesting regions are only a small portion of the whole domain. However, the side effect is the interpolation error. Although most of popular AMR codes support high order interpolation methods and conservative interpolation to reduce the numerical error, the elementary conservation still could be broken(T. Grassi, Bovino, Haugbølle, et al. 2017). The problem is because gradient limiters and interpolation weights could be different for each species. Even if each individual chemical species is conserved during the interpolation, the summation of atoms in species could not be conserved. This means the amount of species should be normalized twice after advection. One is for this elementary conservation and another one is for the mass conservation.

However, even if the memory requirement is affordable, the performance problem is still significant. It is unpractical to wait over forty days for a simulation originally taking one hour. Several ways could improve the performance:

- Parallelization: Parallelization is a common way to improve the performance of simulations. Currently, `openmp` and `MPI` have been widely applied in numerical codes. ENZO also support the usage of `MPI`, which can accelerate the simulation by requesting more processors from more nodes in cluster. Figure 3.1 also shows the simulation is almost twice faster by using two nodes rather than one node. However, it is also unpractical to offload the heavy computation by requesting hundreds of nodes. Graphics card could be another choice of parallelization. By using the graphics processing units (GPU) on graphics cards, hydrodynamics

simulations could be ten times faster than CPU(Schive et al. 2010). Note this also depends on the frequency and the number of units. Usually this kind of tests are done under the same “cost” of a specific cluster. Several studies are working on moving the differential equation solvers onto GPU (**Niemeyer2014**; **Curtis2016**; **odeint**; **cuda-sim(culsoda)**; **Stone2018**). However, these studies are focusing on combustion engineering and biochemistry. They have not been ported to astrochemistry and have several drawbacks. Most solvers are explicit and cannot handle the level of stiffness in astrochemistry. Some of them use the first order BDF method and lose the accuracy. The higher order solver does not handle the sparsity matrix causes the uncertainty on performance.

- Reducing the size of network: Because the computational load is proportional to the number of reactions, it is intuitive to reduce the load by reducing reactions. The idea also has been studied for a long time. It can be done by pre-selecting species and related reactions(Nelson and Langer 1999). Another approach assumes that some species stay on equilibrium to focus on solving the non-equilibrium species(**Glover2010**; Lam 1993), where those equilibrium species are chosen by the rapid reaction rates. The approach may not be valid for a wide range of parameters because the float of reaction rates. A further approach then try to dynamically reduce the reactions to avoid the limitation(Tupper 2002; Grassi et al. 2012).
- Reducing the time steps of chemistry: This could be the simplest way to accelerate the simulations, but the method only works when the chemistry timescale is much longer than the hydrodynamics timestep. If the hydrodynamics timestep is much shorter chemistry timescale, the chemistry calculation can be operated every several hydrodynamics timesteps and keep a good approximation. This approach has been implemented in our ENZO codes. However, if the simulation is under a fast heating/cooling, this approach will lose the accuracy.
- Neural network: Due to the advance of deep learning, neural network becomes a popular skill in a variety of fields. As the fundamental neural network, artificial neural network also has been widely applied in computational chemistry, biochemistry, etc(Goncalves et al. 2013). A recent work also apply the skill to create a emulator of UCLCHEM(De Mijolla et al. 2019). As a black box, the emulator skip the cost of solving differential equation but reproduce the nonlinear results. The problems become how to train a valid model and guarantee the correctness in a wide range of parameters and whether there is a general way to train a effective model.

Chapter 4

Introduction to Paper

4.1 Paper I

In this paper, we focus on the deuterium fractionation in massive prestellar cores. The reason is that N_2H^+ and its deuterated form N_2D^+ have been observed in several prestellar cores and are trusted to be good tracers(Caselli2002; Kong2016; Jonathan C. Tan, Kong, et al. 2013). The cold and dense gas is also an ideal place of deuterium fractionation (See Section 2.1). Their emission lines are also useful to interpret the kinematics in the cores. As a follow-up work of (Goodson et al. 2016), we also use the chemical network in (Kong et al. 2015) but with slightly updates from (Majumdar et al. 2016). The network contains species composed by heavy elements, C, N, and O, so that allow us to investigate the influence of depletion factor. We then use ENZO and KROME to couple the chemical network with magnetohydrodynamics simulation of prestellar core.

We focus on the influences of chemical parameters to an identical prestellar core. The initial conditions of the prestellar core are fixed to mass = $60 M_\odot$, surface density = 0.3 gcm^{-2} , radius = 0.1 pc. The free-fall time is expected to be 76 kyr. The prestellar core also have a velocity turbulent field and a cylindrically symmetrical magnetic field to make the core slightly super virial in the beginning. Since there is no driving turbulence field, the core will still collapse. The chemical parameters we studied are the initial OPR $^{\text{H}_2}$, temperature, cosmic-ray ionization rate and depletion factor of CO and N (two factors, one controlling the abundances of C and O, another one control the N abundance) and their composition.

We follow the chemical evolution in the core for 0.8 free-fall time and analyze the average number densities, column densities of N_2H^+ , N_2D^+ . We conclude that one model with high cosmic-ray ionization rate (1.0^{-16}s^{-1}) and high depletion factor ($f_D^{\text{CO}} = 1000$, $f_D^N = 100$) is the best model matching with the observational data in our simulations. The abundances also allow us to estimate the velocity gradient, the velocity dispersion, even the rotational energy, the kinetic energy and the virial parameter of the simulated core under the assumption of optical thin of $\text{N}_2\text{D}^+(3-2)$ emission line. Overall, the rotational energy is little comparing with the gravitational energy, and the velocity gradient need more observational data for comparison. Besides, the core appears subvirial observationally during most time of evolution

because of the contribution of magnetic field cannot be observed from velocity dispersion.

Chapter 5

Outlook

5.1 Ambipolar Diffusion in Prestellar Cores

Following the work done in Paper 1, we plan to include the non-ideal MHD effect, ambipolar diffusion, into the simulation. If we consider the ambipolar diffusion effect, the prestellar core could take longer time to collapse and probability a lower cosmic-ray ionization rate still can reproduce the high deuterium fraction in observation. Also, due to the cosmic ray and depletion factors, the mean ion mass could be reduced and then enhance the ambipolar diffusion effect.

5.2 Cloud Collision

GMC collisions are thought as a mechanism to trigger star formation. (Wu, Loo, et al. 2015; Wu, Jonathan C. Tan, et al. 2017) has use `ENZO` and `GRACKLE` to study the influence of heating/cooling of photon dissociation region (PDR) model in GMC collisions. The heating/cooling function could help the formation of dense structures which are the progenitors of stars. The PDR model also allow the prediction of molecular lines like CO(8-7) emission line. However, these models does not really couple with chemistry and make the prediction have some uncertainty. We plan to couple the simulation with the reduced network of `UCLCHEM` to further improve the accuracy of CO and probe other tracers of dense gas (HCN, HCO⁺). Because the gas-grain interactions are also supported by the network, it also allow us to study the depletion in dense gas. The results can be used to improve the initial condition of the prestellar core simulation.

5.3 Porting astrochemical tools onto GPU

The performance issue has been addressed in Section 3.3. To run a huge network in high resolution, improving the performance is inevitable. So far, because the chemical networks are much slower than hydrodynamics, we use the way reducing the cycle to accelerate the simulation. To further improve the performance, we would like to have a high order implicit ODE solver executable on GPU. We first plan to

port the available `lsoda` GPU solver to fit the requirement of astrochemistry. `lsode` is another solver has been widely use and take advantage of the sparsity of matrix, but it is written in Fortran77. Porting it onto GPU could be another project.

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Part II

Appended papers

Paper 1

Deuterium Chemodynamics of Massive Pre-stellar Cores

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Paper 1. Deuterium Chemodynamics of Massive Pre-stellar Cores

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Deuterium Chemodynamics of Massive Pre-Stellar Cores

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ABSTRACT

High levels of deuterium fractionation of N₂H⁺ (i.e., D_{frac}^{N₂H⁺ \gtrsim 0.1) are often observed in pre-stellar cores (PSCs) and detection of N₂D⁺ is a promising method to identify elusive massive PSCs. However, the physical and chemical conditions required to reach such high levels of deuteration are still uncertain, as is the diagnostic utility of N₂H⁺ and N₂D⁺ observations of PSCs. We perform 3D magnetohydrodynamics simulations of a massive, turbulent, magnetised PSC, coupled with a sophisticated deuteration astrochemical network. Although the core has some magnetic/turbulent support, it collapses under gravity in about one freefall time, which marks the end of the simulations. Our fiducial model achieves relatively low D_{frac}^{N₂H⁺ \sim 0.002 during this time. We then investigate effects of initial ortho-para ratio of H₂ (OPR_{H₂}), temperature, cosmic-ray (CR) ionization rate, CO and N-species depletion factors and prior PSC chemical evolution. We find that high CR ionization rates and high depletion factors allow the simulated D_{frac}^{N₂H⁺ and absolute abundances to match observational values within one freefall time. For OPR_{H₂}, while a lower initial value helps the growth of D_{frac}^{N₂H⁺, the spatial structure of deuteration is too widespread compared to observed systems. For an example model with elevated CR ionization rates and significant heavy element depletion, we then study the kinematic and dynamic properties of the core as traced by its N₂D⁺ emission. The core, undergoing quite rapid collapse, exhibits disturbed kinematics in its average velocity map. Still, because of magnetic support, the core often appears kinematically sub-virial based on its N₂D⁺ velocity dispersion.}}}}

Key words: astrochemistry – hydrodynamics – methods:numerical – stars:formation

1 INTRODUCTION

Massive stars play an important role in the universe. Their radiative, mechanical and chemical feedbacks influence the evolution of interstellar media and thus star formation and galactic evolution. However, the formation mechanism of massive stars is still unclear. Their short lifetimes and relative rarity make observational studies challenging. Candidate theories to explain massive star formation include Turbulent Core Accretion and Competitive Accretion. Turbulent Core Accretion (e.g., McKee & Tan 2003) proposes that a combination of supersonic turbulence and magnetic fields supports massive pre-stellar cores (PSCs) against fragmentation and that these then collapse to a central star-disk system. However, the collapse is not necessarily as ordered as in the case of low-mass star formation, especially if there is significant turbulence in the PSC. The Competitive Accretion model of Bonnell et al. (2001) proposes that large numbers of low-mass stars form in a protocluster clump, with a few of them later accreting chaotically, by Bondi-Hoyle ac-

cretion of gas supplied by the collapsing clump, to become massive stars. To distinguish these two theories, one of the decisive differences is in the existence of massive, coherent PSCs.

One way to identify PSCs and also characterise their dynamical history is via the deuterium fractionation of certain chemical tracers. In particular, N₂H⁺ and its deuterated form, N₂D⁺, have been observed in PSCs (e.g., Caselli et al. 2002; Tan et al. 2013; Kong et al. 2016). Due to the cold ($T < 20$ K) and dense ($n_{\text{H}} > 10^5 \text{ cm}^{-3}$) environments of PSCs, the deuteration process gradually increases the deuterium fraction of N₂H⁺ ($D_{\text{frac}}^{\text{N}_2\text{H}^+} \equiv [\text{N}_2\text{D}^+]/[\text{N}_2\text{H}^+]$). Also, as deuteration of N₂H⁺ is sensitive to the ortho-to-para ratio (OPR) of H₂, which is expected to gradually decline after formation of molecular gas, D_{frac}^{N₂H⁺ could be a proxy to estimate the chemical age of PSCs.}

A variety of astrochemical studies have been carried out to model deuterium fractionation. Walmsley et al. (2004a) considered a reduced chemical network, including the spin states of H₂, H₂⁺, H₃⁺

and H_2D^+ . This network assumed heavy elements, like C, N, O, etc, are fully depleted. Extending this work, Flower et al. (2006a), Hugo et al. (2009), Pagani et al. (2009) and Sipilä et al. (2010a) included updated reaction rates for spin states and deuterated forms of H_2 and H_3^+ . Vastel et al. (2012) presented networks including molecular species with up to three atoms. Kong et al. (2015) extended these works to include H_3O^+ to acquire more precise results. As a consequence, the abundances of electrons, water, HCO^+ , DCO^+ , N_2H^+ , N_2D^+ are improved and have a good agreement with the even more extensive network of Sipilä et al. (2013). More recently, Majumdar et al. (2016), based on the work of Wakelam et al. (2015), presented a complete network including spin state chemistry with 13 elements (H, He, C, N, O, Si, Fe, Na, Mg, Cl, Ph, F).

The astrochemical models described above have been utilised mostly in single-grid or one-dimensional simulations. However, with this limitation, these simulations could not follow the chemodynamical influence of turbulence, which requires ideally three-dimensional simulations. To investigate such effects, Goodson et al. (2016) developed an approximate model to estimate the levels of $[\text{N}_2\text{H}^+]$ and $[\text{N}_2\text{D}^+]$, and thus the degree of deuteration. This study used pre-computed tables of chemical evolution of the single-grid models of Kong et al. (2015) to parameterize the growth rate of N_2H^+ and N_2D^+ via the abundances of each of them. With this approximate model, the coupled deuterium fractionation chemistry with 3-D magnetohydrodynamics (MHD) simulations of massive turbulent, magnetized PSCs was studied. However, in addition to its approximate nature, the limitations of this method are the lack of information upon the abundances of other species.

In contrast, the study of Körtgen et al. (2017) followed the dynamical influence on deuterium fractionation by coupling the chemical network of Walmsley et al. (2004a) to PSC hydrodynamic simulations. Recently, the synthetic observations of one selected model from Körtgen et al. (2017) are performed in Zamponi et al. (2020). Then, in a follow-up work, the deuterium fractionation of magnetized and turbulent filaments was investigated by Körtgen et al. (2018). These studies coupled the deuterium fractionation network to MHD simulations by utilizing the KROME package (Grassi et al. 2014) to solve the chemical reactions in each time step of the simulation. Some limitations of this work include the assumption of fully-depleted heavy elements and the use of threshold-based sink particles to replace high density regions. Furthermore, the increasing size of the chemical reaction network limits the simulation efficiency. In a recent work, Bovino et al. (2019) also applied KROME to investigate deuterium fractionation on multiple scales of clumps and cores. They considered heavy elements, like C, N, and O and focused on discussing the influence of time-dependent depletion/desorption reactions to CO depletion structures inside the clumps and cores.

In this paper, we re-examine the work of Goodson et al. (2016) and further discuss the influence of other chemical parameters in the standard, fiducial PSC model of the Turbulent Core Accretion theory of McKee & Tan (2003). We discuss our numerical methods, including choices of initial conditions and chemical modeling, in §2. The results are presented in §3, including the basic results of the fiducial model (§3.1), the effects of different chemical conditions (§3.2), average core properties and comparison to observed systems (§3.3), and core structure, kinematics and dynamics (§3.4). A summary and our conclusions are given in §4.

2 NUMERICAL METHODS

We use the adaptive mesh refinement (AMR) code ENZO (v2.5) (Bryan et al. 2014) to run our magnetohydrodynamics (MHD) simulations and adopt the MUSCL-Dedner method with Harten-Lax-van Leer (HLL) Riemann solver to solve the ideal MHD equations. The PSC is assumed to be isothermal and collapses under its self-gravity. We set the ratio of specific heats to be $\gamma = C_p/C_V = 1.001$ to approximate isothermal conditions, with fiducial temperature of 15 K. With this approximation, the highest density regions reached in the simulations would have a temperature of about 20 K, so the collapse is slightly stabilised by this extra thermal pressure compared to the pure isothermal case. However, since thermal pressure is generally quite small compared to turbulent ram pressure and magnetic pressure, this effect has a very minor influence on the dynamics. We also run the astrochemical evolution at fixed temperature, i.e., 15 K, in the fiducial case. The self-gravity is solved by the fast Fourier transform technique under periodic boundary, which is the same as the boundary condition of the gas.

Higher spatial resolution grids are introduced by a Jeans length refinement criterion, i.e., so that the local Jeans length is resolved by 8 cells. However, note that only up to 5 levels of refinement are introduced, so this Jeans (Truelove et al. 1997) criterion is not satisfied at higher densities. The equivalent resolution is 2048^3 in our settings (64^3 in the base level). No sink particles are used in our simulations, so the time step becomes smaller and smaller as time evolves. In practice, this means the simulation can be run for about one free-fall time. We note that fragmentation is suppressed by the presence of magnetic fields in the PSC, together with the assumed initially centrally concentrated density profile.

2.1 Initial conditions

We start from the fiducial (S3M2) model described in Goodson et al. (2016). The model initialises a spherical, turbulent, magnetised PSC according to McKee & Tan (2003). The core has $60 M_\odot$ and its density decreases with radius with a power-law exponent $k_\rho = -1.5$, i.e., $\rho \propto r^{-1.5}$. This profile is implemented in an approximate way, described below. The mass surface density of the surrounding clump environment, which sets the bounding pressure, is set to $\Sigma_{\text{cl}} = 0.3 \text{ g cm}^{-2}$, and the relationships described in McKee & Tan (2003) give the radius of the PSC to be

$$R_c = 0.057 \Sigma_{\text{cl}}^{-1/2} \text{ pc} \rightarrow 0.104 \text{ pc} \quad (1)$$

and the number density of hydrogen nuclei at the surface to be

$$n_{\text{H},s} = 1.11 \times 10^6 \Sigma_{\text{cl}}^{3/2} \text{ cm}^{-3} \rightarrow 1.82 \times 10^5 \text{ cm}^{-3}. \quad (2)$$

The average number density in the core is $\bar{n}_{\text{H}} = 1.97 \times 10^5 \text{ cm}^{-3}$ and thus the associated free-fall time of the average density is

$$\bar{t}_{\text{ff}} = \sqrt{\frac{3\pi}{32G\rho}} \rightarrow 76 \text{ kyr}. \quad (3)$$

The model assumes the PSC is magnetized by a cylindrically symmetric B -field along the z-axis. The field strength is determined by the mass-to-flux ratio normalized with the critical value M_Φ (Mouschovias & Spitzer 1976)

$$\mu_\Phi = \frac{M}{M_\Phi} = \frac{2\pi G^{1/2} M}{\Phi} \quad (4)$$

where Φ is the magnetic flux. The value of μ_Φ is 2, making the system slightly supercritical. Thus the average B -field in the PSC is about 0.44 mG (see below).

2.1.1 Density profile

We use the same density profile described in Goodson et al. (2016) to initialize the PSC:

$$\rho(r) = \rho_0 + \frac{\rho_c - \rho_0}{1 + (r/R_f)^{k_\rho}} \left(0.5 - 0.5 \tanh \left[\frac{r - R_c}{R_s} \right] \right), \quad (5)$$

where ρ_c and ρ_0 are the defined central gas density and ambient gas density. The flattening radius, $R_f = 0.15R_c$, removes the singularity at the origin. The central density is then defined to be $\rho_c = \rho_s[1.0 + (R_c/R_f)^{k_\rho}]$, where $\rho_s = n_{\text{H},s}\mu_{\text{H}}$ is the density at the surface and μ_{H} is the mass per H.

To balance pressures at the surface of the PSC, we set a density jump $\rho_0 = 0.3\rho_s$ and give the ambient gas a higher temperature $T_0 = (10/3)T_c \rightarrow 50$ K at the core surface. The tanh function ensures a smooth transition at the surface. In our case, almost all hydrogen exists in molecular form. Therefore, the particle number density at the surface of the core is

$$n_s \approx 1.82 \times 10^5 \text{ cm}^{-3} \times \frac{\mu_{\text{H}}}{\mu} \approx 1.093 \times 10^5 \text{ cm}^{-3}, \quad (6)$$

where the mean molecular weight $\mu \approx 2.3516$ amu. Therefore, the values of ρ_0 , ρ_s and ρ_c are 1.2905×10^{-19} , 4.30153×10^{-19} , and 7.82733×10^{-18} g cm⁻³.

2.1.2 Magnetic field profile

We use the method described in Goodson et al. (2016) to set the magnetic field, which is described by:

$$B(\xi) = B_0 + \frac{B_c - B_0}{1 + (\xi/R_f)^{0.5}} \left(0.5 - 0.5 \tanh \left[\frac{\xi - R_c}{R_s} \right] \right), \quad (7)$$

where $\xi = \sqrt{x^2 + y^2}$ is the cylindrical radius. The magnetic field is assumed to decrease with cylindrical radius as a power law $B(\xi) \propto (\xi/R_c)^{-0.5}$, but also flatten in the center and taper at the surface, as was implemented for the density profile. The power law keeps the magnetic pressure $(B^2/8\pi) \propto r^{-1}$. Then, $B_0 = B_s$ represents both the ambient magnetic field and the surface magnetic field. The central magnetic field is given by $B_c = B_s[1.0 + (R_c/R_f)^{0.5}]$ and the surface field is given by

$$B_s = \frac{3G^{1/2}M_c}{2\mu_{\Phi}R_c^2}. \quad (8)$$

To obtain $\mu_{\Phi} = 2$ for the PSC, implies that $B_s = 0.22$ mG.

2.1.3 Turbulence

We initialize the turbulent velocity field according to a Burgers turbulence power spectrum $E(k) \propto k^{-2}$. The value of wavenumber k runs through all integers satisfying $1 < |kL|/(2\pi) < N/2$, where N and L are the number of cells and the length along one edge of the domain. The field is initialized as a solenoidal (divergence-free) field. The code implementation applying FFTW is utilised from the GANDALF code (Hubber et al. 2018).

The magnitude of the initial turbulent velocity field is scaled to have a one-dimensional velocity dispersion $\sigma = 0.99 \text{ km s}^{-1}$, which causes the core to be moderately supervirial in the beginning. We calculate the total momentum inside the core and shift the mean velocity to zero to ensure the core stays at the domain center. There is no driven field and the energy of the turbulence dissipates as time evolves.

We note that we use a different power spectrum from Goodson

et al. (2016), which was $E(k) \propto \text{const}$, yielding more small scale power. The case of the Burgers turbulent scaling is expected to be more valid in molecular clouds, although there are few constraints on the particular regions that may be massive PSCs. In addition, we note that Körtgen et al. (2017, 2018) also used the Burgers power spectrum for their initial conditions.

2.2 Chemical model

The chemical model used in the simulation is updated from Kong et al. (2015). The new chemical network is composed of the same 132 species and a total of 3466 reactions. For simplicity, we have not considered time-dependent depletion/desorption in our model. This also means we are able to investigate the effect of the level of heavy element depletion in a controlled manner. This approach is justified in the limit that the depletion time ($t_D \propto n_{\text{H}}^{-1}$) is relatively short compared to the dynamical or free-fall time ($t_{\text{ff}} \propto n_{\text{H}}^{-1/2}$) of the PSC so that the level of heavy element depletion has approached an approximate equilibrium prior to the collapse phase of the core.

The main differences between our network and that of Kong et al. (2015) are minor changes in the dissociative recombination rates of all the forms of H_3^+ . These values were interpolated from an external table (Pagani et al. 2009) in Kong et al. (2015). We have updated these values using the analytic expressions of Majumdar et al. (2016). After these modifications, there is a 5% difference in the obtained equilibrium value of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ resulting in a standard test case compared to that of Kong et al. (2015). Figure 1 demonstrates this result of our updated network. We use the fiducial parameters of Kong et al. (2015) with initial OPR^{H₂} = 0.1 to do the test. The final equilibrium value of our model is 0.1716, which is close to the value of 0.181 found by Kong et al. (2015).

We utilize KROME (Grassi et al. 2014) to import the chemistry solver into ENZO. KROME is responsible for calculating the local equilibrium states of species in cells. The advection between cells is handled by ENZO. By default, ENZO calls the chemistry solver at each time step. However, due to the size of the network, we speed up the calculation by only evaluating the chemical evolution at coarser time intervals. Fortunately, the deuterium fractionation reactions proceed much more slowly than the dynamical evolution of the gas. Thus, we are able to run the chemistry solver with a longer time step, i.e., after every 0.01 global initial free-fall times. The chemical model is applied everywhere in the domain, i.e., including regions outside the PSC.

In our fiducial case, we use the same parameters as in Kong et al. (2015): gas temperature $T = 15$ K, cosmic-ray ionization rate $\zeta = 2.5 \times 10^{-17} \text{ s}^{-1}$, heavy element depletion factor $f_D = 10$, ratio to Habing FUV field $G_0 = 1$, visual extinction $A_V = 30$ mag, dust-to-gas mass ratio DGR = 7.09×10^{-3} , dust particle radius $a_0 = 1.0 \times 10^{-5}$ cm and dust grain density $\rho_{\text{GRAIN}} = 3.0 \text{ g cm}^{-3}$. All elements are assumed to initially be in atomic form, except for ortho/para- H₂ and HD (see Table 2).

As C and O-bearing molecules have been measured to deplete more efficiently from the gas phase compared to N-bearing ones (e.g., Caselli et al. 2002; Bergin et al. 2002), we use $f_D^{\text{C,O}}$ to represent the depletion factor of C and O separately from the other heavy elements, i.e., N, represented by f_D^N . We use various values of gas temperature, CO depletion factor, initial OPR^{H₂} and cosmic-ray ionization rate in our simulations. Table 1 shows the different choices used. The abundances of ortho- and para- H₂ are controlled by OPR^{H₂} and the abundances of atomic heavy elements are decided by the depletion factors ($f_D^{\text{C,O}}$ and f_D^N). However, we

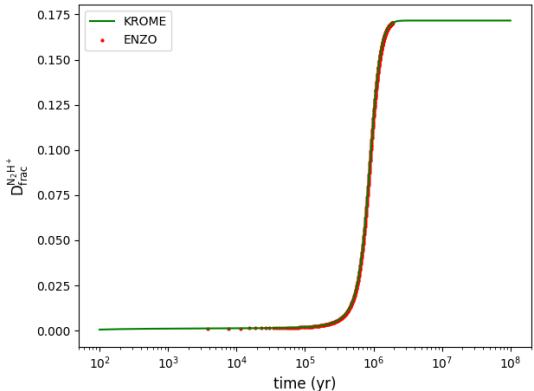


Figure 1. Test case of evolution of $D_{\text{frac}}^{N_2H^+}$ with initial OPR $H_2 = 0.1$. Green line shows the evolution in a single grid run by KROME. Red points shows the result run by ENZO+KROME in a 4^3 cube with static uniform gas. The chemical abundance evolution is consistent in the two cases.

also investigate two models in which the core is chemically aged by running the network at a constant density of $n_H = 1.97 \times 10^5 \text{ cm}^{-3}$ for 3×10^5 and 6×10^5 yr.

3 RESULTS

We evolve each model listed in Table 1 for 80% of the initial mean free-fall time of 76 kyr, i.e., for 61 kyr. At the last time step, the number density n_H reaches $> 10^{11} \text{ cm}^{-3}$. Simulations are not run for longer times, given that we do not include any sub-grid model for the formation of a protostar and any feedback that it would generate. Since all simulations start from identical physical initial conditions (except for the 10 K case), they should have the same density, temperature, velocity distribution, etc, at the end. However, the differences in investigated model parameters cause them to have different chemical compositions. First we focus on comparing the chemical compositions and related derived properties, especially $D_{\text{frac}}^{N_2H^+}$. Then we examine kinematic and dynamic properties of the core, especially as traced by N_2D^+ .

3.1 Fiducial model

Figures 2 and 3 show the results of the fiducial simulation, with views of the core projected along the x -axis and z -axis, respectively. In these figures, the columns from left to right show outputs at 0, 0.2, 0.4, 0.6 and 0.8 t_{ff} . The rows from top to bottom are the mass surface density, radial velocity weighted by N_2D^+ abundance, column density of N_2H^+ , column density of N_2D^+ and deuterium fraction of N_2H^+ . To compare with observational data, we follow the criteria in Goodson et al. (2016), setting a minimum density threshold $n_H = 8 \times 10^5 \text{ cm}^{-3}$, corresponding to roughly 10% of the critical density of the (3-2) transitions (Miettinen & Offner 2013), when we calculate the column density of N_2H^+ and N_2D^+ . The effect of whether or not to impose this threshold is examined later.

We see from Figures 2 and 3 that the core collapses mostly along the large scale B -field direction, i.e., the z -axis, to form a flattened and eventually filamentary structure. When viewed along the x -axis (Figure 2), the maximum mass surface density grows from around 0.18 to 4.1 g cm^{-2} and the magnetic field remains roughly perpendicular to the filament. The average line-of-sight velocities show large-scale structures, dominated at first by the large scale

modes of the turbulent initial conditions. Note that at $t = 0$ in Figs. 2 and 3, we have weighted the velocities simply by density because no N_2D^+ has yet formed, while at later times they are weighted by N_2D^+ abundance. At the end of the simulation, the line-of-sight velocity field shows a high dispersion near the core center, with values ranging from -2.0 to $+3.6 \text{ km/s}$. The overall velocity dispersion is discussed in more detail in §3.4 in relation to the dynamical state of the core.

The column densities of N_2H^+ and N_2D^+ grow as time evolves. The peak value of $N_{N_2H^+}$ increases to more than 10^{14} cm^{-2} , while that of N_2D^+ reaches about 10^{11} cm^{-2} . However, note that these values apply to very localised regions within the core, with average values being much lower. The value of $D_{\text{frac}}^{N_2H^+}$ also increases with time, reaching a peak value of about 2×10^{-3} .

In Figure 3, the core is viewed projected along the z -axis, which is the orientation direction of the initial large-scale, cylindrically symmetric magnetic field. Comparing with Figure 2, the spatial structure of the core is now somewhat more dispersed, although the mass surface density reaches similar peak values. The magnetic field orientations that arise in this plane are more random, since they lack the component of the original large-scale field and arise solely from the motions due to initial turbulence and gravitational collapse that drag the B -field with them. The line-of-sight velocities also reach similar local dispersions around the core center as were seen in the x -axis projection. The column densities of N_2H^+ and N_2D^+ and their ratio, $D_{\text{frac}}^{N_2H^+}$, also reach similar values as seen in the previous view. We conclude that the large-scale magnetic field does not significantly change the patterns of most of these results. In particular, the projections along the x -axis are sufficient for us to analyze the astrochemical results.

In our fiducial model, we see that the growth of $D_{\text{frac}}^{N_2H^+}$ is quite slow so that after 80% of an initial free-fall time the peak values are only ~ 0.002 , which is smaller than seen in many observed PSCs (see §1). In contrast, Goodson et al. (2016) reported a peak value of $D_{\text{frac}}^{N_2H^+}$ about 50 times higher than ours at the equivalent time. In Figure 4, we plot the deuterium fraction versus column density of all the cells in our simulations and compare to the results of Goodson et al. (2016). The left panel shows the result of the fiducial model, which indicates that the differences arise mostly from the high column density regions. Our derived values of $D_{\text{frac}}^{N_2H^+}$ reach a near constant maximum level and do not rise with increasing column density, which is in contrast to the results of Goodson et al. (2016). We discuss the reasons for this different behaviour in Appendix A, where we explain it as being due to deficiencies in the approximate method of Goodson et al. (2016).

In summary, our fiducial physical and chemical model cannot achieve the high values of $D_{\text{frac}}^{N_2H^+} \gtrsim 0.1$ that are observed in many PSCs. One possible explanation is that the physical model is wrong: i.e., the collapse is occurring too quickly and should rather take place over several or many free-fall times (e.g., Kong et al. 2015, 2016). Another possible explanation is that the chemical model is incorrect and that different chemical conditions can allow higher values of $D_{\text{frac}}^{N_2H^+}$ to be achieved quickly within about one free-fall time. This second possibility is investigated next in §3.2.

3.2 Effect of chemical conditions on abundances of N_2H^+ and N_2D^+

High values of $D_{\text{frac}}^{N_2H^+}$, up to ~ 0.5 , have been observed in several star-forming regions that are thought to harbour massive PSCs (e.g.,

model	OPR^{H_2}	T	ζ	f_D^N	$f_D^{\text{C},\text{O}}$	$D_{\text{frac}}^{\text{N}_2\text{H}^+}$
Fiducial	0.1	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	10	-
OPR0.01	0.01	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	10	-
T10	0.1	10K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	10	-
CR10x	0.1	15K	$2.5 \times 10^{-16} \text{ s}^{-1}$	10	10	-
fDCO100	0.1	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	100	-
fDCO1000	0.1	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	1000	-
Aged3e5	1.729×10^{-2}	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	10	7.58×10^{-3}
Aged6e5	2.881×10^{-3}	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	10	5.38×10^{-2}
OPR0.01_fDCO1000	0.01	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	1000	-
OPR0.001_fDCO1000	0.001	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	10	1000	-
fDN100_fDCO1000	0.1	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	100	1000	-
fDN1000_fDCO1000	0.1	15K	$2.5 \times 10^{-17} \text{ s}^{-1}$	1000	1000	-
CR4x_fDCO1000	0.1	15K	$1.0 \times 10^{-16} \text{ s}^{-1}$	10	1000	-
CR4x_fDN100_fDCO1000	0.1	15K	$1.0 \times 10^{-16} \text{ s}^{-1}$	100	1000	-

Table 1. List of PSC simulations and their parameters. We first try modifying one parameter at a time compared to the fiducial model to understand its influence, and then test the combination of different parameters. The parameters that are varied are initial OPR^{H_2} , temperature (T), cosmic-ray ionization rate (ζ), depletion factor of C and O ($f_D^{\text{C},\text{O}}$), and depletion factor of N (f_D^N). The "Aged" models use the same initial parameters as the fiducial model, but the chemistry has been evolved first at constant density for 3×10^5 and 6×10^5 years before the core starts to collapse. We note the values of OPR^{H_2} and $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ at the beginning of hydrodynamic simulations in the table.

Species	Abundance ($n_{\text{species}}/n_{\text{H}}$)
p-H ₂	$5/11 = 0.455$
o-H ₂	$5/110 = 0.0455$
HD	1.5×10^{-5}
He	1.0×10^{-1}
N	2.1×10^{-6}
O	1.8×10^{-5}
C	7.3×10^{-6}
GRAINO	1.32×10^{-12}

Table 2. The initial abundances of the fiducial model. The depletion factor = 10 reduces the abundances of C, O and N by a factor of 10 from the default Solar metallicity ISM values.

Miettinen et al. 2012; Kong et al. 2016). It is clear that our fiducial model cannot reach such high values in one free-fall time. In the following subsections, we discuss some possible ways to reach higher values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ in our simulated PSC. In particular, we investigate the effect of several model parameters, as summarised in Table 1. Our overall goal is to see under what conditions very high values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$, similar to those of some observed PSCs, can be achieved in the context of the presented physical model. If such high values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ are not achievable, then this would strengthen the case that a different physical model is needed, e.g., one involving less rapid collapse of the PSC via increased magnetic field support, as has been discussed by, e.g., Kong et al. (2016).

Maps of N_2H^+ and N_2D^+ column densities and their ratio, $D_{\text{frac}}^{\text{N}_2\text{H}^+}$, projected along the x -axis, of the various investigated models at $0.8 t_{\text{ff}}$ are plotted in Figures 5 and 6. We show versions of the maps (first, third and fifth rows) in which only contributions from cells with number densities $> 8 \times 10^5 \text{ cm}^{-3}$, corresponding to about 10% of the critical density of the (3-2) rotational transitions, are counted. However, to investigate the effect of this choice, we also show versions of the maps (second, fourth and sixth rows) without use of this threshold.

3.2.1 Initial OPR^{H_2}

The OPR^{H_2} is expected to decline with time in molecular clouds (Flower et al. 2006a). The higher energy of ortho-H₂ suppresses formation of H_2D^+ and thus N_2D^+ and so limits the growth of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ (e.g., Pagani et al. 1992; Gerlich et al. 2002; Walmsley et al. 2004b; Sipilä et al. 2010b). To investigate the effect of initial OPR^{H_2} compared to the fiducial model, we use a value that is ten times smaller, i.e., $\text{OPR}^{\text{H}_2} = 0.01$, in the OPR0.01 model. In Figure 5, we see this change causes the deuterium fraction to increase by about one order of magnitude at all locations, with little dependence on column density. Thus reducing the initial OPR^{H_2} level is a way to boost $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ in the core. However, we note that in our models this does not lead to a strong spatial concentration of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ associated with high column density regions. Such a concentration would likely require a similar spatial concentration of OPR^{H_2} in the core region. We also note that achieving a low value of OPR^{H_2} would require "chemical aging" of the core over a timescale that is expected to be longer than the initial free-fall time of our model (e.g., Kong et al. 2015), and that this should be considered together with the evolution of other chemical species. Such models are presented below in §3.2.5.

3.2.2 Temperature

Kong et al. (2015) showed that for temperatures below $\sim 20 \text{ K}$, there is little variation of the growth timescales and equilibrium levels of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$. However, to check this and to also examine any effects on the dynamical evolution, e.g., affecting fragmentation, we run another isothermal PSC at 10 K in the T10 model. In Figure 5, we see that the $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ map of the T10 model shows a slightly wider distribution than the fiducial model. However, as expected, $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ is at a similar level as the fiducial model. Furthermore, the physical structures that form during the collapse are broadly similar, indicating that thermal pressure support plays a relatively minor role compared to magnetic fields and turbulence.

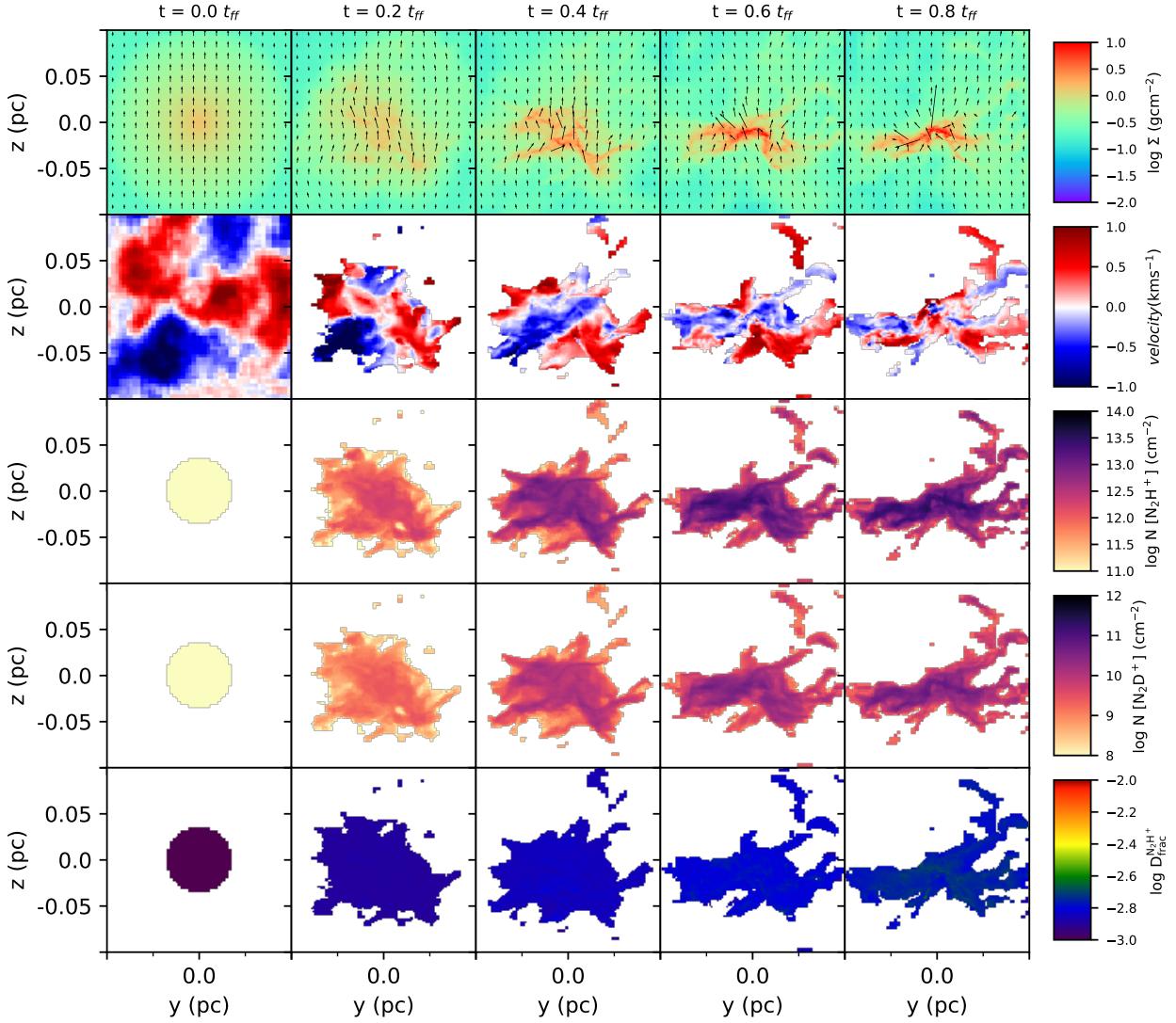


Figure 2. Time evolution of the fiducial model. The columns from left to right show the snapshots at times = 0, 0.2, 0.4, 0.6 and 0.8 of the initial core mean free-fall time of 76 kyr. Each row shows properties projected along the x -axis. From top to bottom these are: mass surface density; line-of-sight velocity weighted by N_2D^+ (except at $t = 0$, when it is weighted by Σ) (see text); column density of N_2H^+ ; column density of N_2D^+ ; and the deuterium fraction, $D_{\text{frac}}^{N_2H^+}$. The lines in the first row indicates the average (mass-weighted) projected magnetic field orientations.

3.2.3 Cosmic-ray ionization rate

The cosmic-ray ionization rate, ζ , influences $D_{\text{frac}}^{N_2H^+}$ significantly in two aspects: the growth timescale and the final equilibrium value. The timescale to reach the equilibrium state becomes ~ 100 times shorter and the equilibrium value of $D_{\text{frac}}^{N_2H^+}$ ~ 6 times smaller as ζ increases from 10^{-18} to 10^{-15} s^{-1} (Kong et al. 2015). In the CR10x model, we try a cosmic-ray ionization rate of $2.5 \times 10^{-16} \text{ s}^{-1}$, i.e., $10\times$ higher than that of the fiducial model and in line with the value measured in diffuse clouds (e.g., Indriolo & McCall 2012).

The result shows that $D_{\text{frac}}^{N_2H^+}$ not only increases globally (since we are in the limit where the equilibrium value has not been reached), but has higher values along the dense filaments. In other words, it shows a more concentrated spatial structure. We can see the same phenomenon by comparing with the map plotted without the density threshold. To verify this, we plot $D_{\text{frac}}^{N_2H^+}$ versus column density in the middle panel of Figure 4. We see that $D_{\text{frac}}^{N_2H^+}$ is increasing rapidly until $N_H \sim 3 \times 10^{23} \text{ cm}^{-2}$, after which it stays quite constant. Thus we see that higher cosmic-ray ionization rates enhance the influence

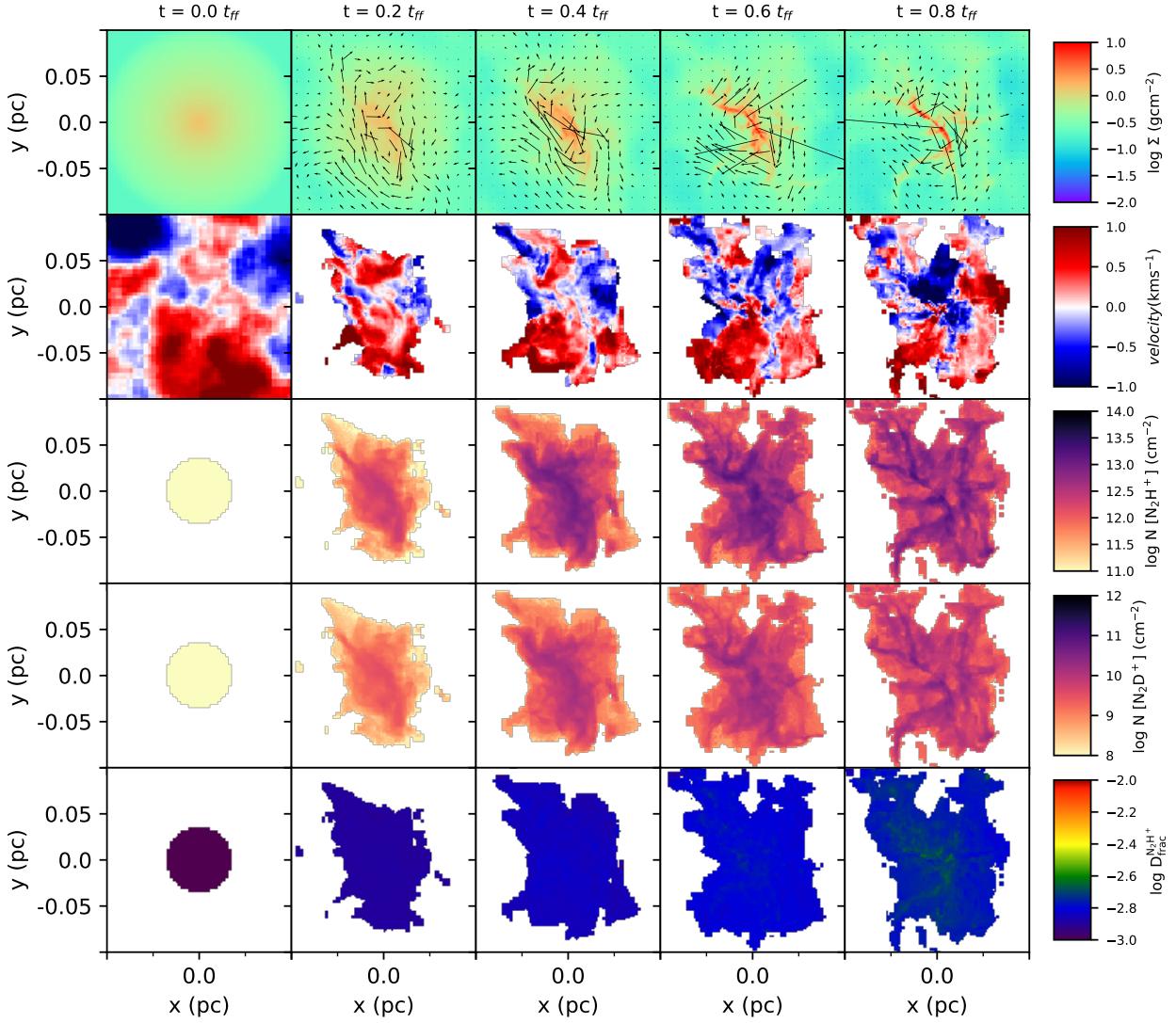


Figure 3. The same as Figure 2, but now with the panels showing quantities projected along the z -axis.

of density and cause $D_{\text{frac}}^{N_2H^+}$ to be a better tracer of high-density regions.

As a caveat, we note that in actual molecular clouds, where cosmic rays are mostly propagating in from the surroundings, we expect higher density regions to have lower cosmic-ray ionization rates, due to shielding by surrounding material and magnetic fields. Still, Padovani et al. (2018, see their figure F.1), predict values close to 10^{-16} s^{-1} for H_2 column densities of about 10^{23} cm^{-2} , including the effects of attenuation. Further independent constraints on the levels of the cosmic-ray ionization rate are needed when making detailed comparisons of our models with observed sources.

3.2.4 CO depletion

As CO destroys H_3^+ and its deuterated forms, a higher level of CO depletion is beneficial to deuterium fractionation (Dalgarno & Lepp 1987; Crapsi et al. 2005). The level of CO depletion itself also depends on density, temperature, cosmic-ray ionization rate, etc., and will evolve in time towards an approximate equilibrium level. However, for simplicity, we do not include freeze-out reactions and ice phase species in our network. Instead, we model CO depletion as a fixed input parameter, following the treatment of Kong et al. (2015). The level of depletion is parameterized by the CO depletion factor, f_D^{CO} , which is the ratio of expected CO abundance if all C is in the form of gas-phase CO and the actual gas phase CO abundance. This method is expected to be approximately valid in cold ($T \lesssim$

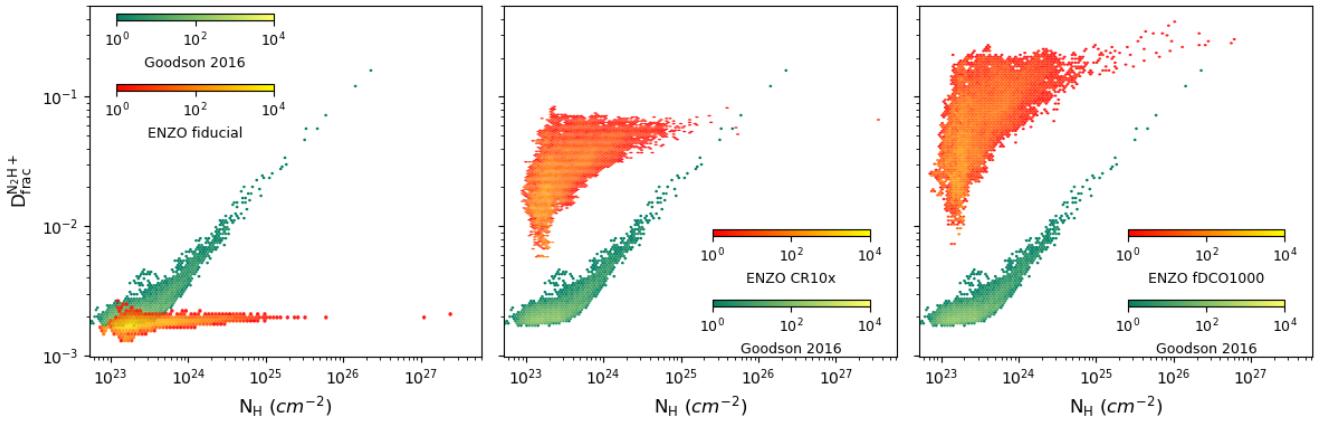


Figure 4. Scatter plot of total H column density versus $D_{\text{frac}}^{\text{N}_2\text{H}^+}$. In each panel, the green-yellow colour-map shows the fiducial model of Goodson et al. (2016) at the end of the evolution, i.e., at $0.8 t_{\text{ff}}$. From left to right, we plot three of our models in red-yellow colour-map: fiducial, CR10x and fDCO1000 (see text and Table 1).

20 K), dense ($n_{\text{H}} \gtrsim 10^4 \text{ cm}^{-3}$) protocluster clump environments in which the freeze-out time is short compared to the free-fall time (Walmsley 1991). Note that in our physical model the surrounding clump gas has a density of $n_{\text{H,cl}} = 6 \times 10^4 \text{ cm}^{-3}$.

With reference to the fiducial model, which has $f_D^{\text{C,O}} = 10$, we explore the influence of two higher values of $f_D^{\text{C,O}} = 100$ and 1000 in the fDCO100 and fDCO1000 models, respectively. In Figure 6, both models show more concentrated spatial structures. In the densest regions, the 10 times higher depletion factor in the fDCO100 model causes $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ to become about 10 times higher than in the fiducial case. Similarly, the 100 times higher depletion factor in the fDCO1000 model causes $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ to increase by about two orders of magnitude, i.e., reaching values >0.1 in less than one free-fall time.

The relationship between $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ and column density of the fDCO1000 model is also plotted in Figure 4 (right panel). It shows the fDCO1000 model has an overall higher level of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ than the other models so far considered.

3.2.5 Initial chemical age

We also investigate starting the simulation from an older chemical age. To obtain chemical abundances at a specific chemical age, we first calculate the mean density of the PSC to be $\bar{n}_{\text{H}} = 1.97 \times 10^5 \text{ cm}^{-3}$. We use this mean density and other fiducial model parameters of T and ζ to evolve a single grid chemical model to obtain abundances of species at a given chemical age. The effect is similar to using a smaller initial OPR^{H_2} , but now the chemical composition is no longer starting from pure elemental abundances (of the heavy elements).

We investigate two different chemical ages: 3×10^5 years and 6×10^5 years. Figure 7 shows the time evolution of key species in this PSC prior evolutionary phase. We see that by 3×10^5 years, equilibrium levels of CO, N₂ and N₂H⁺ have been achieved, while N₂D⁺ is still growing. This is because, at this time, OPR_{H₂} is undergoing a gradual decline. The values of OPR_{H₂} at $(3 \text{ and } 6) \times 10^5$ yr are $(1.73 \text{ and } 0.29) \times 10^{-2}$, while the values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ are $(0.76 \text{ and } 5.4) \times 10^{-2}$, respectively (see Table 1).

In Figure 6, we see that the models Aged3e5 and Aged6e5 that have this described prior chemical evolution are able to reach higher

values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$. Especially model Aged6e5 can reach $D_{\text{frac}}^{\text{N}_2\text{H}^+} \sim 0.1$, similar to the levels achieved in model fDCO1000. However, now the effect is more global, i.e., there is not as strong a spatial concentration of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ following the column density structure. We also see that the similar levels of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ that arise in these models are achieved with quite different absolute abundances of N₂H⁺ and N₂D⁺. We will return to these metrics as a way to distinguish between models in §3.3.

3.2.6 Combination of initial OPR_{H₂} and CO Depletion

In §3.2.1, we saw that lower initial OPR_{H₂} helps the growth of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$, while its spatial structure is not enhanced with density. In contrast, the models of §3.2.4 with a higher CO depletion factor have a clearer structure of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ following column density. Therefore, we try a series of combinations of initial OPR_{H₂} and CO depletion to survey whether $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ can have a stronger correlation with column density. We consider two possible CO depletion factors: 10 and 1000, and four possible values of initial OPR_{H₂}: 3, 0.1, 0.01 and 0.001. From the results of these eight models, initial OPR_{H₂} does not appear to enhance the spatial structure of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$.

Although the initial OPR_{H₂} does not further enhance the spatial structure, the models with high CO depletion factor and low initial OPR_{H₂} still show the possibility to obtain high values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$. We thus focus on two models, OPR0.01_fDCO1000 and OPR0.001_fDCO1000, which are listed in Table 1 and which will be considered further in the comparison to observed systems (see below).

3.2.7 Combination of N and CO depletion factors

In §3.2.4, we found that increased CO depletion factor enhances the spatial concentration of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$. In reality, N depletion should also happen where CO depletion occurs, although perhaps at a more modest level (Flower et al. 2006b). The level of N depletion will naturally affect the absolute abundances of N₂H⁺ and N₂D⁺.

To investigate the influence of N depletion, we run simulations using two higher N depletion factors ($f_D^{\text{N}} = 100, 1000$), and with CO depletion factor set to 1000. The results are plotted in the left

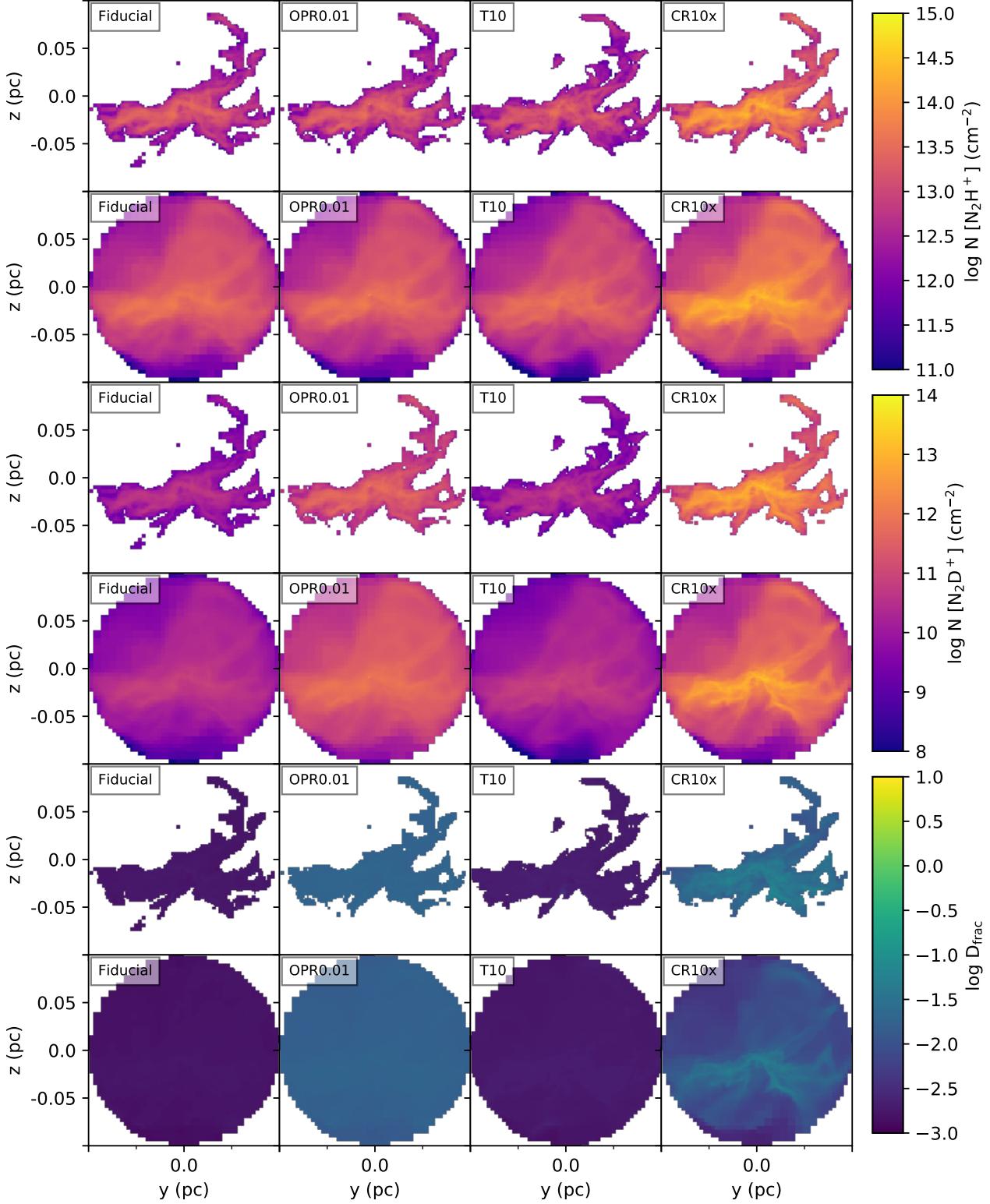


Figure 5. The projections of N_2H^+ , N_2D^+ , and $D_{\text{frac}}^{N_2H^+}$ along the x -axis at $0.8 t_{\text{ff}}$. These quantities are plotted in two cases (with density threshold and without density threshold) to display the general concentration properties (see text). The ambient gas is subtracted by limiting the volume inside 0.1 pc sphere. Each column represents a different model (left to right): fiducial; OPR0.01; T10; CR10x, with parameters of these models listed in Table 1. The remaining models are plotted in Figure 6 and 8

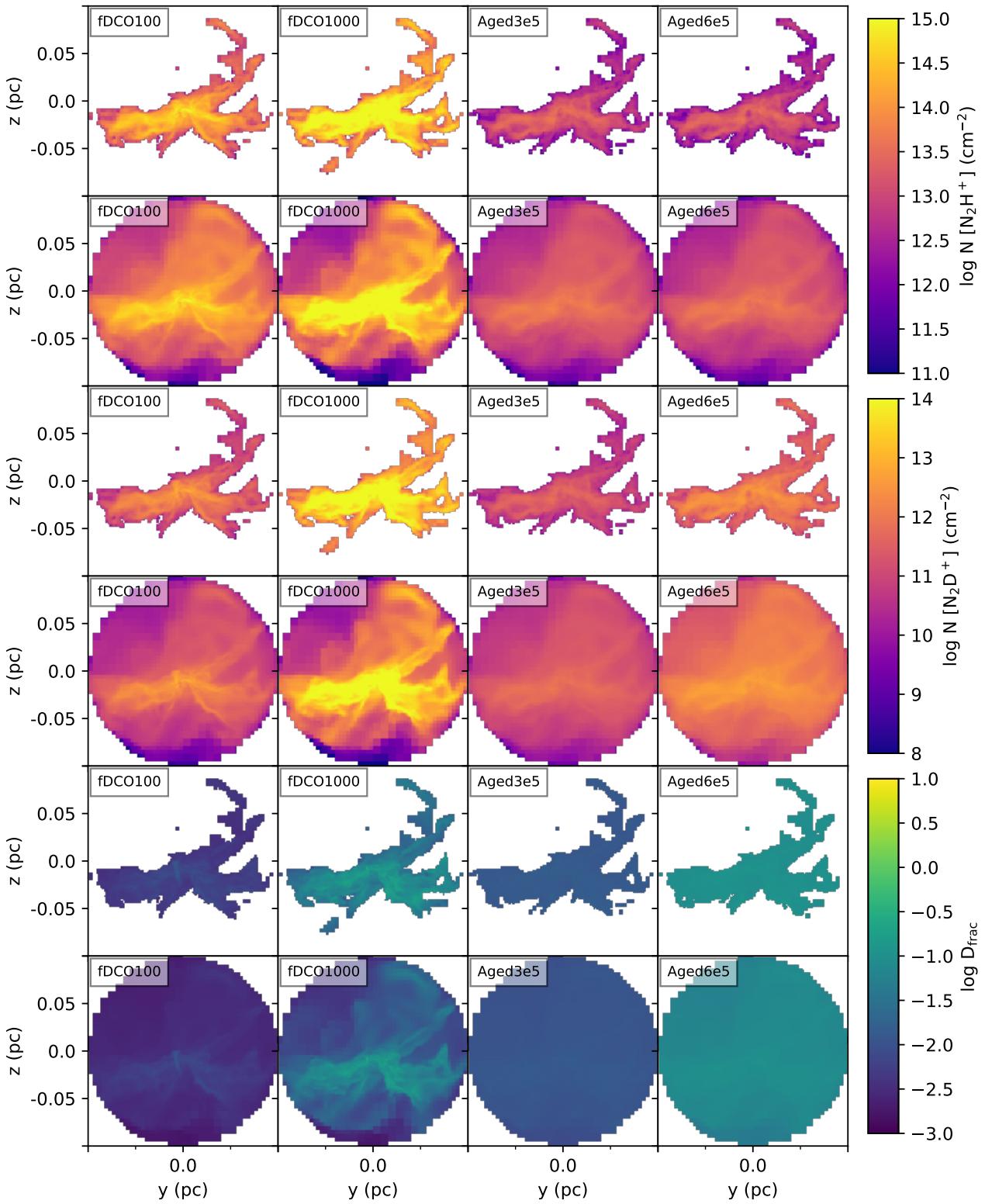


Figure 6. Same as Figure 5, but now for models (left to right): fDCO100; fDCO1000; Aged3e5; Aged6e5, with their parameters listed in Table 1.

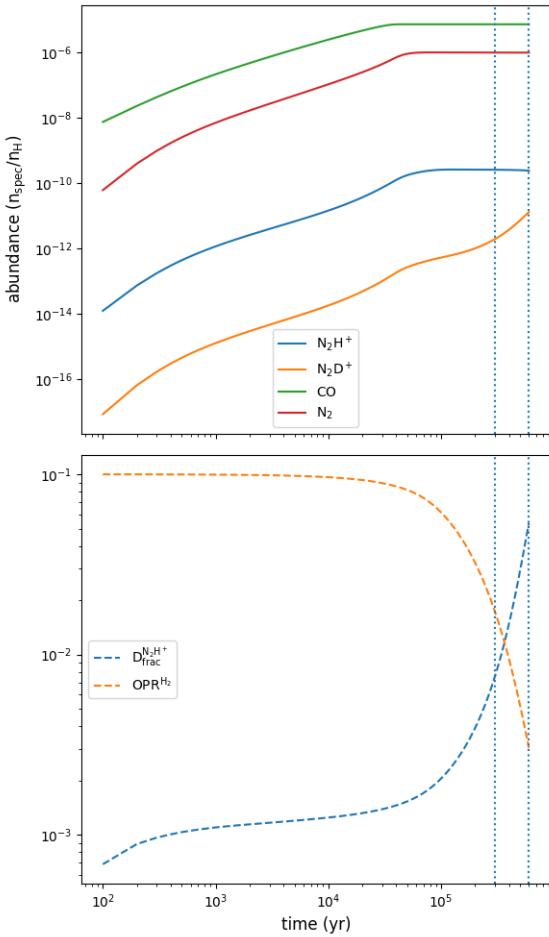


Figure 7. The time evolution of key species (CO, N₂, N₂H⁺, N₂D⁺) (top panel) and OPR^{H₂} and D_{frac}^{N₂H⁺ (bottom panel) in the prior evolutionary phase of the PSC. Vertical lines indicate times at 3 × 10⁵ years and 6 × 10⁵ years.}

two columns of Figure 8. It is easy to see that the column densities of N₂H⁺ and N₂D⁺ are reduced and the spatial structures of D_{frac}^{N₂H⁺ are enhanced (note that in this figure the range of D_{frac}^{N₂H⁺ now extends up to values of 10).}}

3.2.8 Combining higher cosmic-ray ionization rates with other models

In §3.2.3, it was shown that the cosmic-ray ionization rate has a strong impact to the level of D_{frac}^{N₂H⁺ and its spatial structure. However, we have not yet investigated its influence on the models with higher depletion factors. Thus combinations of a higher cosmic-ray ionization rate of $\zeta = 1.0 \times 10^{-16}$ with the fDCO1000 and fDN100_fDCO1000 models are carried out. The results are also plotted in Figure 8. These models do not show a strong effect of further enhancement of the spatial structure of D_{frac}^{N₂H⁺. However, due to the effect of shortening chemical time scales, the enhanced cosmic ray models have generally enhanced levels of D_{frac}^{N₂H⁺ by about a factor of 10 than in the lower cosmic ray cases. Final values of D_{frac}^{N₂H⁺ that are > 1 can be achieved in some regions.}}}}

It is worth noting that high cosmic-ray ionization rates may lead to high rates of desorption of species from grain surfaces such

that the highest values we have used for gas phase depletion factors may not be reasonable in this regime. To investigate this, we have run some test models with the astrochemical network of [Walsh et al. \(2015\)](#), which includes the influences of thermal desorption, photodesorption and grain surface reactions. Cosmic ray especially contributes to the UV field produced by exciting H₂ molecules. These results show that in gas with n_H = 10⁶ cm⁻³ at T = 15 K and cosmic ray ionization rate of $\zeta = 1.0 \times 10^{-16}$ the CO depletion factor reaches ~ 500 in 100,000 years. However, there are significant uncertainties in the modeling of cosmic-ray induced desorption processes, so we consider our most extreme case with a CO depletion factor of 1000 under such conditions to be a reasonable choice, especially since much of the core material of interest soon achieves densities several times higher than n_H = 10⁶ cm⁻³.

3.3 Average Core Properties and Comparison to Observed Cores

3.3.1 Average core abundances and D_{frac}^{N₂H⁺}

We have seen from the results presented so far that there are different combinations of astrochemical conditions that can help achieve a given value of D_{frac}^{N₂H⁺ within a single free-fall collapse time of a massive PSC. To break these degeneracies when comparing to observed systems, we will need to use the full information set available, i.e., measurements of the absolute abundances of N₂H⁺ and N₂D⁺, as well as their ratio that defines D_{frac}^{N₂H⁺. Additional constraints on CO depletion factor and cosmic-ray ionization rate are also useful, if available.}}

We calculate the average column densities of N₂H⁺ and N₂D⁺ in our simulations. Assuming the N₂D⁺(3-2) emission area can be well resolved, we calculate the average over this resolved area of the map that contains N₂H⁺ and N₂D⁺ emission (recall that these are sight lines that contain at least one cell above the adopted density threshold n_{H,crit} = 8 × 10⁵ cm⁻³). The average column densities of N₂H⁺ and N₂D⁺ in the core are shown in Figure 9a, with the lines tracking the evolution of the simulations in time, i.e., rising up from the lower left to the upper right. The behaviour of N₂H⁺ is that it tends to grow to reach a maximum value and then declines by a small amount. On the other hand, N₂D⁺ tends to keep increasing during the evolutionary phases that we follow. The largest factor that leads to variation among the models is the adopted N depletion factor.

Several observed cores and sub-regions of cores are also shown in Figure 9a. These exhibit values of column densities and D_{frac}^{N₂H⁺ that are similar to those of some of the simulations. We note that many of the observed cores are of relatively low-mass (e.g., those of Cheng et al., in prep. and many of the sources of [Miettinen et al. \(2012\)](#)) and low total H column, so that one does not necessarily expect agreement in terms of absolute columns of N₂H⁺ and N₂D⁺ with the simulations, which are for a relatively high density, 60 M_⊙ core. Thus, we shall also focus on abundances of the species relative to H.}

To estimate the H column density in the simulations to be used in the abundance estimates, which are shown in Figure 9b, we only count material from cells above the n_{H,crit} density threshold, with an additional constraint that the material be located within 0.1 pc distance of the core centre, i.e., excluding surrounding clump environment material. We note that when such measurements are done observationally, then, to be similar to this theoretical method, the clump background contribution should be subtracted off the esti-

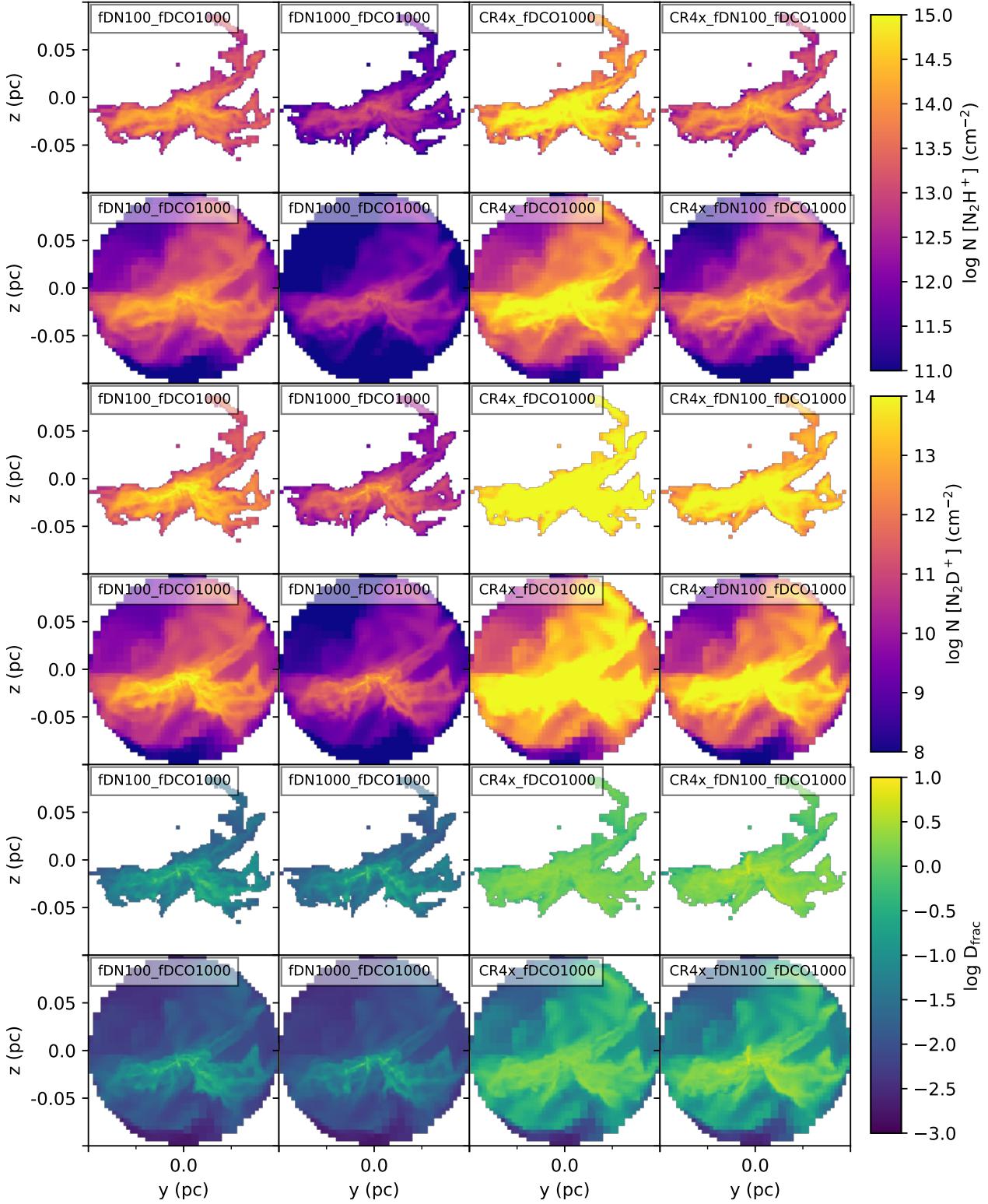


Figure 8. Same as Figure 5, but now for models (left to right): fDN100_fDCO1000; fDN1000_fDCO1000; CR4x_fDCO1000; CR4x_fDN100_fDCO1000, with their parameters listed in Table 1. Due to the high levels of N_2D^+ and $D_{\text{frac}}^{N_2H^+}$ in these models, different scale bars for N_2D^+ and $D_{\text{frac}}^{N_2H^+}$ are used in this figure.

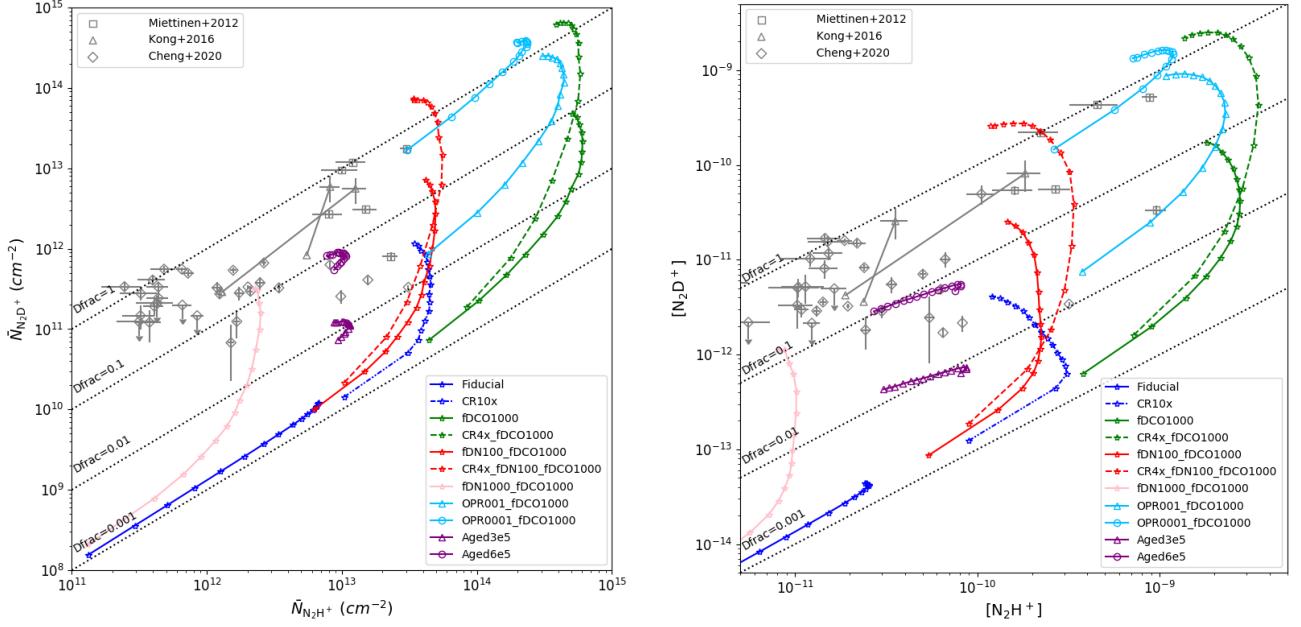


Figure 9. (a) Left: Time evolution (from 0.05 t_{ff} to 0.8 t_{ff}) of the average column densities of N_2H^+ and N_2D^+ for various simulations as labelled. The grey squares, triangles and diamonds are observational data from Miettinen et al. (2012); Kong et al. (2016) and Cheng et al. (in prep), respectively. (b) Right: As (a), but now showing average abundances of N_2H^+ and N_2D^+ with respect to H nuclei.

mate of the core region. As expected, the simulations show qualitatively similar behaviour in their abundance evolution as shown in their absolute column density evolution.

We note that if a simpler circular aperture with a 0.1 pc radius is used in the analysis, then the estimates of absolute column densities of N_2H^+ and N_2D^+ generally decrease, typically by factors of 2 to 3, since significant area from non-emitting regions is included. However, the estimates of the abundances with respect to H change by smaller amounts. The density threshold is another factor influencing the average column densities. If we remove the density threshold but still limit the valid domain inside the 0.1 pc radius sphere, the column densities of N_2H^+ can grow by factors of 1.3 to 3.2 (time-averaged in the various models). Similarly, the column densities of N_2D^+ grow by factors of 1.1 to 3.2, but always by a slightly smaller factor than N_2H^+ . For the column density of H, we find that N_{H} increases by a factor of 1.74 when the density threshold condition is relaxed.

Overall, considering the comparison of simulations with observed cores, we see that the observed cores exhibit a range of values of $\bar{N}_{\text{N}_2\text{H}^+}$ from about 10^{11} to a few $\times 10^{13} \text{ cm}^{-2}$ and of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ from a few $\times 0.01$ to about 1. It is to be remarked that there can be significant uncertainty in the estimates of the observed column densities, e.g., due to excitation temperature uncertainties (e.g., see the case of the C1-S and N cores observed by Kong et al. 2016) or the cores of (Miettinen et al. 2012), where a mixture of LTE and non-LTE assumptions were employed. Estimates of the abundances of the species bring in additional uncertainties associated with the measurement of N_{H} . The observed values of $[\text{N}_2\text{H}^+]$ range from about 10^{-11} to 10^{-9} , while those of $[\text{N}_2\text{D}^+]$ range from about 10^{-12} to close to 10^{-9} . Focusing on the abundance plots, we see that the range of observed core properties is well covered by the simulations. For example, the model CR4x_fDN100_fDCO1000 reaches the high values of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ that are observed in some cores with typical abundances of $[\text{N}_2\text{H}^+] \sim \text{few} \times 10^{-10}$. We see that variation in

f_D^N , i.e., the parameter controlling the degree of N depletion from the gas phase, is one way that the simulations can be tuned to match a given observed system. Given the uncertainties in predicting the degree of N depletion from first principles, one strategy when modeling a population of cores in the same region (assumed to have similar environmental conditions) is to set this empirically based on observations of N_2H^+ and then use any observed differences in N_2D^+ abundance as a measure of evolutionary stage, with absolute age differences estimated from such simulations.

3.3.2 Spatial structure of $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$

We have seen that different models lead to different degrees of spatial concentration of $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$ that follow the PSC structure, i.e., the level of deuteration can increase rapidly with density so that the PSC appears as an "island" of high $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$ within the surrounding clump environment. While there are to date relatively few observed maps of $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$, especially on the scale of PSCs, in a number of observational studies, the morphology of N_2D^+ emission appears more concentrated than that of N_2H^+ (e.g., Kong et al. 2016; Barnes et al. 2016). Among the models that we have considered so far, those having high CO depletion factors or high cosmic-ray ionization rates show significant degrees of spatial concentration. In the last sub-section, we compared core averaged column densities and abundances with observational data and found that the CR4x_fDN100_fDCO1000 model is one of the best of those explored so far in matching the observed systems.

In Figure 10 we examine the dependence of N_2H^+ and N_2D^+ column densities and $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$ on total H column density, N_{H} , in the CR4x_fDN100_fDCO1000 model at $t = 0.8t_{\text{ff}}$ for two cases of with and without the density threshold being needed for N_2H^+ and N_2D^+ emission. We see that there is a general rise in these quantities with N_{H} . The use of a density threshold in the simulations for N_2H^+

and N_2D^+ emission causes a truncation at low values of N_{H} , as expected.

Figure 10 also shows observational data from various cores, i.e., the same data shown in Figure 9. These data include core-averaged properties, as well as a small strip map from Cheng et al. (in prep.), which probes a range of conditions in a linear feature in the Vela C GMC that spans from two early-stage protostellar cores and a bridge feature containing a PSC candidate in between. These systems are not as massive as our simulated core and are also generally of lower column density, as can be seen in Figure 10. Nevertheless, they provide a number of probes of independent regions that span conditions relevant to the outer boundary of our simulated massive PSC. The comparisons that are shown in Figure 10 indicate a preference for not using the density threshold condition for N_2H^+ and N_2D^+ emission. Still, a self-consistent test of the model will require new observations to make resolved maps of N_2H^+ and N_2D^+ of a massive PSC.

3.3.3 Chemical history of the core defined by N_2D^+

In §3.3.1 and §3.3.2 we have considered the time evolution of average column densities, abundances and the spatial structures of different simulated cores. We then selected the CR4x_fDN100_fDCO1000 model as one of the best cases for matching some highly deuterated observed systems. We now focus on the time evolution of the average properties of this core.

In a number of observational studies (e.g., Tan et al. 2013; Kong et al. 2018), the PSC boundary is defined by the region (either in position-position space or in position-position-velocity space) that is detected in N_2D^+ (3-2) line emission. In our simulations, following Goodson et al. (2016), we have assumed N_2D^+ emission arises only if the H nuclei number density of a cell is larger than $8 \times 10^5 \text{ cm}^{-3}$. We thus measure the effective area of the core, $A_{\text{c,eff}}$, in a given image as the area of all the pixels that contain at least one cell along the line of sight that meets this density threshold. Then, we define the effective radius of the core as $R_{\text{c,eff}} \equiv (A_{\text{c,eff}}/\pi)^{1/2}$. The time evolution of the effective radii as viewed from different directions are plotted in Figure 11a. In the case of the x and y projections, we see that the core radius grows quickly from an initial value of 0.037 pc to about 0.055 pc, being fairly constant (or even declining) after $t = 0.2t_{\text{ff}}$. In the case of the z projection, $R_{\text{c,eff}}$ grows from 0.037 pc to about 0.08 pc, which is reached by $t = 0.6t_{\text{ff}}$, after which it remains nearly constant. The larger effective area when viewed in the z -projection (i.e., viewing the $x - y$ plane) is clear from Figures 2 and 3.

Figures 11b and c show the time evolution of the average H nuclei number density, \bar{n}_{H} , and the average total H column density, \bar{N}_{H} , for the core defined by its N_2D^+ emitting region. Two methods for estimating densities and column densities are used. The first method, which was introduced in §3.3.1, only counts cells that satisfy the density threshold condition and that are located within a 0.1 pc radius of the center of the simulation domain. This criterion counts only the densest and the most limited part inside the core resulting in the $\bar{n}_{\text{H},\text{max}}$ and $\bar{N}_{\text{H},\text{min}}$. However, since from an observational point of view it is hard to make such a selection based on such a density threshold, in our second method this density threshold condition is removed, while still keeping the 0.1 pc radius constraint. This estimate gives us $\bar{n}_{\text{H},\text{min}}$ and $\bar{N}_{\text{H},\text{max}}$ because of the extended low density region. It is noticeable that the average volume density estimated by the latter method remains steady at around $10^6 (\text{cm}^{-3})$ during the course of the simulation. The average density in the actual N_2D^+ -emitting cells is about twice as large at

the beginning and seven times larger at the end. Similarly, the two methods give a range of column densities, with the differences being at the level of about a factor of three at early times and less than a factor of two at late times. Overall, \bar{N}_{H} stays in a range from 10^{23} to 4×10^{23} . For the first method, i.e., with the density threshold condition (solid lines), $\bar{N}_{\text{H},\text{min}}$ keeps growing in the x and y projections, but is almost constant in the z projection. For the second method, i.e., without the density threshold condition (dashed lines), $\bar{N}_{\text{H},\text{max}}$ is relatively constant (but with a slow, later time increase) in the x and y projections, while in the z projection it steadily decreases during most of the evolution of the simulation.

In Figures 11d, e and f, the column densities of N_2H^+ and N_2D^+ and the implied value of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ are examined during the evolution of the core and with the two methods of total H column density estimation, i.e., with and without the use of the density threshold. The column densities of N_2H^+ are seen to quickly establish relatively high values and then undergo a modest decline, while those of N_2D^+ undergo a more gradual rise and then reach near constant levels. There are relatively modest variations due to different methods of column density estimation and different viewing angles.

One promising direction to improve observational constraints is to survey a sample of PSCs in the same region, i.e., assumed to have similar cosmic-ray environmental conditions, and examine the distribution of N_2H^+ column densities to determine the maximum achieved value and the distribution up to this value. This information can then potentially better constrain the types of acceptable astrochemical models (including a combination of cosmic-ray ionization rates and achieved levels of N species depletion) that are valid for the cores in the region. Then associated measurements of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ can further constrain the astrochemical conditions, along with the dynamical evolution of the cores. In the next sub-section we examine the detailed kinematic and dynamical properties that can be derived for the core based on observations of its N_2D^+ emission.

3.4 Core Structure, Kinematics and Dynamics

Our goal here is to consider the structure, kinematics and dynamics of the PSC, especially as traced by its N_2D^+ emission. Note, all the simulations we have run have effectively the same dynamical properties, and indeed start with the same physical conditions of the particular choice of turbulent velocity field. However, of the models that we have presented so far, we consider that the CR4x_fDN100_fDCO1000 is best at reproducing observed features of massive PSCs. In particular, it can reach high levels of $D_{\text{frac}}^{\text{N}_2\text{H}^+} \gtrsim 1$ and also do this in a way that $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ increases with density (or column density), so that the core appears as a local maximum in the $D_{\text{frac}}^{\text{N}_2\text{H}^+}$ map. However, it remains to be established if its kinematic properties, as traced by N_2H^+ and N_2D^+ , match those of observed systems. Thus, here we focus on the kinematic and dynamical properties of the core in the CR4x_fDN100_fDCO1000 model as traced by these species.

3.4.1 Evolution of Core Size and Associated Mass

The evolution of $R_{\text{c,eff}}$ versus time in the simulation for the CR4x_fDN100_fDCO1000 model has been shown in Figure 11a, including for the three different viewing directions. To estimate the mass associated with the core, we apply the two methods, i.e., with and without density threshold, that we used to estimate column den-

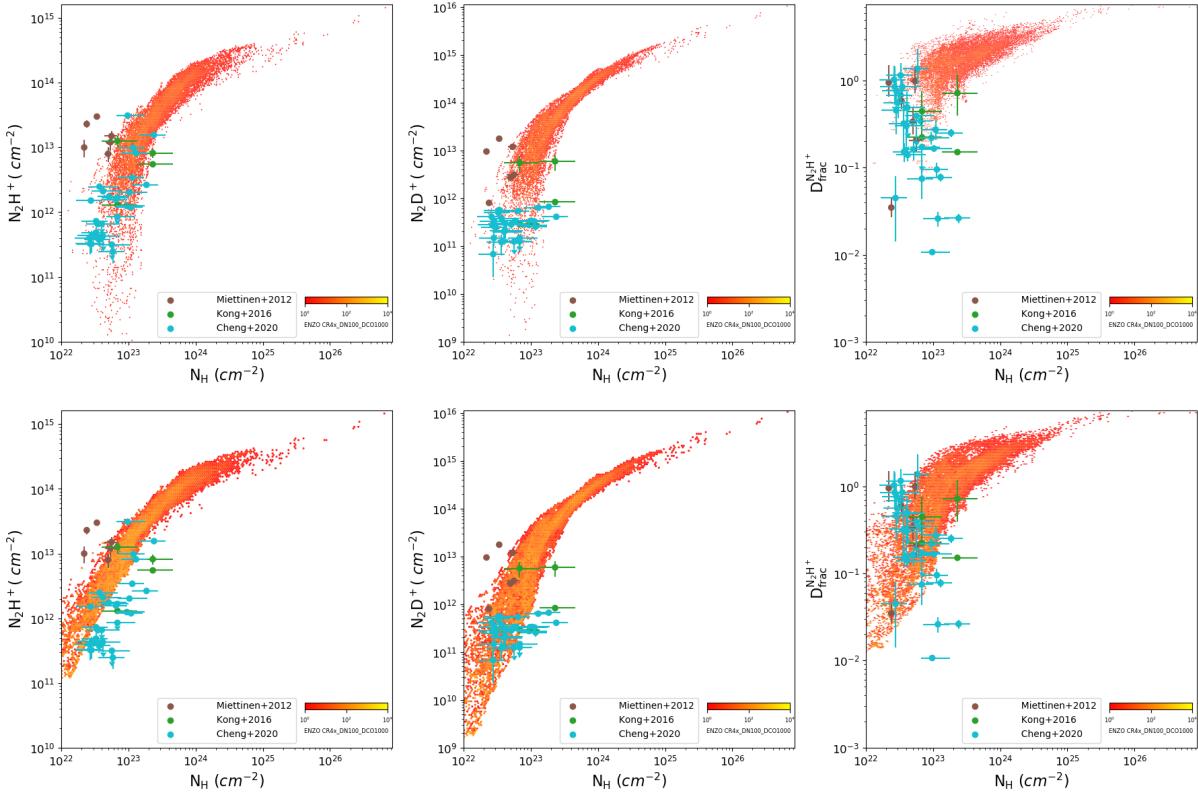


Figure 10. The panels from left to right show the column density of N_2H^+ , the column density of N_2D^+ and $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$ with respect to the column density of H nuclei (N_{H}) in the CR4x_fDN100_fDCO1000 model. The top row shows results based on the method of requiring a density threshold for N_2H^+ and N_2D^+ emission, while the bottom row shows the case without this threshold. Observational data points, which are the same as presented in Figure 9, are plotted as labelled.

sities in §3.3. Note, both methods require the core material to be within 0.1 pc distance of the initial center of core domain.

The time evolution of the core mass, M_c , measured in these different ways, including using different viewing directions, is shown in Figure 12a. From the physical model of a $60 M_{\odot}$ core, we see that the mass potentially traced by N_2D^+ increases from about $8 M_{\odot}$ at the beginning of the simulation to reach just over $30 M_{\odot}$ in the case of the first method requiring a density threshold (note, the same mass is derived independent of viewing direction). With the second method, the traced core grows from $17 M_{\odot}$ to reach just over $40 M_{\odot}$ by the end of the simulation (with minor variations caused by different viewing directions).

3.4.2 First Moment Maps, Velocity Gradients and Implied Rotation

In Figure 13, we show the first moment maps as viewed along each principal axis, weighted by N_2D^+ , of the core from simulation CR4x_fDN100_fDCO1000 at the final time corresponding to $t = 0.8t_{\text{ff}}$. Note that the large scale B -field is initially in the direction of the z -axis and we have already seen that the collapse occurs most easily along the direction of this axis. Thus the spatial distribution of the dense gas and the N_2D^+ emission is more concentrated towards the $z = 0$ plane.

We use the first moment maps to evaluate the implied average velocity gradient (magnitude and direction) in each image plane, with the averaging done using the column densities of N_2D^+ . Such a gradient can be interpreted as rotation of the core with an angular

momentum, J , vector in the plane of the sky. These vectors are also shown in each of the panels. The magnitudes of the velocity gradients are 5.57 , 23.1 , and $6.73 \text{ km s}^{-1} \text{ pc}^{-1}$ for the $y - z$, $x - z$ and $x - y$ planes, respectively. If we define north to be 0° in each panel, i.e., along the z -axis in the first two panels and along the y -axis in the third panel, then their position angles are -117° , -54° , and -53° , respectively. Since N_2D^+ is concentrated in the central region, the velocity gradients are dominated by the small area in the center. Overall, the kinematics traced by the first moment maps of N_2D^+ appear relatively disordered, which likely reflects the fact that the core is undergoing approximately free-fall global collapse from turbulent initial conditions, i.e., the collapse is only modestly impeded by the support of turbulence and magnetic fields.

In Figure 12b, we show the time evolution of the magnitudes of the velocity gradients of the core. Two ways are used to estimate the velocity gradient. The first way, illustrated in Figure 13, estimates the velocity gradient from the N_2D^+ -weighted radial velocities. For a second method, we use an area-weighted radial velocity from the first moment map to estimate a gradient. In Figure 12b, it is clear that the area-weighted velocity gradient is much smaller than the N_2D^+ -weighted velocity gradient by about one order of magnitude. Since the strongest gradients and the highest densities are concentrated near the center, the overall average core velocity gradient is flattened if it is weighted by area.

There are relatively few spatially resolved massive pre-stellar and/or early-stage protostellar cores observed in N_2D^+ . For the early-stage protostellar core C1-Sa, Kong et al. (2018) estimated a velocity difference of 0.8 km/s across its 0.044 pc diameter, i.e., a

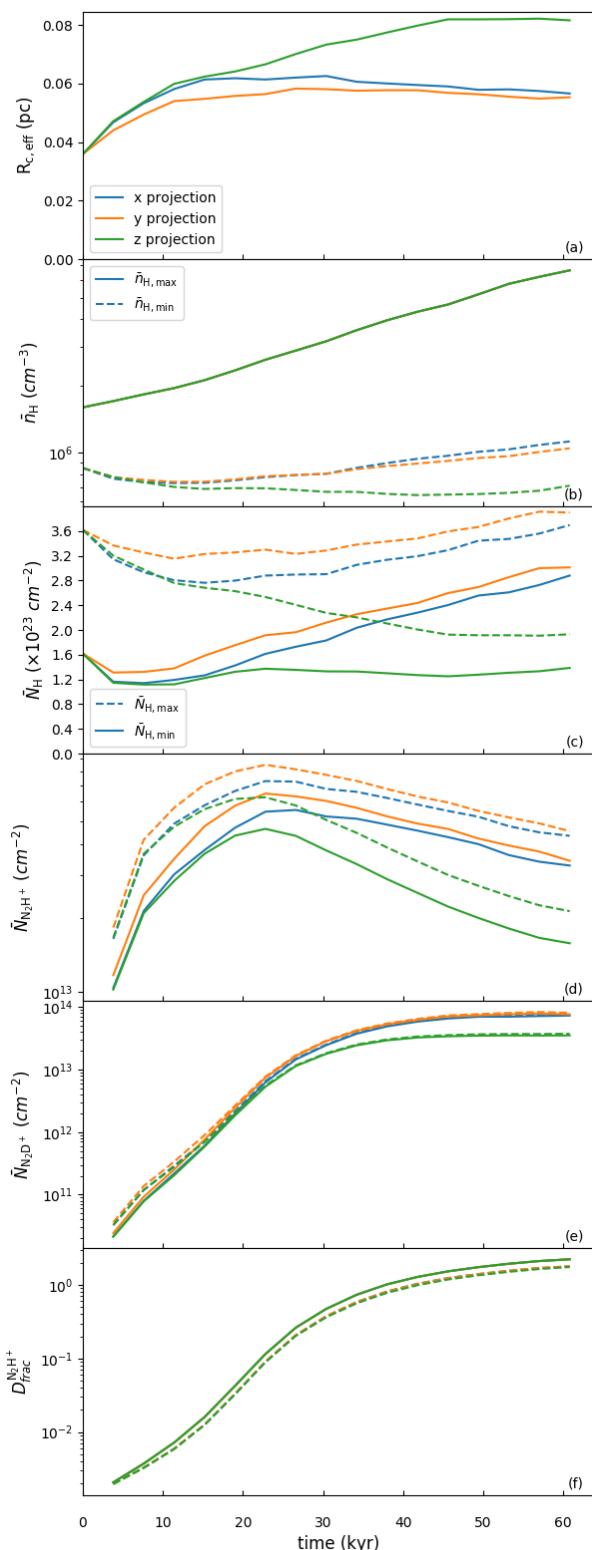


Figure 11. Time evolution of average quantities based on the N_2D^+ emission area (see text) of CR4x_fDN100_fDCO1000 model: (a) effective radius (here and in subsequent panels, results for x , y , z projection are shown with colors as labelled); (b) average number density of H nuclei (\bar{n}_{H}) (here and in subsequent panels solid line(s) show results considering only the cells where $n_{\text{H}} > n_{\text{H},\text{crit}}$; dashed lines are results when other, lower density cells along the line of sight are included in the average); (c) average column density (\bar{N}_{H}); (d) average column density of N_2H^+ ; (e) average column density of N_2D^+ ; (f) average $D_{\text{frac}}^{\text{N}_2\text{H}^+}$.

velocity gradient of about $18 \text{ km s}^{-1} \text{ pc}^{-1}$. Such a gradient should be compared to our area-weighted estimates: the velocity gradient of C1-Sa is much larger than we see in our PSC simulation. The reason for this could be that the gradient in C1-Sa is already dominated by the presence of the protostar and/or that it is an overall more compact structure (by a factor of about 3) compared to our simulated PSC. On the other hand, for the PSC C1-S, which at its outer scale has a radius of 0.045 pc, Kong et al. (2018) did not discern any significant velocity gradient in the N_2D^+ first moment map. From their Figures 9j and 9n, we see that most of the radial velocities are within 0.1 km/s of the core centroid velocity, but with a few localised regions, especially near the outer boundaries, that show velocity excursions up to about 0.5 km/s. Overall, a qualitative visual comparison between the first moment maps of our simulated core and those of C1-S suggests that C1-S is more quiescent, which likely indicates that it has a lower level of internal turbulence and/or smaller infall motions.

In the survey of Kong et al. (2017), the N_2D^+ -defined core C9A shows a strong velocity gradient in $\text{N}_2\text{D}^+(3-2)$: on one side the mean velocities are -1.5 km/s , whereas they are $+0.5 \text{ km/s}$ on the other. The core diameter is about 0.1 pc, so the velocity gradient is $\sim 20 \text{ km s}^{-1} \text{ pc}^{-1}$. Kong et al. (2017) speculated that given the bimodal morphology of its 0th-moment map, it is possible that two N_2D^+ cores are being seen in the process of merging. However, the magnitude of this velocity gradient is within the range seen for our simulated single PSC. The other large cores in the Kong et al. (2017) sample, i.e., B1A, B1B and H2A, do not show such large velocity gradients as C9A.

The average velocity gradients can be interpreted in terms of global rotation of the core and thus used to estimate the rotational energy: $E_{\text{rot}} = (1/2)I\omega^2 = J^2/(2I)$, where I is the moment of inertia. In many observational cases, especially those with limited resolution, a simplifying assumption for density distribution is often needed for the calculation of I . For example, for a uniform sphere $I = (2/5)M_cR_c^2$, while for a singular polytropic sphere $I = (2/3)[(3 - k_\rho)/(5 - k_\rho)]M_cR_c^2$ for a given power law index k_ρ .

For our simulated core, we can obtain the actual moment of inertia from the full density distribution. However, we choose to work with the 2D mass surface density distribution, which in principle could be observed in real PSCs, e.g., via high resolution extinction mapping, rather than with the 3D density distribution. Also the average velocity gradients that are derived from line-of-sight radial velocities yield a core rotational axis that is restricted to be in the plane of the sky. Thus we evaluate I with reference to this rotational axis, which runs through the centre of mass of the core as viewed in a given mass surface density map, i.e., for a given viewing direction. Thus, the defined core area is treated as a thin slab rotating along the axis.

With this method for estimating I , the rotational energy evolution of the core is shown in Figure 12c for the four combinations of two mass cases (red and blue colors) and two velocity gradient cases (solid and dashed line styles), described above. One sees that the different methods of estimating velocity gradients translate into about two orders of magnitude differences in the amount of rotational energy.

Next, we compare these estimates to the actual rotational energy of the core, but with this estimate still based on radial velocities along the line of sight to the projected plane, i.e., we sum the angular momenta of all the 3D cells with respect to the projected center of mass. Then the total rotational energy is estimated via $J_{\text{tot}}^2/(2I)$. This estimate is shown with the dot-dashed lines in Figure 12c. A conclusion from this comparison is that the rotational energy esti-

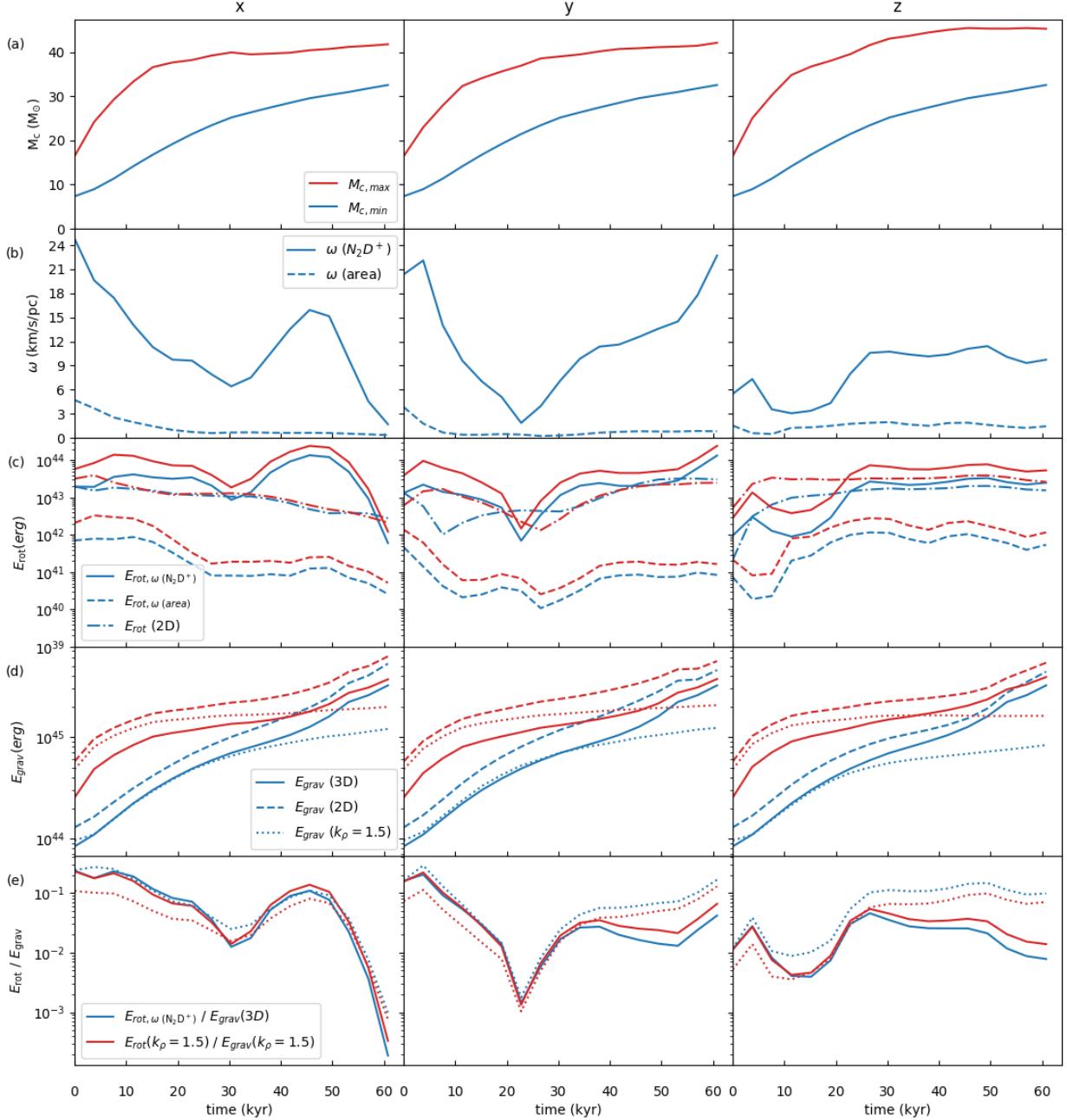


Figure 12. Time evolution of PSC kinematic and dynamic properties. The results based on different viewing directions, i.e., projecting along the x , y and z axes, are shown in the left, middle and right columns, respectively. From top to bottom: (a) core masses, $M_{c,\min}$ (blue line), estimated using only cells that meet the density threshold condition, and $M_{c,\max}$ (red line), estimated including lower density material along the line of sight (see text); (b) velocity gradient, ω , of N_2D^+ -weighted radial velocities (solid line) and area-weighted radial velocities (dashed line); (c) rotational energy, E_{rot} , estimated from (i) N_2D^+ -weighted velocity gradient and projected 2D plane moment of inertia (solid lines), (ii) area-weighted velocity gradient and projected 2D plane moment of inertia (dashed lines), (iii) the total angular momentum of each projected plane (dot-dash line); (d) gravitational energy estimated from (i) a simple polytropic model (dotted lines, $k_\rho = 1.5$), (ii) pairwise summation of cells in the 2D plane (dashed lines), (iii) pairwise summation of cells in 3D (solid lines); and (e) ratio of rotational energy to gravitational energy estimated from (i) 2D plane method (solid lines), (ii) polytropic model method (dotted lines).

mated by the flux-weighted N_2D^+ velocity gradient is much closer in value to the actual rotational energy of the core based on line of sight velocities.

Figure 12d shows the time evolution of the gravitational energy, E_{grav} , of the core. As with rotational energy, for observed cores one typically needs to make simplifying assumptions about the

mass distribution of the core in order to estimate E_{grav} . For the case of an uniform sphere, $E_{\text{grav}} = (3/5)GM_c^2/R_c$. In the case of a singular polytropic sphere with density power law index k_ρ , $E_{\text{grav}} = (3 - k_\rho)GM_c^2/(5 - 2k_\rho)R_c$ (Bertoldi & McKee 1992). For the $k_\rho = 1.5$ index that is relevant for the initial condition of our PSC, $E_{\text{grav}} = (3/4)GM_c^2/R_c$, i.e., only a 25% difference compared

to the uniform sphere. This result is shown by the dotted lines in Figure 12d for the two different mass estimation methods, i.e., with and without the density threshold along the line of sight (blue and red, respectively).

If the mass surface density map of a core is well resolved, we can use this distribution to carry out a pair by pair calculation for each pixel to estimate the gravitational energy, i.e., only using projected separations in the sky plane. Figure 12d also shows this method, using dashed lines, for the simulated PSC. Since the distances between mass elements are underestimated in this method, E_{grav} is overestimated, i.e., compared to the singular polytropic sphere model estimate.

Finally, only in the case of a simulated PSC, we can use the full three-dimensional data to calculate the gravitational energy by pairing cells in 3D. This estimate is shown in Figure 12d with the solid lines. We see that over most of the evolution of the simulation this method has close agreement with the polytropic sphere model estimate for the case where a density threshold is adopted (blue lines). However, at late times the 3D pairwise estimate becomes significantly higher, as the mass distribution deviates away from the simple polytropic model. For the method without the use a density threshold (red lines), we see that for most of the evolution both the polytropic model estimate and the 2D pairwise estimate are significantly larger than the true value yielded by the 3D pairwise method.

In Figure 12e, the ratio of rotational energy to gravitational energy is plotted. We focus on an estimate based on the N_2D^+ -weighted velocity gradient rotational energy using the 2D mass surface density map divided by the gravitational energy using the same 2D map. We find that the ratio $E_{\text{rot}}/E_{\text{grav}}$ starts at a level of $\gtrsim 0.1$ when viewed along the x and y axes, but about an order of magnitude lower when viewed along the z axis. This variation is due simply to stochastic sampling of large scale modes of the turbulent velocity field, since it is present at $t = 0$. We note that the core was initialized with solenoidal turbulence, but with no large scale rotation beyond that arising from random sampling of this turbulent velocity field. In this sense the degree of rotation of these structures is the smallest it can be for such turbulent cores.

Then $E_{\text{rot}}/E_{\text{grav}}$ decays to a level of about 0.01 during the first 20 to 30 kyr, likely due to decay of the initial turbulent motions, but then shows an increase, especially when viewed along the x axis, driven by infall motions. Finally there is a later decline to smaller values, driven in part by the late-time increase in E_{grav} .

For polytropic models, the ratio of rotational energy and gravitational energy can be expressed as:

$$\frac{E_{\text{rot}}}{E_{\text{grav}}} = \frac{(3 - k_{\rho})MR^2\omega^2/[3(5 - k_{\rho})]}{(3 - k_{\rho})GM^2/[(5 - 2k_{\rho})R]} = \frac{(5 - 2k_{\rho})\omega^2R^3}{3(5 - k_{\rho})GM}. \quad (9)$$

We use this equation to estimate $E_{\text{rot}}/E_{\text{grav}}$ using unresolved core properties and the N_2D^+ -weighted velocity gradient, and also show this estimate in Figure 12e. We find that this simple method of estimating the energy ratio follows that based on the 2D mass surface density map reasonably well (within a factor of about two), especially at earlier times (up to ~ 30 kyr).

3.4.3 Velocity Dispersion and Implied Virial State

In Figure 14, we plot histograms of the velocity distribution of the CR4x_fDN100_fDCO1000 model viewed along the x and z axes at the final time step. The bin count in each cell is weighted by the abundances of either N_2H^+ or N_2D^+ . The histograms are normalized so that the area under each distribution equals unity. We

plot three kinds of histogram. The first is the velocity distribution of the original data, i.e., the bulk motions of the gas due to turbulence, infall, rotation, etc. (black lines). The second is the total velocity distribution of bulk motions plus thermal broadening of the N_2H^+ or N_2D^+ species (transparent blue lines). For the assumed temperature of $T = 15$ K, this thermal broadening is equivalent to a 1-D velocity dispersion of $\sigma \approx 0.065 \text{ km s}^{-1}$ for these species (note the 1-D sound speed in the gas given a mean particle mass of $2.33m_{\text{H}}$ is $0.230(T/15 \text{ K})^{1/2} \text{ km s}^{-1}$). The last histogram considers both the thermal broadening effect and the hyperfine structure of the $\text{N}_2\text{H}^+(3-2)$ emission or $\text{N}_2\text{D}^+(3-2)$ emission (green lines).

All these distributions are dominated by a strong central peak. The effect of thermal broadening is very minor in comparison to the intrinsic bulk velocity dispersion. In contrast, hyperfine structure broadening acts to smooth and widen the distribution, but we note that a Gaussian fitted to the central part of the distribution still retains a similar width.

The velocity dispersions of the intrinsic distributions (black lines in Figure 14) are $\sigma_{x,\text{N}_2\text{H}^+} = 0.685 \text{ km s}^{-1}$, $\sigma_{x,\text{N}_2\text{D}^+} = 0.848 \text{ km s}^{-1}$, $\sigma_{z,\text{N}_2\text{H}^+} = 0.859 \text{ km s}^{-1}$ and $\sigma_{z,\text{N}_2\text{D}^+} = 0.983 \text{ km s}^{-1}$. We see that the velocities along the x -axis have a narrower distribution and a higher peak. However, the velocity distribution along the z -axis does have a localized sharp central peak shape. Additionally, we see that N_2H^+ has a slightly narrower distribution than N_2D^+ . The reason for this is because we have assumed that N_2H^+ and N_2D^+ are emitted from the same volume (i.e., using the same density threshold). However, velocity dispersions are higher in the center due to infall motions, and this is where N_2D^+ is relatively enhanced with respect to N_2H^+ .

In Figure 15a, we plot the time evolution of the total 1-D velocity dispersion, $\sigma_{\text{N}_2\text{D}^+}$, as measured by the velocities of the N_2D^+ -emitting cells along the three viewing directions (black solid lines). Note that here, since we will use this velocity dispersion for a dynamical analysis, we include the thermal sound speed of 0.23 km s^{-1} added in quadrature to the bulk velocity dispersion as measured from the raw velocities of the N_2D^+ -emitting cells, weighted by their abundance of N_2D^+ . Note, to estimate this from an observed spectrum would require removing the contribution of thermal and hyperfine structure broadening, before then adding in the 1-D sound speed. Also, an observational study would need to correct for any optical depth effects in the measured spectrum.

In the x and y directions, the velocity dispersion at first decreases modestly due to the decay of the initial turbulence. It then increases again after about 0.5 t_{ff} . In the z direction it exhibits a more monotonic increase. The velocity dispersion generally stays in a range of 0.4 to 0.8 km/s, except for a late increase seen in the z direction, where it rises to just over 1.0 km/s.

In the context of the fiducial Turbulent Core model of McKee & Tan (2003), there is a predicted pre-stellar core velocity dispersion based on its mass and the mass surface density of the surrounding clump environment. The mass averaged velocity dispersion of the core in the fiducial case is given by (see also Tan et al. 2013):

$$\sigma_{\text{c,vir}} = 1.09 \left(\frac{M_c}{60 M_{\odot}} \right)^{1/4} \left(\frac{\Sigma_{\text{cl}}}{1 \text{ g cm}^{-2}} \right)^{1/4} \text{ km s}^{-1}. \quad (10)$$

Note, this assumes the core is in virial and pressure equilibrium. As the core starts to undergo collapse and/or turbulence begins to decay, one expects deviations from the estimate of eq. (10). If we adopt a fixed value of $\Sigma_{\text{cl}} = 0.3 \text{ g cm}^{-2}$, which was used to set up the initial PSC, then we can use the evolving estimates of M_c to calculate $\sigma_{\text{c,vir}}$. Virial velocity dispersion estimates using this method are shown in Figure 15a with dashed lines. Since we have two definitions

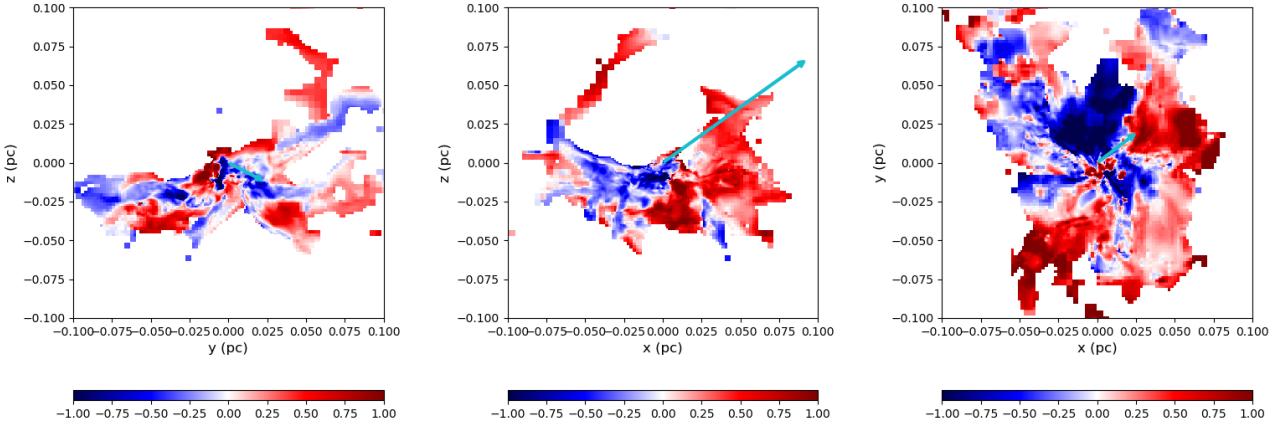


Figure 13. The first moment map along each principal axis at $t=0.8t_{\text{ff}}$. The cyan vectors indicate the directions and magnitudes of the average velocity gradients weighted by N_2D^+ flux on each plane.

of core mass, $M_{\text{c},\text{min}}$ and $M_{\text{c},\text{max}}$, the two estimates of velocity dispersion are plotted in blue and red, respectively. Alternatively, we can measure Σ_{cl} directly from the mass surface density map using the annular area of the region from R_{c} to $2R_{\text{c}}$, which follows the method applied to analysis of observed cores by Tan et al. (2013) and Kong et al. (2017). This estimate is also shown in Figure 15a (dotted lines). Again, we plot both estimates of $M_{\text{c},\text{min}}$ (blue) and $M_{\text{c},\text{max}}$ (red). We see that the actual value of $\sigma_{\text{N}2\text{D}^+}$ is quite similar to the estimates based on virial equilibrium. However, for much of the evolution and for two out of the three viewing directions, the core can appear kinematically sub-virial, i.e., with its observed velocity dispersion being lower than the virial equilibrium prediction by up to a factor of two. There is a tendency for velocity dispersion to rise above the virial equilibrium prediction in the final stages of the evolution of the PSC, i.e., just before the onset of star formation.

Observationally, the studies of Tan et al. (2013) and Kong et al. (2017) have identified 12 PSC candidates from their N_2D^+ (3-2) emission in IRDCs. The ratio of the observed N_2D^+ (3-2) velocity dispersion to that predicted by equation (10) has been found to be equal to 0.83 ± 0.15 for the 6 cores studied by Tan et al. (2013) and 0.80 ± 0.08 for the 6 cores studied by Kong et al. (2017). We see that such ratios are quite consistent with the results shown in Figure 15a. Thus, what is to be concluded here is that the overall velocity dispersion of N_2D^+ emission in the simulated PSC is always within a factor of two of the value predicted by virial equilibrium, even though the core is undergoing quite rapid collapse.

The total 1D velocity dispersion, $\sigma \equiv \sigma_{\text{N}2\text{D}^+}$, allows us to estimate the total internal kinetic + thermal energy of the core, assuming $\sigma_{3\text{D}} = 3^{1/2}\sigma$. In Figure 15b, we plot the evolution of the kinetic + thermal energy estimated as $E_k = (3/2)M_c\sigma^2$ (solid lines), with the two different mass cases, i.e., with density threshold (blue) and without density threshold along the line of sight (red), as before. We also show the actual kinetic + thermal energy as estimated from the 3D velocities of the simulation data. The two methods generally track each other reasonably well, though with deviations that can be at the level of a factor of a few. We notice in the y direction the velocity dispersion is relatively low during much of the evolution compared to other directions, so this leads to a relatively larger deviation given the assumption of isotropic velocity dispersion that is made in the simple estimate. At late times, the actual kinetic + thermal energy in the core increases significantly, which we attribute to

the emergence of a modest number of very dense cells, with deep gravitational potentials that induce fast motions.

In Figure 15c, we plot the gravitational energy of the core material that is traced by N_2D^+ , which is a repeat of the information shown in Figure 12d, shown here for convenience. Then in Figure 15d we plot the evolution of the ratio of kinetic + thermal to gravitational energies, i.e., E_k/E_{grav} . The first ratio to consider is that of the actual kinetic + thermal energy estimated from 3D motions to that of the gravitational energy estimated from a pairwise cell calculation in 3D (solid lines) (see §3.4.2). This ratio starts with relatively high values, which can be > 1 , and then declines, which is due to the decay of the initial, supervirial turbulence. The ratio declines to about 0.4, which is close to virial equilibrium, and then bounces back to a range between 0.5 to 0.8.

Other methods of estimating E_k/E_{grav} are those that can be done observationally, i.e., with E_k estimated from σ of N_2D^+ and E_{grav} from either a 2D mass surface density map (i.e., 2D pairwise summation) (dashed lines) or from the simple spherical polytropic assumption (dotted lines). If we consider the density threshold (blue lines), the 2D method yields an estimate generally within a factor of two of the actual 3D result, but the error could be larger in the later stages. The polytropic estimate yields a slightly better accuracy, within about a factor of two of the 3D result. When the density threshold is not used, then the mass is overestimated, leading to lower values being inferred for E_k/E_{grav} .

The magnetic energy of the core is expected to play an important dynamical role in the evolution of the core. We calculate $E_{\text{mag}} = \int B^2/(8\pi)dV$, with the integration approximated as a summation over the volume of the cells that make up the volume of the core. The ratio $E_{\text{mag}}/E_{\text{grav}}$ is shown in Figure 15e. For the case with the core mass defined with the density threshold (blue line) there is a short rising phase in the first 10 kyr when this ratio approaches unity. After this there is a decline to values approaching ~ 0.1 . For the core defined without the density threshold used along the line of sight (red lines), the ratio shows a simpler, monotonic decline from about 0.7 to about 0.2.

The ratio of total internal energies ($E_k + E_{\text{mag}}$) to gravitational energy is shown in Figure 15f. For the actual energy ratio, it shows a declining evolution from values ~ 2 down to a range of 0.5 to 1. Applying the kinetic + thermal energy estimated from $\sigma_{\text{N}2\text{D}^+}$ gives an even monotonic decline down to a range of 0.3 to 0.5.

In Figure 15g, we plot the virial parameter, $\alpha \equiv 5\sigma^2 R_c/(GM_c)$, which is often used in observational studies and is equivalent to twice the ratio of kinetic energy and gravitational energy in the uniform sphere case. The virial parameter decreases in the first $0.5 t_{\text{ff}}$ in the x and y directions, but keeps increasing in the z direction. Its value ranges from 0.3 to 3.0. For much of the evolution the virial parameter can take values that are < 1 , so that a core would be interpreted as being kinematically "sub-virial". We see that the value also depends on how the mass is measured in the core, i.e., illustrating the importance of background subtraction methods to remove the contribution of material along the line of sight.

Finally, as another metric to examine the collapsing state of the core, we plot the mass surface density distribution function (area weighted) in Figure 16. A collapsing core is expected to develop regions of very high values of mass surface density, which could serve as a diagnostic of the evolutionary state of the PSC. From the distributions shown in Figure 16, we do see the steady development of a high- Σ tail to the distribution, however the shape is relatively constant in the range from 0.1 to 100 g cm^{-2} from $t = 0.6$ to $0.8 t_{\text{ff}}$.

Observationally, mid-infrared extinction mapping methods (e.g. Butler & Tan 2012; Butler et al. 2014) based on *Spitzer*-IRAC $8 \mu\text{m}$ imaging data of Galactic plane infrared dark clouds have relatively poor angular resolution of $2''$ and tend to saturate at values of $\Sigma \sim 0.5$ to 1 g cm^{-2} . Thus such maps are not currently able to probe most of the dynamic range shown in the Σ -PDFs of Figure 16. High angular resolution observations of sub-mm/mm dust continuum emission, e.g., with *ALMA*, are another method to probe the Σ -PDFs of PSCs, but one would also need to allow for the potential effects of temperature and (grain growth/composition driven) opacity variations (e.g., Lim et al. 2015; Demyk et al. 2017a,b) in the core.

4 SUMMARY AND CONCLUSIONS

We have presented a simulation of a massive, magnetized PSC, i.e., the starting assumption of the Turbulent Core Model of massive star formation (McKee & Tan 2002; McKee & Tan 2003). The main novel aspect is the self-consistent coupling of a detailed astrochemical gas phase network that follows the evolution of a number of species that can trace dense, cold gas, especially N_2H^+ and N_2D^+ . The latter species in particular has been used as a diagnostic tracer of massive PSCs (e.g., Tan et al. 2013; Kong et al. 2017, 2018). This simulation work extends the methodology of a previous study of Goodson et al. (2016), which only approximated the chemistry with semi-analytic growth functions for the abundances of these species and which was a lower resolution, uniform grid simulation. One of the main differences of this work compared to that of Goodson et al. (2016) is that we predict lower levels of deuteration. The difference is because the previous semi-analytic method does not explicitly follow the time evolution of other species, especially ortho- and para- H_2 . Furthermore, this previous method was not able to fully follow advective diffusion in a turbulent core.

With the methodology developed here, we first investigated a fiducial astrochemical model and found that only relatively low levels of deuteration could be achieved during the approximately free-fall collapse, i.e., only reaching $D_{\text{frac}}^{\text{N}_2\text{H}^+} \sim 0.002$, which is much lower than found in many observed PSCs. We then carried out an extensive survey of different astrochemical model parameters and properties, i.e., initial ortho-to-para ratio of H_2 , cosmic-ray ionization rate, heavy element depletion factors, degree of chemical

pre-aging, to investigate how high levels of deuteration, approaching unity, can be achieved, which are needed to explain the observed PSCs. We have found that a model with relatively high cosmic-ray ionization rates ($1.0 \times 10^{-16} \text{ s}^{-1}$) and high depletion factors of CO (1000) and N (100) is able to match observed properties of some PSCs, i.e., $D_{\text{frac}}^{\text{N}_2\text{H}^+} \sim 0.1$ to 1 and absolute abundances of these species, within the followed evolution of the core, i.e., within $0.8 t_{\text{ff}} = 61 \text{ kyr}$. However, it should be emphasised that this is just an example model and other combinations of parameters are expected to also be able to give similar agreement with these observational metrics. One additional metric that may help break degeneracies is the spatial structure of $D_{\text{frac}}^{\text{N}_2\text{H}^+}$, i.e., how it correlates with column density. However, to date such observations have not yet been acquired to be able to test the models.

In kinematic and dynamic aspects, we examined the capability of N_2D^+ as a tracer. We used the N_2D^+ abundances to imitate observational analyses. This showed that the velocity gradients estimated by N_2D^+ give a good estimation of the actual rotational to gravitational energy ratio based on the assumption of a polytropic core structure model. We also found that the PSC, as defined by its N_2D^+ emission, can often appear kinematically sub-virial, even though it is undergoing quite rapid collapse. High resolution observations of the velocity structures of massive PSCs as traced by N_2D^+ are one promising method to help test models of PSC evolution. From the simulation side, a broader range of physical models, e.g., with a varying degree of B -field support and with a range of initial conditions that set core formation, are needed to explore their effects on these kinematic signatures.

Some caveats of our work include that, for simplicity, we do not include freeze-out and desorption reactions and surface chemistry. Instead we control the gas phase C, O and N abundances by parameterized depletion factors. However, the depletion rate of CO is dependent on density. The CO depletion factor is likely to have a spatial structure. In our fDCO100 and fDCO1000 models, it is obvious that they reach different levels of deuterium fractionation. If there is a spatial structure of CO, this could influence the final result of our simulation. In Bovino et al. (2019), they discuss the CO depletion structure of their simulated high-mass pre-stellar region. Their result shows that the CO depletion does have a spatial structure and thus influences the structure of deuterium fractionation also. Thus an additional observation metric of CO depletion structure is likely to be helpful to give better constraints on our PSC models, which ultimately should include time dependent depletion and desorption processes and surface chemistry, although these aspects introduce many additional uncertainties for the modeling. Another, similar caveat is our choice of uniform cosmic-ray ionization rate, rather than allowing for attenuation of this rate in higher density, more shielded regions. Such effects will also be investigated in future work that extends the analysis of this study.

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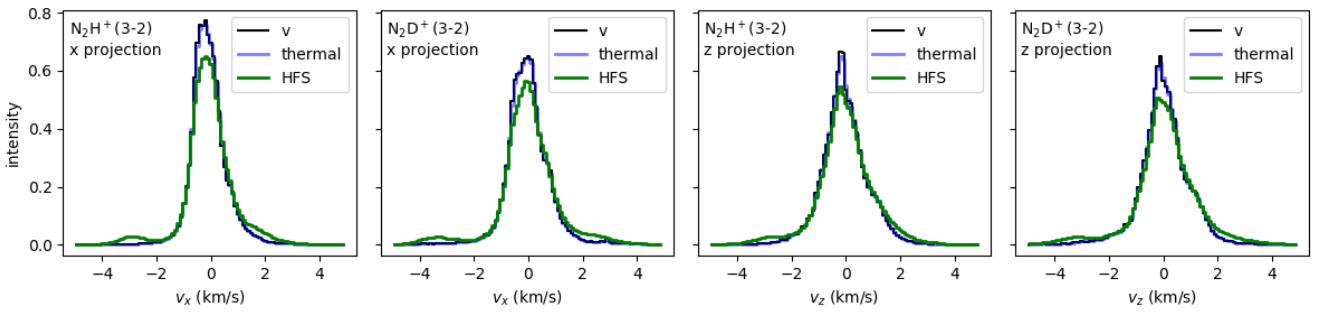


Figure 14. The velocity distribution weighted by N_2H^+ and N_2D^+ abundances along the x -axis and z -axis of fDN100_fDCO1000 model. The left two panels show the distributions in v_x , weighted by N_2H^+ and N_2D^+ as labelled. The right two panels show the distributions in v_z in the same format. The black lines show intrinsic velocity distributions of the snapshot at $0.8 t_{ff}$. The transparent blue line shows the results including thermal broadening at $T = 15$ K. The green line includes both thermal broadening and hyperfine structure broadening of either $\text{N}_2\text{H}^+(3-2)$ emission or $\text{N}_2\text{D}^+(3-2)$ emission. The bin width of the histograms is 0.1 km/s .

DATA AVAILABILITY

The data underlying this article will be shared on reasonable request to the corresponding author.

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APPENDIX A: COMPARISON OF CHEMODYNAMICAL METHODS

Comparing the results of our fiducial case with the corresponding results in Goodson et al. (2016), we notice an apparent discrepancy in the estimated deuteration level. The peak value of deuterium fraction is about 100 times smaller in our simulation. We check

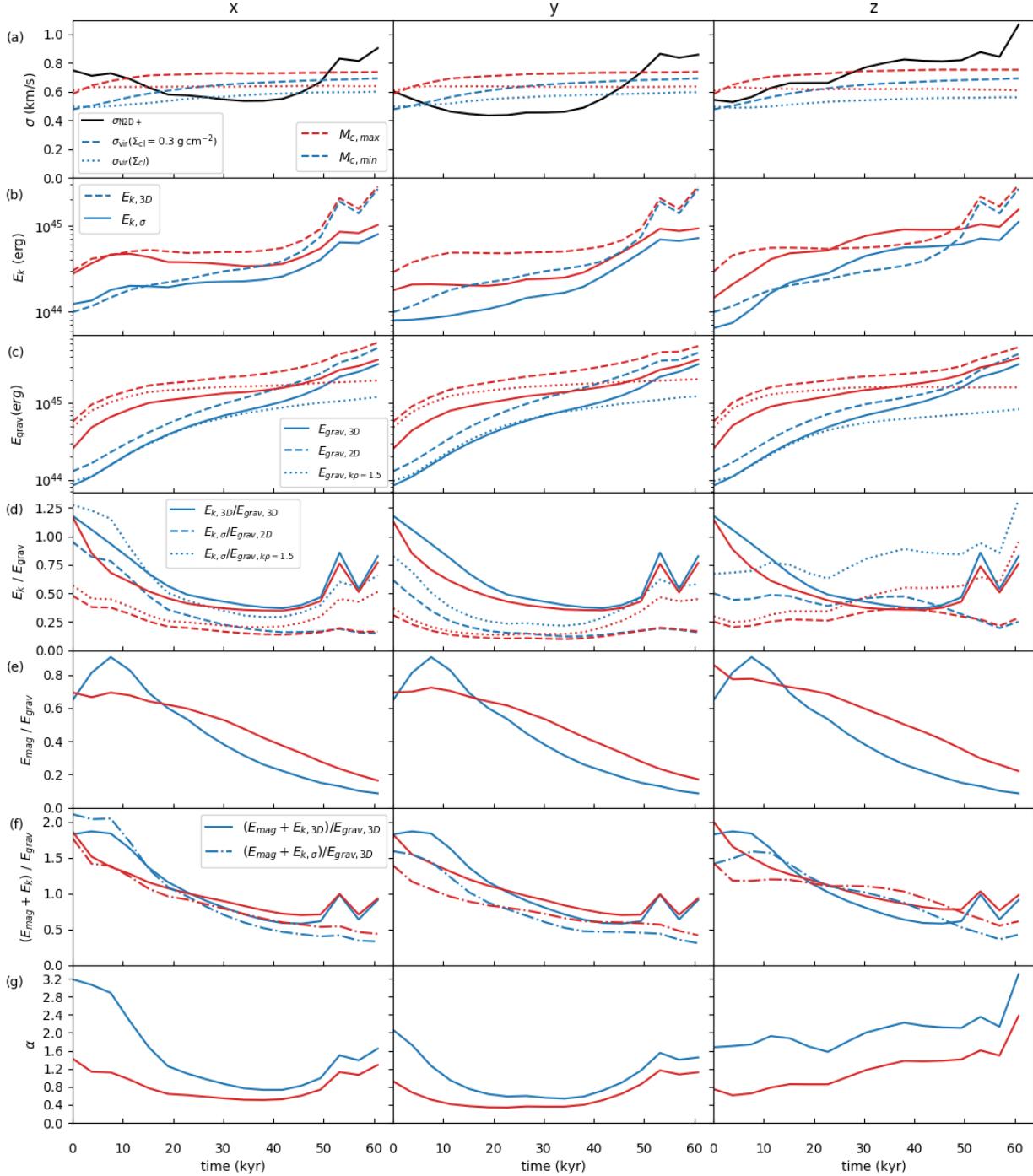


Figure 15. Time evolution of various dynamical properties of the PSC in the CR4x_fDN100_fDCO1000 simulation, with results from the x , y , z viewing directions shown in the columns left, middle, right, respectively. (a) Top row: Total one-dimensional velocity dispersion as traced by N_2D^+ , $\sigma_{\text{N}2\text{D}^+}$, including the thermal sound speed of 0.23 km s^{-1} added in quadrature (black solid lines). Also shown are estimates of the virial equilibrium velocity dispersion for two different assumptions about the clump mass surface density (a constant value of 0.3 g cm^{-2} - dashed lines; a local estimate around the core - dotted lines) and for two mass estimation methods of the core, i.e., with (blue) and without (red) density threshold along line of sight (see text). (b) Second row: Core kinetic + thermal energy, E_k , estimated from the 1D velocity dispersion traced by N_2D^+ (solid lines) and from the actual 3D motion in the simulation (dashed lines), again for the two mass cases described in (a) (blue and red). (c) Third row: Core gravitational energy, E_{grav} , estimated by (i) pairing cells in 3D data (solid lines), (ii) pairing cells in the 2D projected plane (dashed lines), (iii) a polytropic model with $k_\rho = 1.5$ (dotted lines). (d) Fourth row: Ratio of kinetic + thermal energy to gravitational energy estimated by (i) kinetic + thermal energy and gravitational energy from 3D data (solid lines), (ii) kinetic + thermal energy estimated by $\sigma_{\text{N}2\text{D}^+}$ and gravitational energy estimated by pairing cells in the 2D projected plane (dashed lines), (iii) kinetic + thermal energy estimated by $\sigma_{\text{N}2\text{D}^+}$ and gravitational energy estimated by polytropic model with $k_\rho = 1.5$ (dotted lines). (e) Fifth row: Ratio of core magnetic energy, E_{mag} , to gravitational energy (see text). (f) Sixth row: Ratio of the sum of kinetic, thermal and magnetic energies to gravitational energy estimated by (i) kinetic + thermal energy and gravitational energy from 3D data (solid lines), (ii) kinetic + thermal energy estimated by $\sigma_{\text{N}2\text{D}^+}$ and gravitational energy from 3D data (dot-dashed lines). (g) Bottom row: Time evolution of virial parameter based on $\sigma_{\text{N}2\text{D}^+}$ and effective radius for the two mass estimation methods.

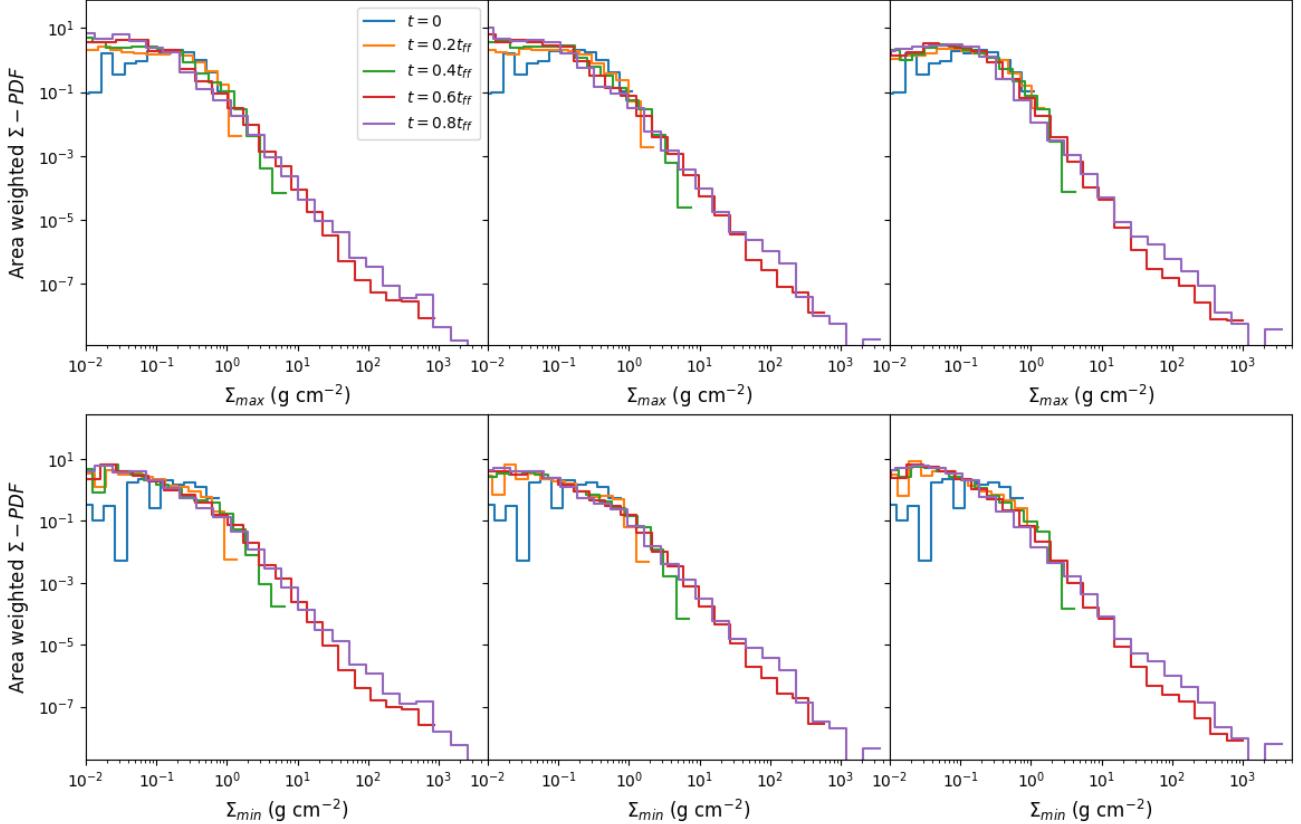


Figure 16. Probability distribution functions of mass surface density. (a) Top row: Distribution of mass surface densities including cells along the line of sight of the emission area. The results for x , y and z projections are shown in the left, middle and right panels, respectively. The different colors show the time evolution of the distributions (see legend). (b) Bottom row: As (a), but now for mass surface density only of the cells where $n_{\text{H}} > n_{\text{H},\text{crit}}$.

the abundances of N_2H^+ and N_2D^+ and notice that N_2H^+ has a good agreement but N_2D^+ was much larger in Goodson et al. (2016) compared to our results (Figure A1). The agreement of N_2H^+ comes from the equilibrium condition. Since the timescale for N_2H^+ to achieve an equilibrium abundance is shorter than one free-fall time, both simulations reach near equilibrium values and match with each other very well. In contrast, the timescale for N_2D^+ to reach its equilibrium state is much longer. It is impossible for N_2D^+ to reach the equilibrium value in one free-fall time (Figure A2). However, we notice N_2D^+ reaches the equilibrium value in the densest grids in Goodson et al. (2016) (Figure A1). The reason why N_2D^+ reaches the equilibrium state in a much shorter time is a result of the approximate interpolation method. The growth rate of N_2D^+ and N_2H^+ are estimated by their current abundances respectively. The interpolation method ignores the information of other species, such as OPR^{H_2} , which suppress the growth of $\text{D}_{\text{frac}}^{\text{N}_2\text{H}^+}$.

Considering two single grid models, which have the same initial parameters as the fiducial model (see Table 1), with $n_{\text{H}} = 10^5$ and 10^7 cm^{-3} , their N_2D^+ growth curves are plotted in Figure A3. If N_2D^+ is 10^{-13} in one grid, the grid could have evolved for around either 27,000 years in the former case or 160,000 years in the latter case. It means they have different history and different chemical ages. Goodson et al. (2016) used the local abundance to predict the growth of N_2H^+ and N_2D^+ . If we check OPR^{H_2} in Figure A3, we even find that they corresponds to $\text{OPR}^{\text{H}_2} \approx 0.091$ and 0.023 respectively. Goodson et al. predicted the growth of N_2D^+ according

to the point data where $\text{OPR}^{\text{H}_2} = 0.023$, but the real OPR^{H_2} is around 0.091 and this suppresses the growth of N_2D^+ . Since the high-density region (e.g., $n_{\text{H}} = 10^7 \text{ cm}^{-3}$) inherits the value from the low-density (e.g., $n_{\text{H}} = 10^5 \text{ cm}^{-3}$) region, it finally gives an overestimated result.

APPENDIX B: ELEMENTAL CONSERVATION

In adaptive mesh refinement (AMR) frameworks, interpolation can cause errors in elemental conservation because of gradient limiters and interpolation weights could be different for each species (Plewa & Müller 1999; Grassi et al. 2017). This means that the total abundances of C, N and O would not be conserved, even if each individual chemical species is conserved during the interpolation. To check elemental conservation, we compare the total abundances of C, N and O with their initial values cell by cell. We find that their mean values have a typical $\sim 2\%$ error for each element in our defined cores. For example, Figure B1 shows the ratio of $[\text{N}]/[\text{N}]_{\text{init}}$. The mean value is 0.977 and the standard deviation is 0.021. Other elements, i.e., C and O, have similar distributions. The error is even smaller if we consider a more extensive domain because the errors are concentrated in the refined regions within our defined cores. Such levels of elemental conservation errors do not significantly influence our results.

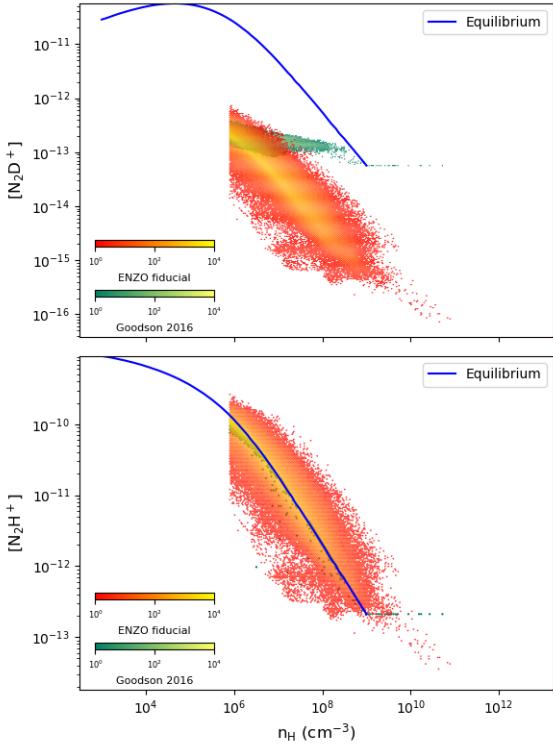


Figure A1. The distribution of N_2D^+ and N_2H^+ versus column density. The blue line indicates the equilibrium value used in Goodson et al. (2016). N_2H^+ (lower panel) matches with this equilibrium value well in both simulations, but the behaviour of N_2D^+ is very different between the two simulations: i.e., in our models gas at high density has a much lower abundance of N_2D^+ .

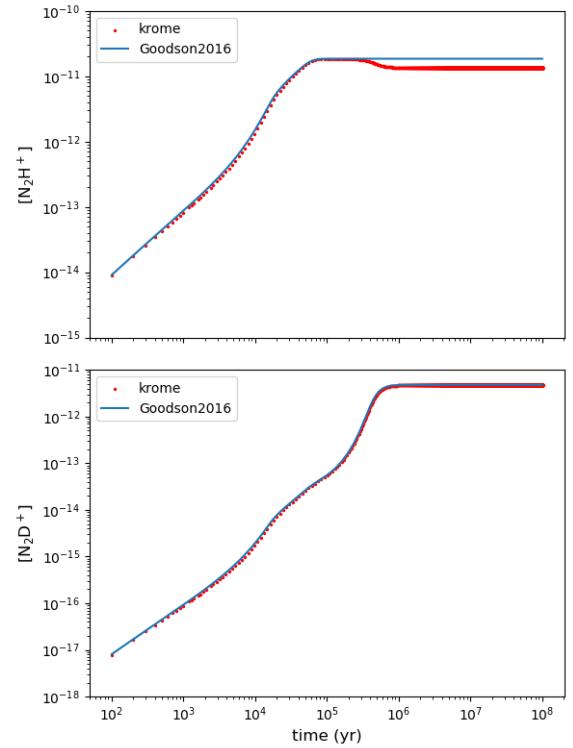


Figure A2. We run single grid models and compare the results with the interpolation data of Goodson et al. (2016), which corresponds to our fiducial model. The parameters have been listed in the Table 1. In single cells, they have a good agreement. We notice that the time scale required for N_2H^+ to achieve the equilibrium value is shorter and is similar to one free-fall time. By contrast, the timescale for N_2D^+ to achieve equilibrium value is longer than one free-fall time.

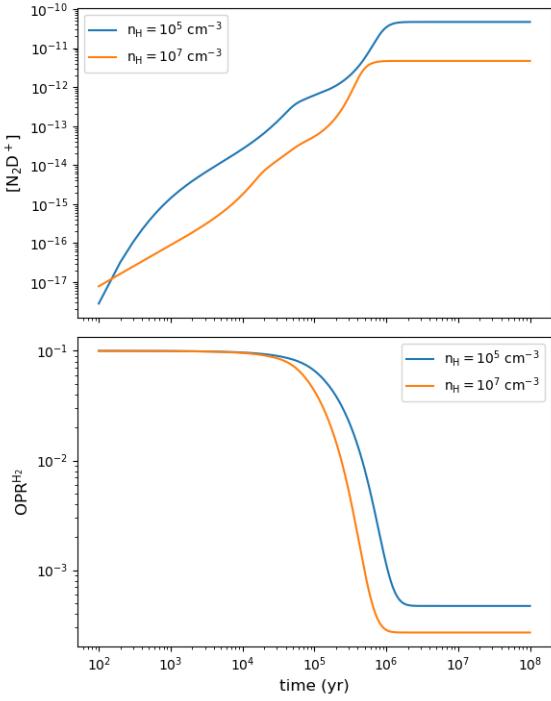


Figure A3. The evolution of N_2D^+ in single grid models. We use two different densities here: 10^5 and 10^7 cm^{-3} .

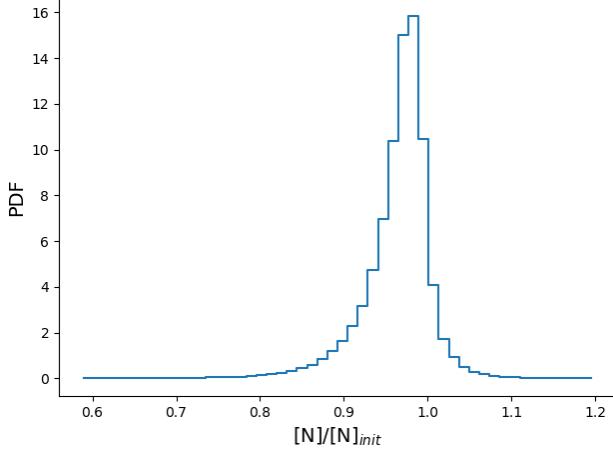


Figure B1. The ratio of $[\text{N}]/[\text{N}]_{\text{init}}$ of cells in the defined core.