



Transport Models for Galactic Cosmic Rays Results of a Markov-Chain-Monte-Carlo Study

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$$\frac{\partial \Psi}{\partial t} = q(\vec{r}, p) + \vec{\nabla} \cdot (D_{xx}\vec{\nabla}\Psi - \vec{V}\Psi) + \frac{\partial}{\partial p}p^2 D_{pp}\frac{\partial}{\partial p}\frac{1}{p^2}\Psi - \frac{\partial}{\partial p}[\dot{p}\Psi - \frac{p}{3}(\vec{\nabla}\cdot\vec{V})\Psi] - \frac{1}{\tau_f}\Psi - \frac{1}{\tau_r}\Psi$$





$$\frac{\partial \Psi}{\partial t} = \overline{q(\vec{r}, p)} + \vec{\nabla} \cdot (D_{xx}\vec{\nabla}\Psi - \vec{V}\Psi) + \frac{\partial}{\partial p}p^2 D_{pp}\frac{\partial}{\partial p}\frac{1}{p^2}\Psi - \frac{\partial}{\partial p}[\dot{p}\Psi - \frac{p}{3}(\vec{\nabla}\cdot\vec{V})\Psi] - \frac{1}{\tau_f}\Psi - \frac{1}{\tau_r}\Psi$$





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- Source distribution
- Diffusion
- Convection (galactic winds)





Aim: Reproduction of the locally measured spectra of cosmic rays
Numerical solution of the transport equation (e.g. GalProp, Dragon)

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4 kpc

. ..

60 kpc



- Diffusion
- Convection (galactic winds)
- Diffusive reacceleration





$$\frac{\partial \Psi}{\partial t} = \overline{q(\vec{r}, p)} + \overline{\nabla} \cdot (D_{xx} \overline{\nabla} \Psi) - \overline{\nabla} \Psi + \frac{\partial}{\partial p} p^2 D_{pp} \frac{\partial}{\partial p} \frac{1}{p^2} \Psi - \frac{\partial}{\partial p} [\dot{p} \Psi - \frac{p}{3} (\overline{\nabla} \cdot \overline{V}) \Psi] - \frac{1}{\tau_f} \Psi - \frac{1}{\tau_r} \Psi$$





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Fitness of model by comparing results with experimental data: $p, \bar{p}, \frac{\bar{p}}{p}, \frac{B}{C}, \frac{Be^{10}}{Be^{9}}$

Why use MCMC and not a simple minimizing algorithm?



- \rightarrow not feasible for such high dimensional problems! Evaluation of a single model can take minutes to hours
- → We do NOT want to find a best fit model Instead, want to explore wide ranges of parameter space (exotic models?) Examine the full potential of these kind of models
- \rightarrow Use MCMC in order to **sample** parameter space in an efficient way

Markov-Chain-Monte-Carlo



Decider

Metropolis-Hastings-Algorithm

(Metropolis, N., Rosenbluth, A.W., Rosenbluth, M.N. Teller, A.H., 1953)

Acceptance criterion by comparing 2 models

$$\alpha \propto \frac{F(n)}{F(p)}$$



Density of chain points ~ Fitness!

Markov-Chain-Monte-Carlo

Decider

Multiple-Try-Metropolis-Algorithm (MTM) (Liu, J. S., Liang, F. and Wong, W. H., 2000)

Acceptance criterion by comparing 2 sets of models

$$\alpha = \min(1, \frac{F(p_1) + \dots + F(p_k)}{F(n_1) + \dots + F(n_{k-1}) + F(p_0)})$$

 \rightarrow Use much more information of parameter space \rightarrow Allows evaluation of several models in parallel





Markov-Chain-Monte-Carlo

Proposal functions (A. Putze et al. 1001.0551)

I. Gaussian function in every dimension

- Step size (σ) must be given
- Most probable point = current point

II. Multivariate Normal Distribution

- build empirical covariance matrix based on chain points of I.
- Correlations/step sizes are given

III. Next Neighbour (NN) Search

- Propose points according to the binned fitness distribution of all known points







Results Efficiency



Proposed points

chain points



Parameter 2

I. Gaussian function

II. Multiv. Norm. Distr.

III. NN Search



Results Efficiency



Results



→ more than 10 million models have been evaluated
> 1 year with 500 markov chains, running simultaneously



ISAPP 2013 - Djurönäset

What about leptons?





Assume local transport = global transport

Positron fraction is impossible to mimic with a pure secondary production of positrons

Most favored sources:

- Pulsars
- Dark Matter annihilation

→ large energy losses due to synch. radiation and inv. compton scattering → short propagation lengths of O(500pc) → *local transport,* highly affected by the local environment



Spiral arm structure of the Milky Way



The local bubble

Separate study, see Gebauer, SK, Weinreuter ICRC13