

Estimation of physical conditions in PDRs Bayesian approach with spatial regularization



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Aims and contributions

New large hyper-spectral surveys in radio-astronomy: game changer for study of star formation, feedback mechanisms and chemistry of interstellar gas. \implies now possible to observe full molecular clouds (10 pc size) at dense core scale (<0.1 pc) spatial resolution. For instance, the IRAM-30m Large Program «Orion B» [1] provides a 10⁶ pixel observation map, covering $\sim 250 \text{ pc}^2$ of the Orion B cloud with emission of dozens of molecules.

Goal To estimate maps of thermal pressure P_{th} , UV radiation intensity G_0 , visual extinction A_V , scaling factor κ and the associated uncertainties.

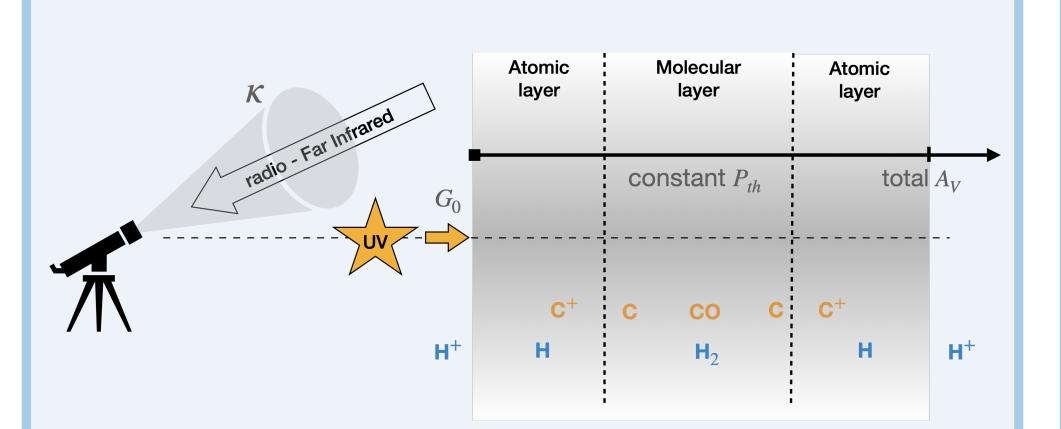
Challenge Variability of the SNR: the brightest regions can be well constrained, but the regions with low SNR lead to degenerate solutions.

Approach \star Mixture of both additive and multiplicative noises + limited detectability \implies more realistic observation model

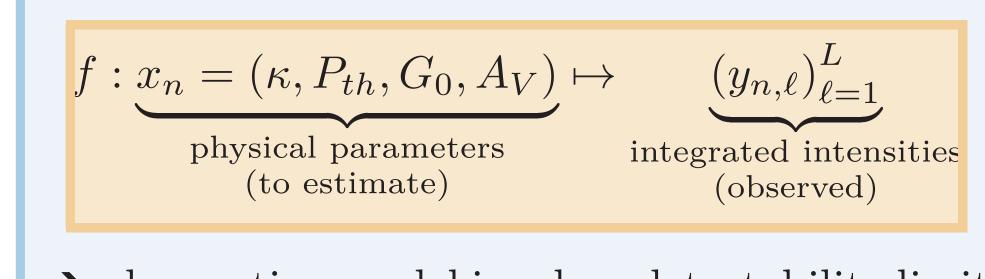
- \star Spatial regularization \implies exploitation of the information contained in neighbouring pixels.
- \star Bayesian approach \implies access to accurate credibility intervals.

Problem formulation

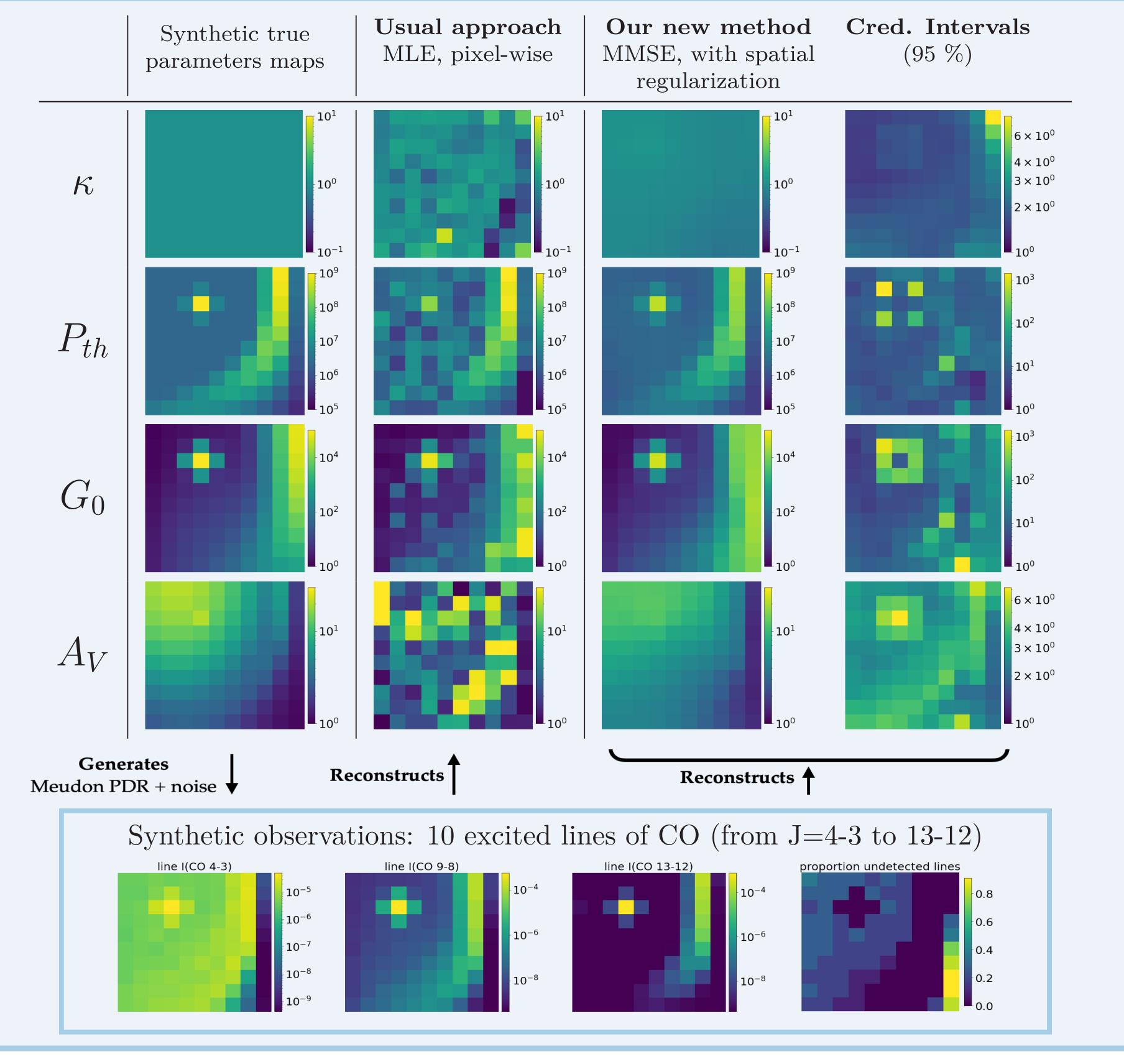
→ Meudon PDR code [2]: solves radiative transfer, chemistry, thermal balance on stationary 1D plane-parallel model.



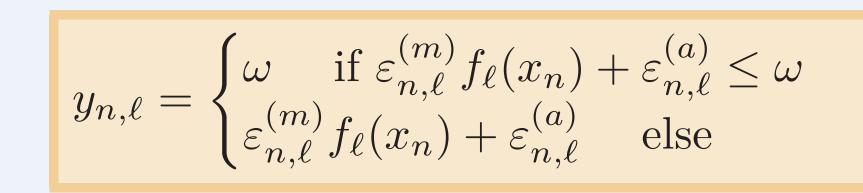
 \rightarrow Assumption: physics in PDR = interpolation of a grid of simulations of Meudon PDR code. Defines function f s.t. for one pixel n:



Application on toy dataset (100 pixels)



 \rightarrow observation model involves detectability limit ω , gaussian noise $\varepsilon_{n,\ell}^{(a)}$ and lognormal noise $\varepsilon_{n,\ell}^{(m)}$



defines a **likelihood** function $\pi(y \mid x)$.

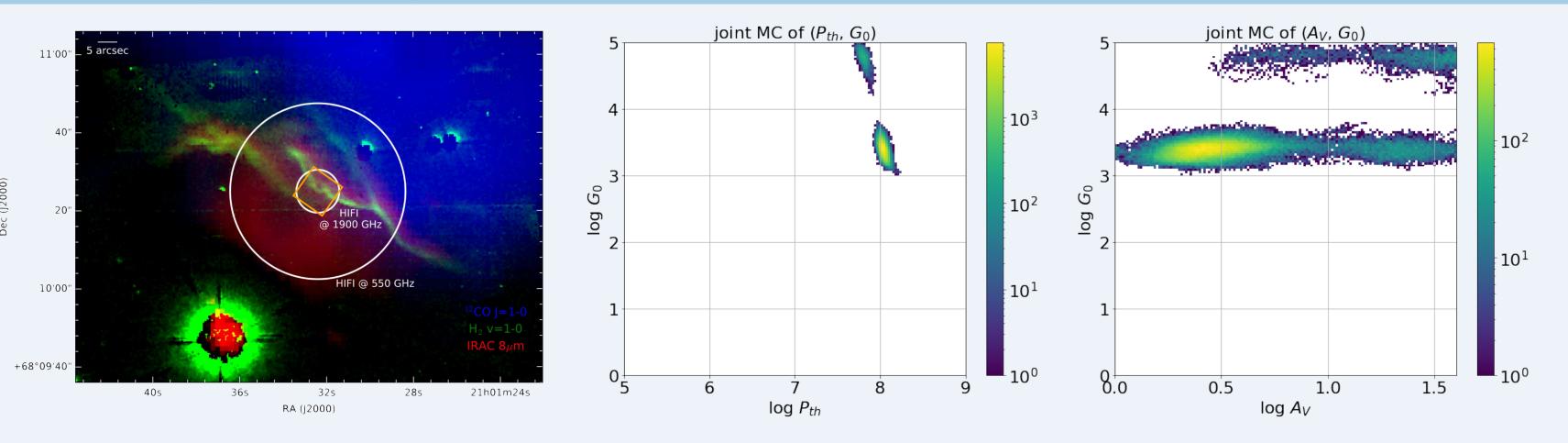
Bayesian approach

- \rightarrow State-of-the-art in millimeter/IR astronomy estimations: Maximum Likelihood Estimates $(MLE) \implies$ very sensitive to noise
- \rightarrow For robust estimators: spatial regularization **prior** $\pi(x)$ (norm of image gradient or laplacian)
- → **Posterior** probability density function:

 $\pi(x \mid y) \propto \pi(y \mid x)\pi(x)$

 \rightarrow To derive estimators and credibility intervals from posterior distribution: sample from it with Monte Carlo Markov Chain (MCMC).

NGC 7023 (1 pixel)



 \rightarrow Our sampler mixes two kernels: one identifies local minima, one efficiently explores them.

 \rightarrow Our ponctual estimator: MMSE (Mininimum) Mean Squared Error) = mean of posterior.

 \rightarrow Estimation from 10 lines of CO and 2 lines of CI.

→ In contrast with [3], G_0 and A_V estimated along with κ and P_{th} .

 \rightarrow Two regions of acceptable solutions identified.

Summary for main mode

	MLE $[3]$	MMSE	2.5%	97.5%
κ	0.7	4.8	2.5	8.6
$P_{th} \ (\times 10^8)$	1	1.1	1.0	1.3
$G_0 \; (\times 10^3)$	2.6	1.6	1.0	2.6
A_V	10	3.1	1.6	30.5

References

[1] Pety et al., The anatomy of the Orion B Giant Molecular Cloud: A local template for studies of nearby galaxies, 2016

[2] Le Petit et al., A model for atomic and molecular interstellar gas the Meudon PDR code, 2009

[3] Joblin et al., Structure of photodissociation fronts in star-forming regions revealed by Herschel observations of high-J CO emission lines, 2018

Conclusions

Achieved:

- \checkmark Spatial regularization \implies more robust estimations of maps of parameters,
- \checkmark Bayesian approach \implies more complete description than ponctual estimators.

Future Work:

- \rightarrow Application to other environments than PDR (shocks, dark clouds, etc.),
- \rightarrow Code acceleration to scale to ~ 10⁶ pixel observations (e.g., Orion B IRAM Large Programm).