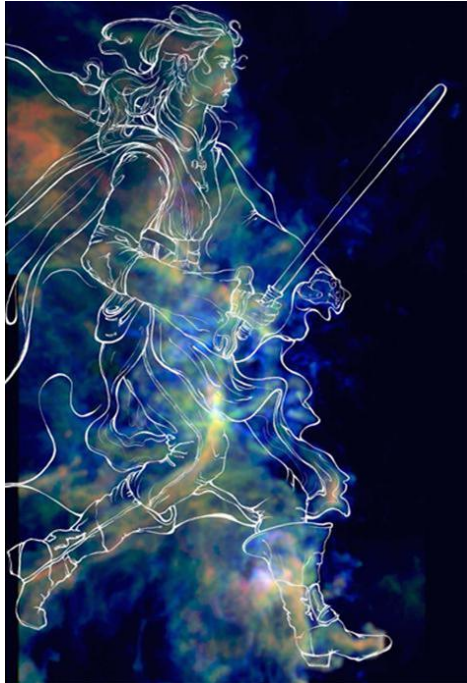
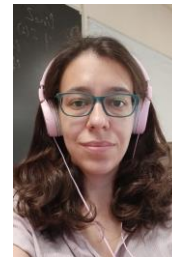


Millimeter rotational lines as powerful diagnostics of the physical conditions inside a Giant Molecular Cloud - The Orion B case



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Column Density estimation technique

$$M_{\text{gas}} \propto \mu m_{\text{H}} N(\text{H}_2) [M_{\odot}]$$

Dust-continuum-emission observations

X_{CO} : The CO-to- H_2 conversion factor

$$N(\text{H}_2) = 1.35 \cdot 10^4 \cdot T_{850\mu\text{m}} \cdot 1.9 \cdot 10^{21} \cdot \tau_{350}^{-1}$$

$$N(\text{H}_2) = X_{\text{CO}} \cdot W(\text{CO})$$

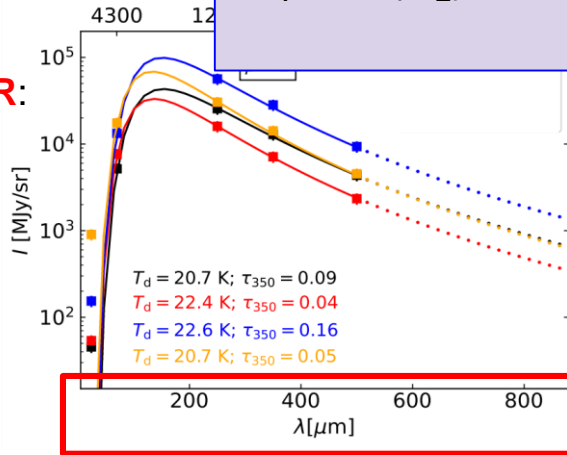
$$A_V \simeq N(\text{H}_2) \cdot 2 / 1.9 \cdot 10^{21} \text{ [mag]}$$

In GMCs ->

dust emits in the FIR:

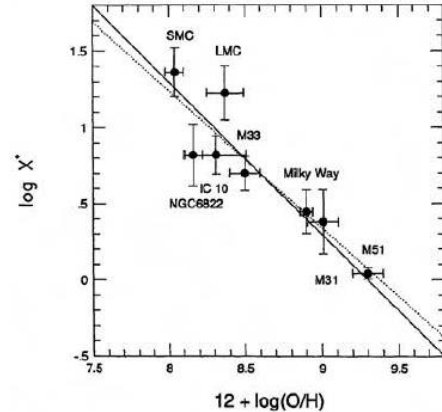
warm Big Grains

$T_{\text{dust}} \sim 15 - 80 \text{ K}$



% uncertainty in the inner disk
of the MW (Bolatto+13)

- Metallicity dependent (Arimoto+96)

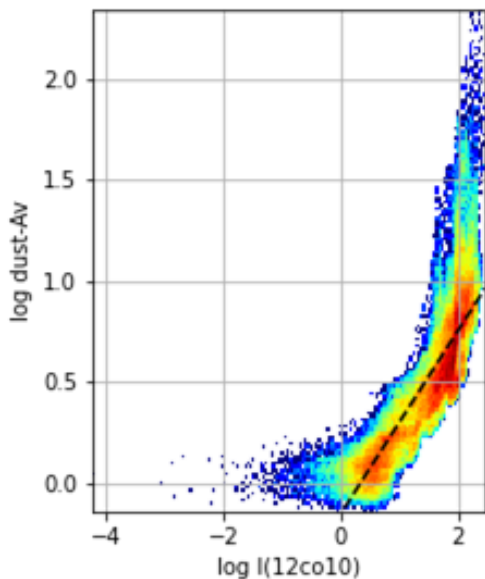


Need space telescopes like Herschel: None in the next decades.

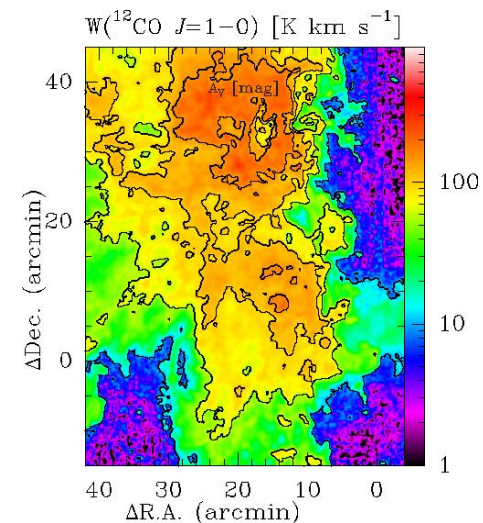
Observations of Orion B

Orion B GMC as a local template for interpreting Galactic and extragalactic molecular line observations.

One square degree map across the full 3mm band (84.5 to 115.5 GHz) with angular resolution: $31'' \sim 60$ mpc



$^{12}\text{CO } J=1-0$	$^{12}\text{CN } N=1-0$	$^{12}\text{CS } J=2-1$
$^{13}\text{CO } J=1-0$	$\text{C}^{17}\text{O } J=1-0$	$\text{C}^{18}\text{O } J=1-0$
$\text{HCO}^+ J=1-0$	$\text{HCN } J=1-0$	$\text{HNC } J=1-0$
$\text{H}^{13}\text{CO}^+ J=1-0$	$\text{N}_2\text{H}^+ J=1-0$	$\text{CH}_3\text{OH } J=2-1$
$\text{c-C}_3\text{H}_2 J=1-0$	$\text{CCH } J=1-0$	$^{32}\text{SO } J=3-2$



Multidimensional data: Principal Component Analysis

We are interested in:

Is there any relation between the features (12D data)?
Clusterization of the data in a lower dimensions?

Principal Component Analysis (PCA):

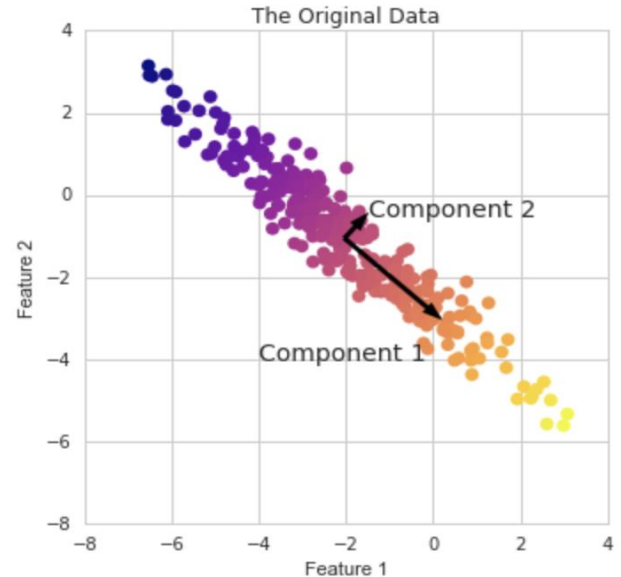
find **linear** relations between the features

Mathematically: find new orthogonal basis with maximum variance of the data along them

Technically: singular value decomposition of the data

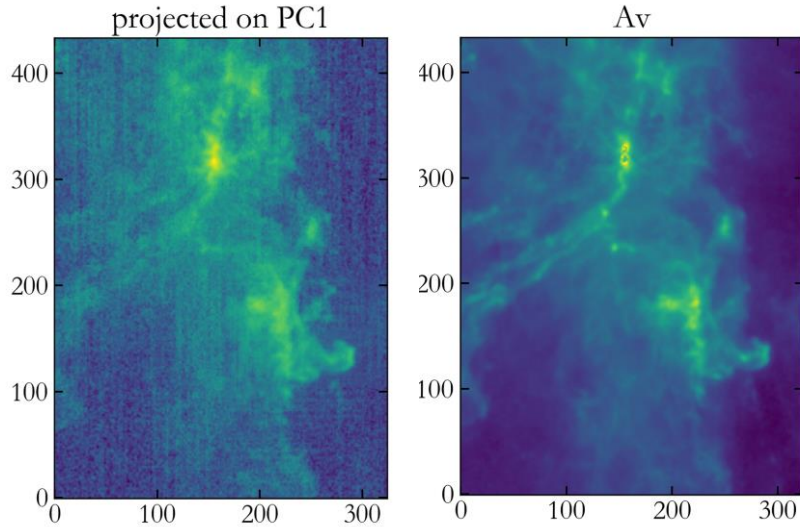
Python: scikit-learn (Pedregosa *et al.*, JMLR 12, pp. 2825-2830, 2011)

2D example

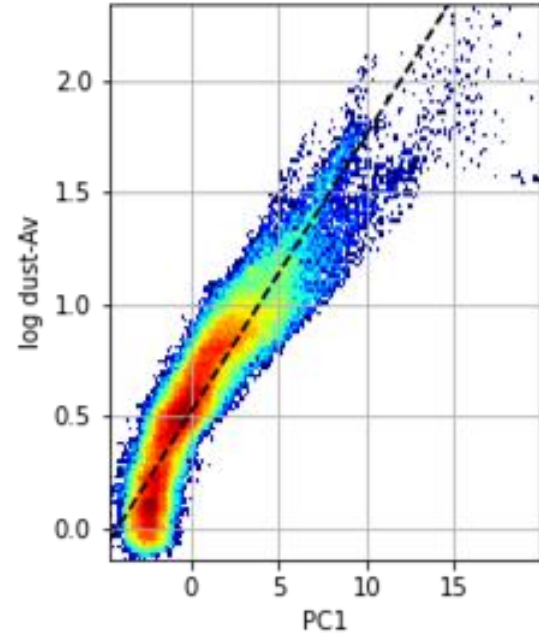


PCA of 12 rotation lines intensities

Data projected to the first Principal Component reflect visual extinction

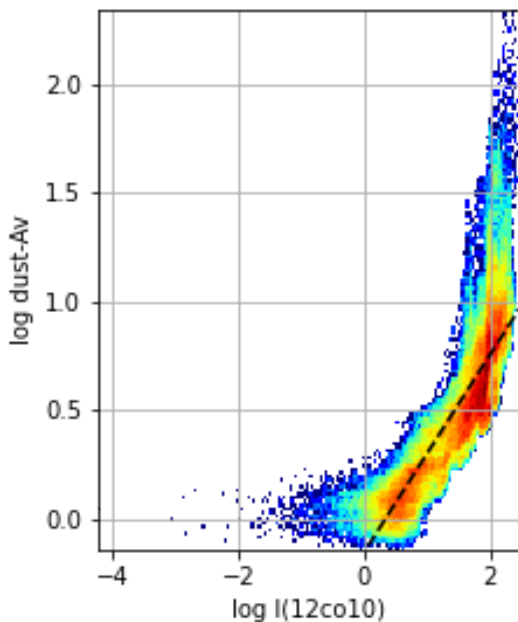


Correlations between data projected of the PCs and Av

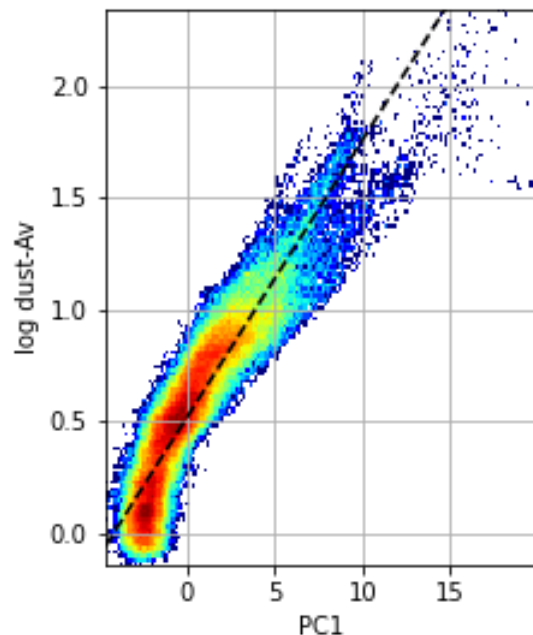


Methods comparison

with X_{CO} method
 $12\text{co}10, r^2 = 70.6\%$

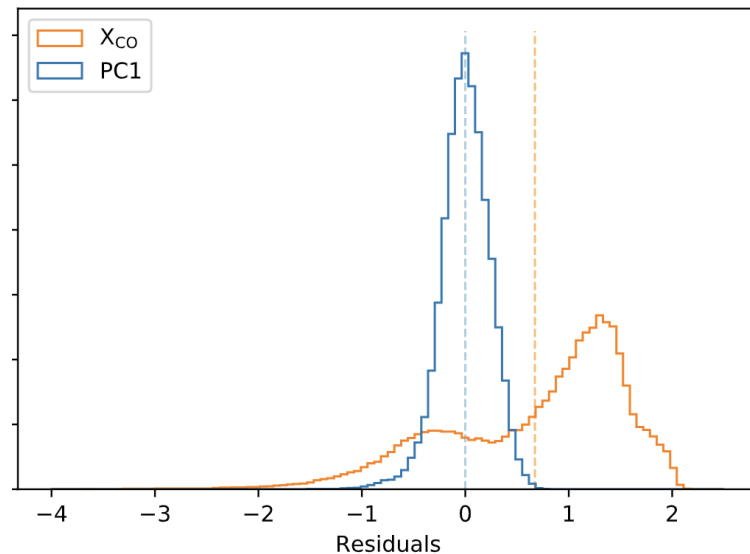


with PC1
 $\text{PC1}, r^2 = 83.3\%$



	bias	std error
X_{CO}	0.67	0.89
PC1	0.00	0.24

comparison of errors on
 $\log \text{dust-Av}$ prediction



Summary

Conclusions

- 12 line intensities from Orion B GMC
- PCA yields better results
- PCs 1, 2, 3: column density, bulk density, FUV radiation

Perspectives

- Interpret the higher principal components
- Non-linearities: random forests, other techniques
- Identify most significant lines

