





















Interpreting column densities

Strategy 1:

Model trap and equation of state

- calculate column density

- compare with image

Strategy 2:

Model trap and extract EOS independent quantities - virial theorem

Strategy 3:

Model trap and extract EOS

required signal to noise

Hydrostatics

Assume locally Homogeneous

 $F(z+\delta z)=P(z+\delta z) imes A$



$$F(z) - F(z + \delta z) = rac{dV}{dz}
ho A \, \delta z$$

gives
 $rac{dP}{dz} = -
ho rac{dV}{dz}$
Isothermal Assumption:
 $rac{dP}{dz} =
ho rac{d\mu}{d\mu}$

Result: Thomas-Fermi

dz

$$\mu(z)=\mu_0-V(z)$$

dz

Punchline:

local properties of inhomogeneous system given by homogeneous equation of state





$$\begin{array}{l} & \mbox{Extracting density} \\ & \mbox{from column density} \\ & \mbox{(not practical -- signal to noise)} \\ & dP = nd\mu + sdT \\ & \mbox{(z)} = \mu_0 - V(z) \end{array} \qquad \mbox{Assume Harmonic trap} \\ \hline & \mbox{(a(z))} = \int d^2r_{\perp}n(\mu_c(z) - V_{\perp}(r_{\perp})) \\ & = \frac{2\pi w^4 m}{\hbar^2 \lambda^2} \int_{-\infty}^{\mu_c(z)} d\mu[\lambda^2 n(\mu)] \\ & = \frac{2\pi w^4 m}{\hbar^2 \lambda^2} P_c(z), \end{array} \qquad \begin{array}{l} \frac{\partial n_a(z)}{\partial z} = \int d^2r_{\perp} \frac{\partial n(\mu_c(z) - V_{\perp}(r_{\perp}))}{\partial \mu} \left(-\frac{2\hbar^2 z}{md^4}\right) \\ & = -\frac{8\pi z w^4}{d^4} \int_{-\infty}^{\mu_c(z)} \frac{\partial n}{\partial \mu} \\ & = -\frac{8\pi z w^4 n_c(z)}{d^4 \lambda^2}, \end{array} \\ \hline & \mbox{Integrate to get "Axial Pensity"} \qquad \end{tabular}$$





Idea: Model Noise **Model Space Data Space** (ex 3D) Data K $\{f_i\}$ $\{F_{\nu}\}$ $\{F_{\nu}^{d}\}$ Model noise $F_{\nu}^{d} = \bar{F}_{\nu}^{d} + \sigma_{\nu}\xi_{\nu}$ Compare model and data $\chi^{2} = \sum_{\nu} \frac{(F_{\nu} - F_{\nu}^{d})^{2}}{\sigma^{2}}$ Ex: Independent Gaussian random variables with mean 0 and Want to choose f standard deviation 1 $F_{\nu} = \bar{F}_{\nu}^{d}$ so that

Idea: Model
$$\chi^{2} = \sum_{\nu} \frac{(F_{\nu} - F_{\nu}^{d})^{2}}{\sigma^{2}}$$
$$\text{If } F_{\nu} = \overline{F}_{\nu}^{d}$$
Most probable value of χ^{2}
$$\overline{\chi^{2}} = N$$
$$= number of pixels$$
Consider space of all f's which give this chi 2

Data

 $\{F^d_\nu\}$

Model noise

Noise

 $F_{\nu}^{d} = \bar{F}_{\nu}^{d} + \sigma_{\nu}\xi_{\nu}$

Ex: Independent Gaussian random variables with mean 0 and standard deviation 1

Maximum EntropyBayesian Principle:If you don't know anything, assume
everything is equally likelyChoose the f's which satisfy
$$\chi^2 = N$$

and carry the least informationMaximize $S = -\sum_i f_i \log(f_i/M)$ $M = \sum_i f_i$