

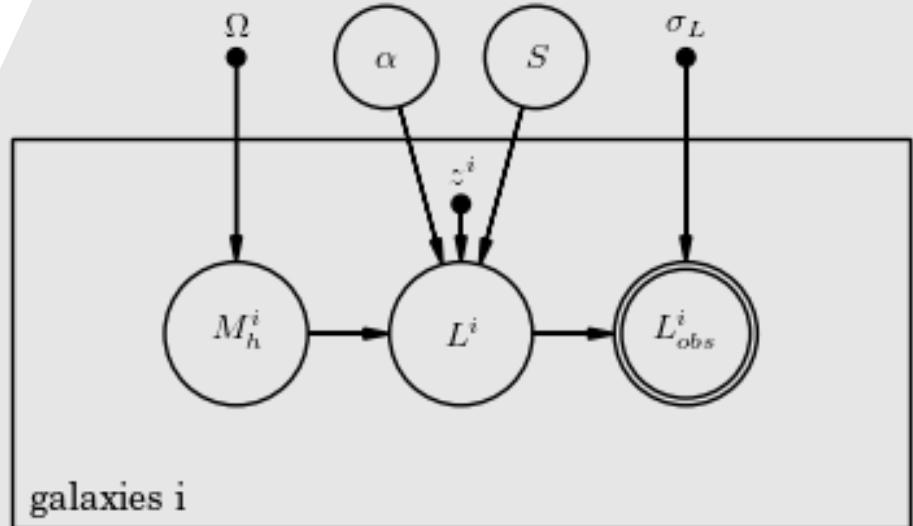
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Linking Halo Mass to Galaxy Luminosity



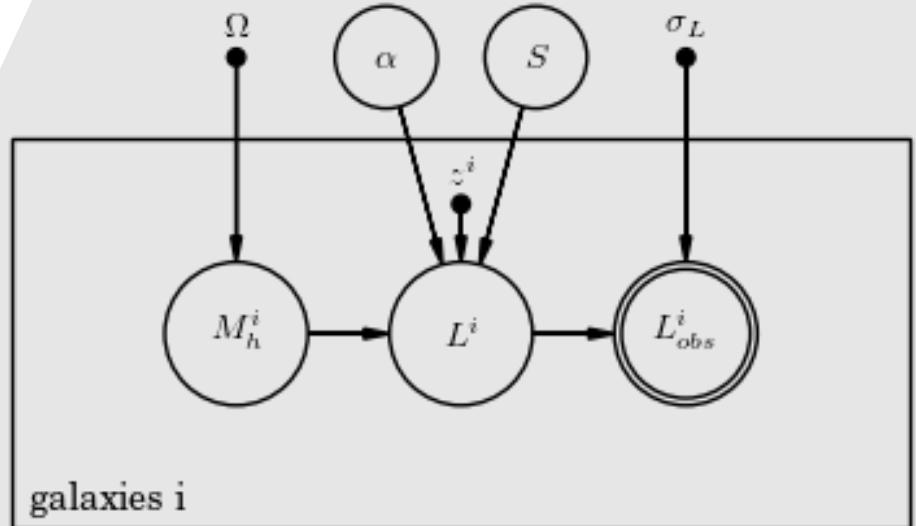
Statistical Model

$$P(\mathbf{M}_h) \propto \frac{dn}{dM} = f(\sigma) \frac{\rho_m}{M} \frac{d\ln \sigma^{-1}}{dM}$$
$$P(\mathbf{L}|\alpha, S, \mathbf{z}, \mathbf{M}_h) \sim \frac{1}{\sqrt{2\pi S^2 \mathbf{L}^2}} \exp\left(-\frac{(\ln \mathbf{L} - \ln \mathbf{u}_L)^2}{2S^2}\right);$$
$$\ln \mathbf{u}_L = \alpha_1 + \alpha_2 \ln\left(\frac{\mathbf{M}_h}{\alpha_3}\right) + \alpha_4 \ln(1 + \mathbf{z})$$
$$P(\mathbf{L}_{obs}|\mathbf{L}) \sim \frac{1}{\sqrt{2\pi \sigma_L^2 \mathbf{L}_{obs}^2}} \exp\left(-\frac{(\ln \mathbf{L}_{obs} - \ln \mathbf{L})^2}{2\sigma_L^2}\right)$$
$$P(\alpha, S) \sim \text{Uniform}$$



Statistical Inference

$$\underbrace{P(\alpha, S | \mathbf{L}_{obs}, \mathbf{z})}_{posterior} = \frac{P(\mathbf{z}) P(\alpha, S) P(\mathbf{L}_{obs} | \alpha, S, \mathbf{z})}{P(\mathbf{L}_{obs}, \mathbf{z})}; \text{ (Bayes Theorem)}$$
$$\propto P(\alpha, S) [P(\mathbf{L}_{obs} | \alpha, S, \mathbf{z})]; \text{ (Normalization)}$$
$$\propto \underbrace{P(\alpha, S)}_{prior} \underbrace{\int \int P(\mathbf{L}_{obs} | \mathbf{L}) P(\mathbf{L} | \alpha, S, \mathbf{z}, \mathbf{M}_h) P(\mathbf{M}_h) d\mathbf{M}_h d\mathbf{L}}_{likelihood}$$



Importance Sampling

- Approximate sampling a distribution by sampling from a more convenient distribution in a ‘fair’ way.

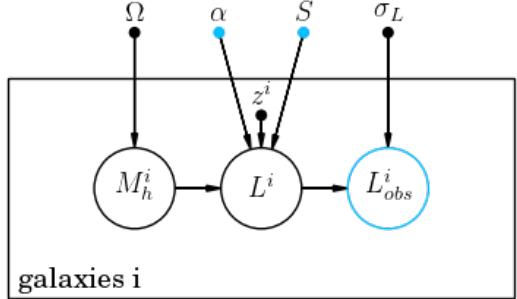
$$\begin{aligned}\mathcal{L} &= \int \int P(\mathbf{L}_{obs}|\mathbf{L})P(\mathbf{L}|\alpha, S, \mathbf{z}, \mathbf{M}_h)P(\mathbf{M}_h)d\mathbf{M}_hd\mathbf{L} \\ &= \int \int \frac{P(\mathbf{L}_{obs}|\mathbf{L})P(\mathbf{L}|\alpha, S, \mathbf{z}, \mathbf{M}_h)P(\mathbf{M}_h)}{Q(\mathbf{M}_h, \mathbf{L})}Q(\mathbf{M}_h, \mathbf{L})d\mathbf{M}_hd\mathbf{L}; \text{ (Insert Q/Q)} \\ &\approx \frac{1}{N_s} \sum_{s \sim Q} \frac{P(\mathbf{L}_{obs}|\mathbf{L}_s)P(\mathbf{L}_s|\alpha, S, \mathbf{z}, \mathbf{M}_{hs})P(\mathbf{M}_{hs})}{Q(\mathbf{M}_{hs}, \mathbf{L}_s)}; \text{ (Discretize)}\end{aligned}$$

- Go from iterating through $[\min(M_h), \max(M_h)]^{100,000} \times [\min(L), \max(L)]^{100,000}$ to sampling from $\hat{P}(\mathbf{M}_h, \mathbf{L})$

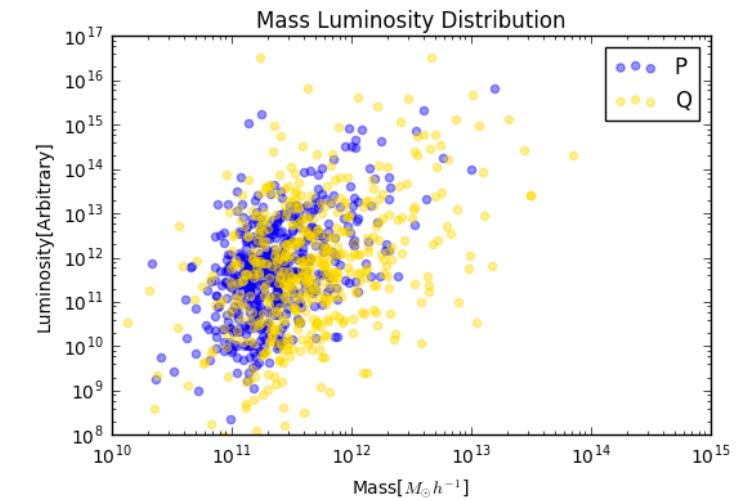
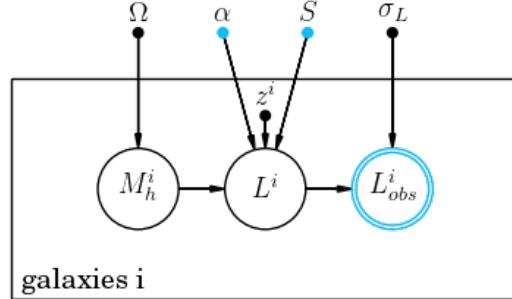
Generating
Q Similar To
P

$$\begin{aligned} Q(\mathbf{M}_h, \mathbf{L}) &= \hat{P}(\mathbf{M}_h, \mathbf{L}) \\ &= \int \hat{P}(\mathbf{M}_h | \mathbf{L}) \hat{P}(\mathbf{L} | \mathbf{L}_{obs}) d\mathbf{L}_{obs} \\ &\approx \frac{1}{N_s} \sum_{s \sim \mathbf{L}_{obs}} \hat{P}(\mathbf{M}_h | \mathbf{L}) \hat{P}(\mathbf{L} | \mathbf{L}_{obs}) \\ \hat{P}(\mathbf{M}_h | \mathbf{L}) &\sim \text{Delta} \left[\exp(\alpha'_1) \left(\frac{\mathbf{M}_h}{\alpha'_3} \right)^{\alpha'_2} (1+z)^{\alpha'_4} \right] \\ \hat{P}(\mathbf{L} | \mathbf{L}_{obs}) &\sim \frac{1}{\sqrt{2\pi\sigma_L^2 \mathbf{L}^2}} \exp \left(-\frac{(\ln \mathbf{L} - \ln \mathbf{L}_{obs})^2}{2\sigma_L^2} \right) \end{aligned}$$

Generate Data and $P(\mathbf{M}_h, \mathbf{L})$



Infer $Q(\mathbf{M}_h, \mathbf{L})$



Validating Q Against P