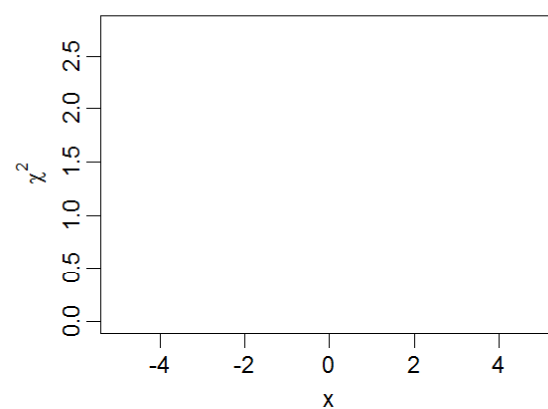
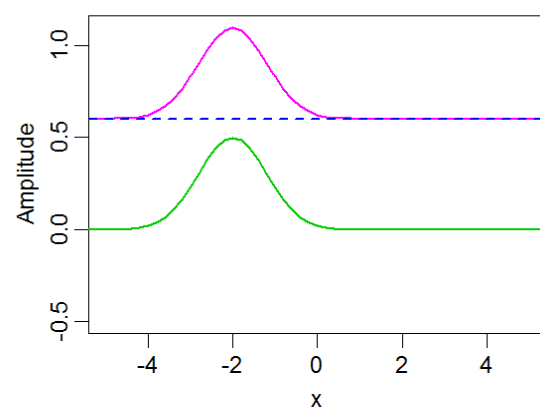


## Simple example with no noise

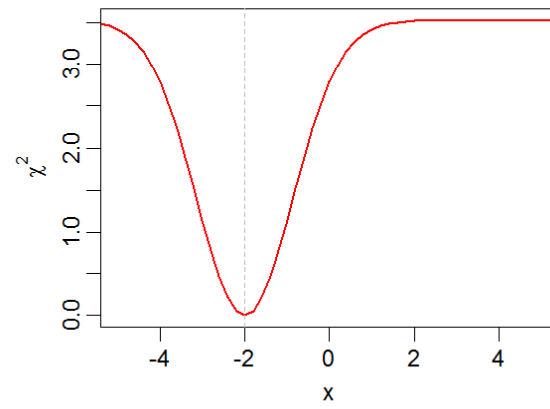
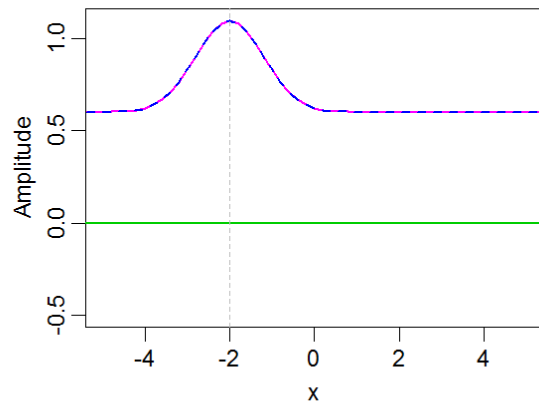
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$



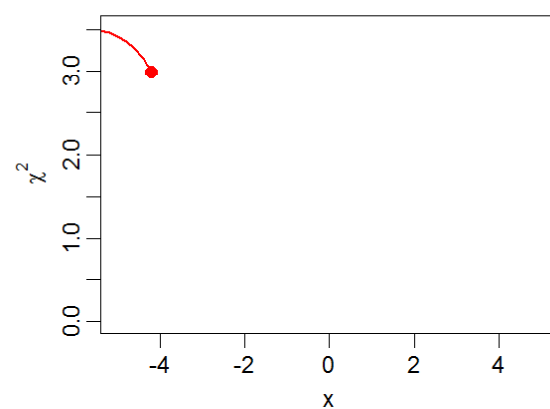
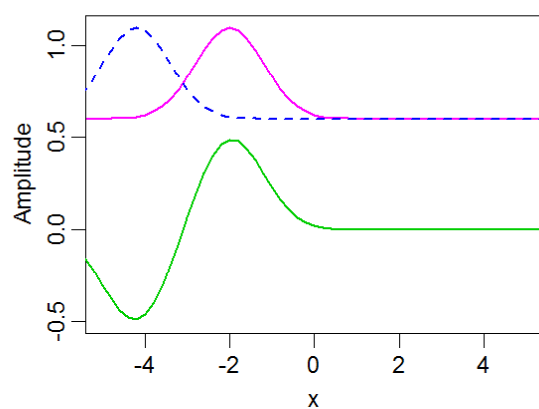
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$



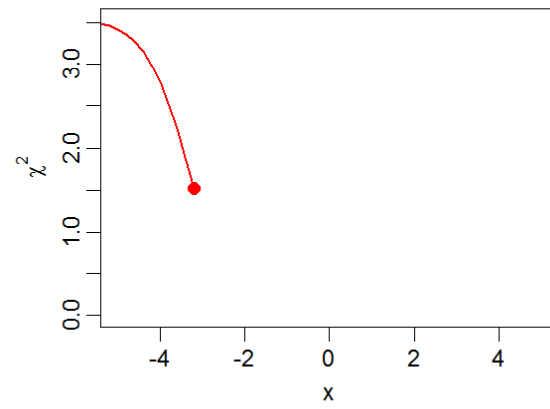
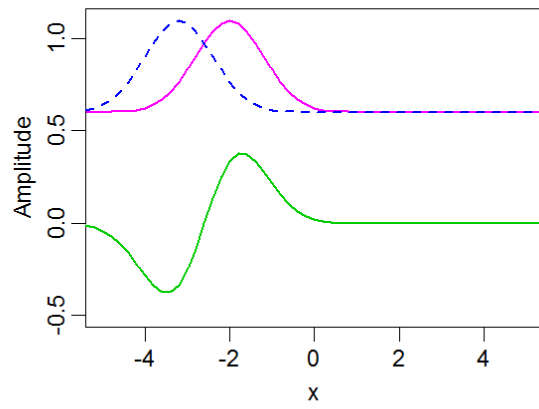
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

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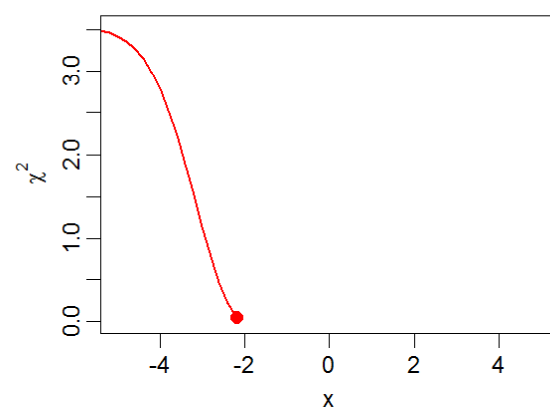
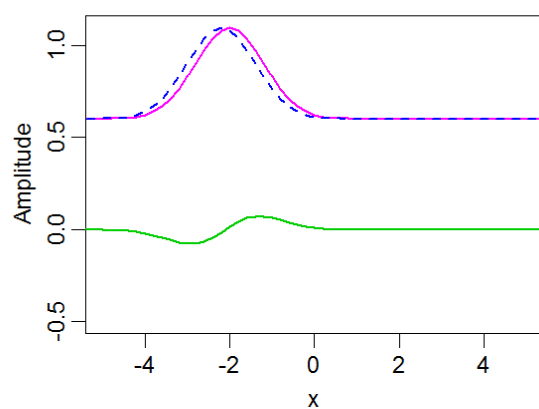


$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2$$

The full definition

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2$$

An unscaled but appealing version

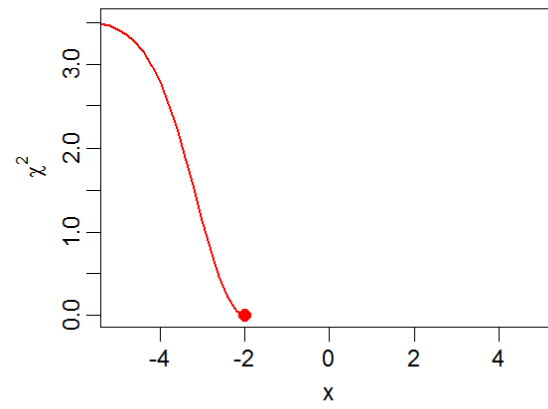
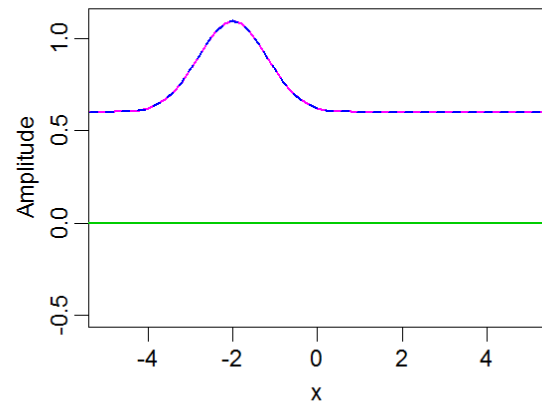


$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2$$

The full definition

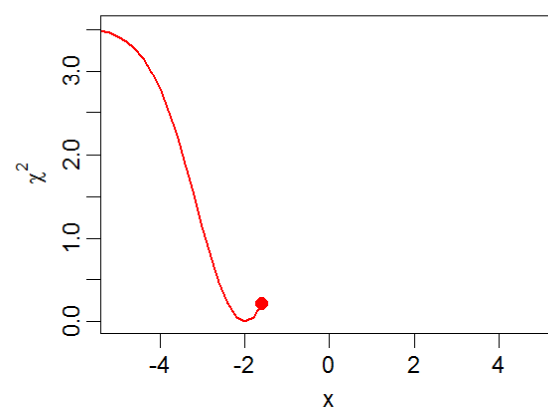
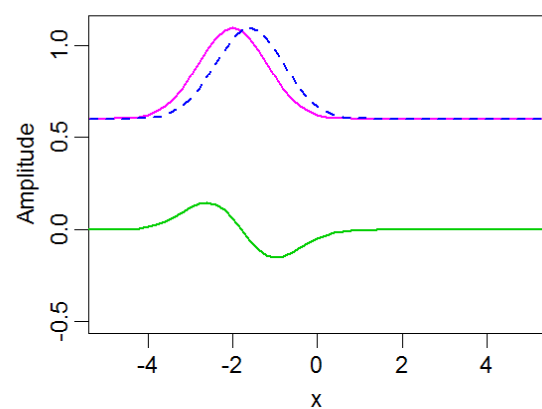
$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2$$

An unscaled but appealing version



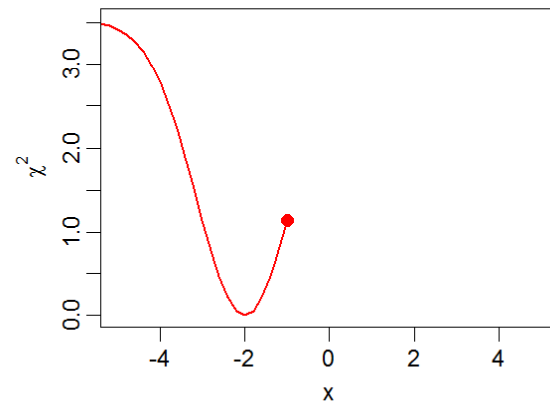
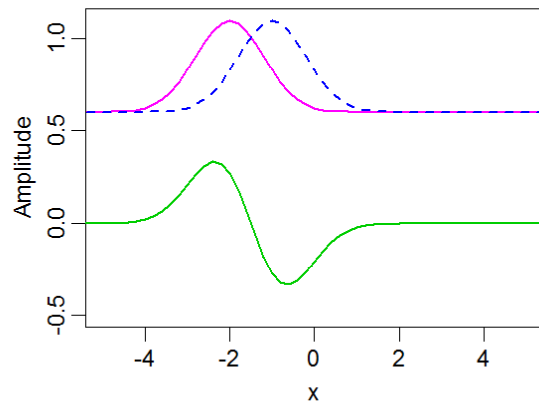
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$



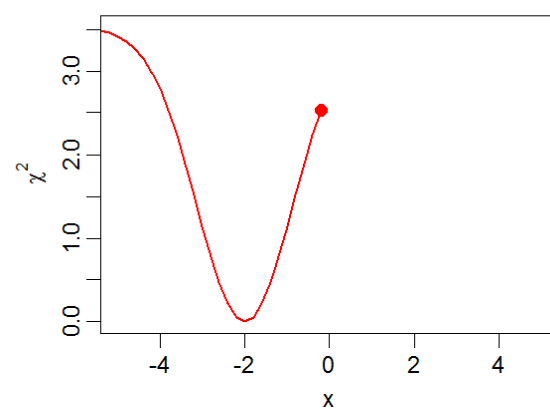
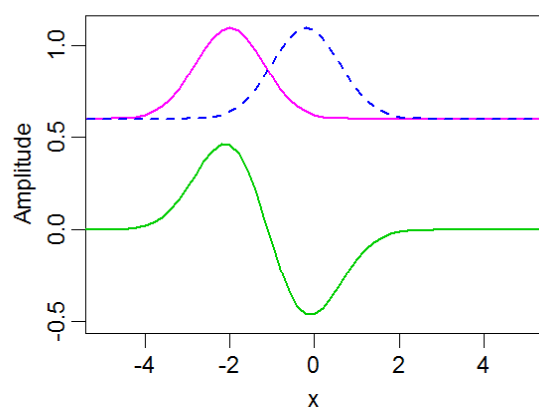
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$



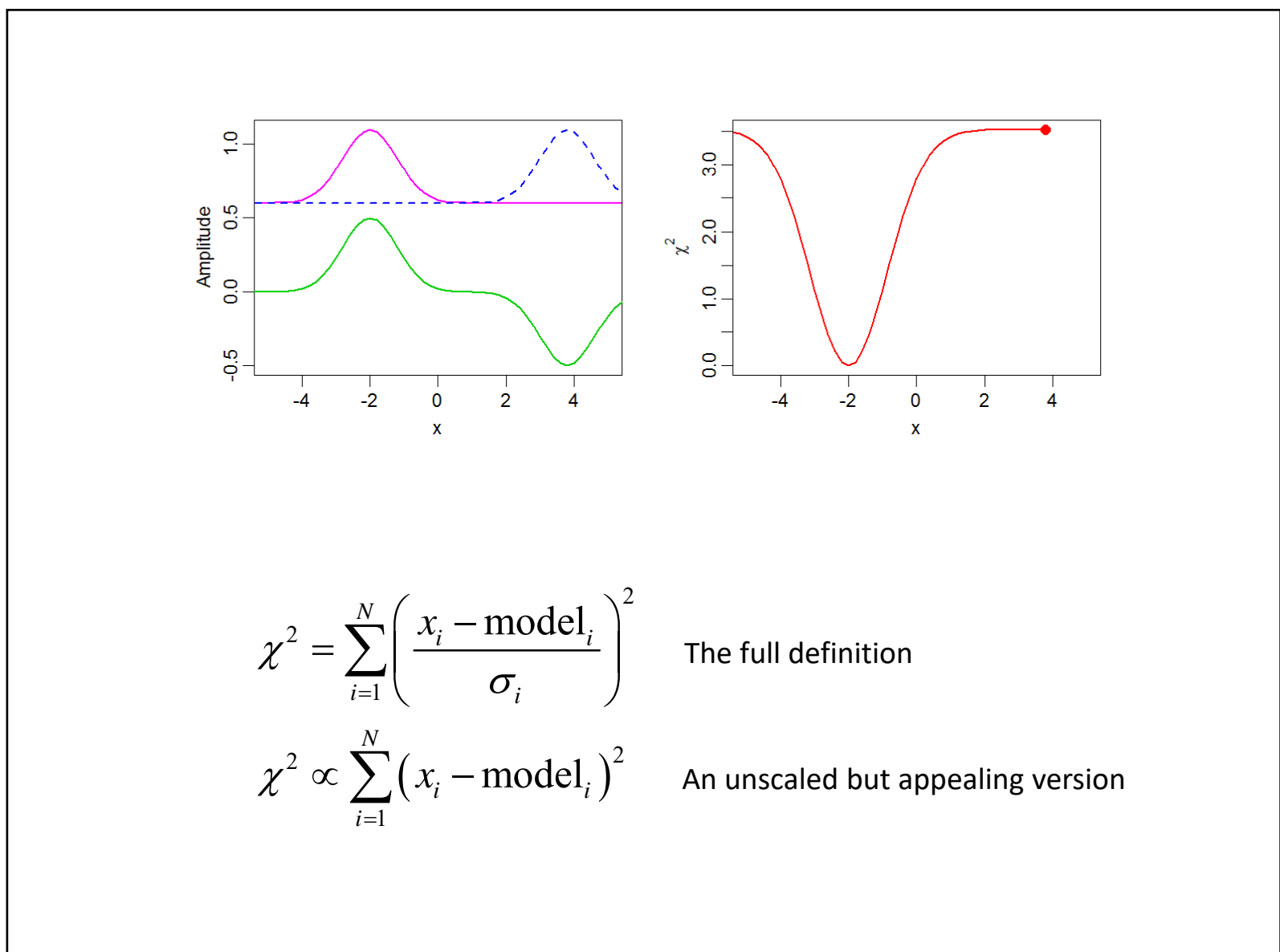
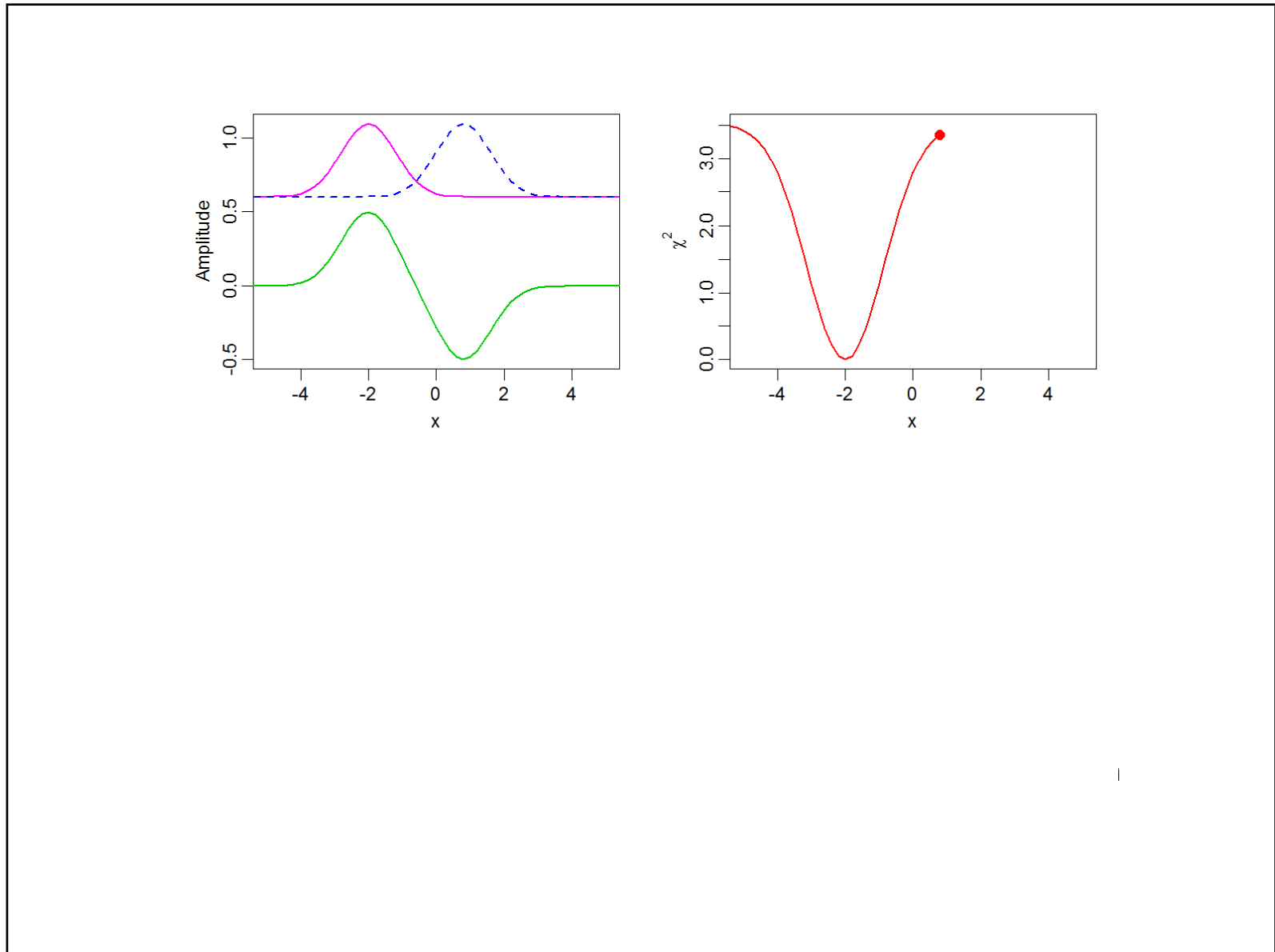
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$



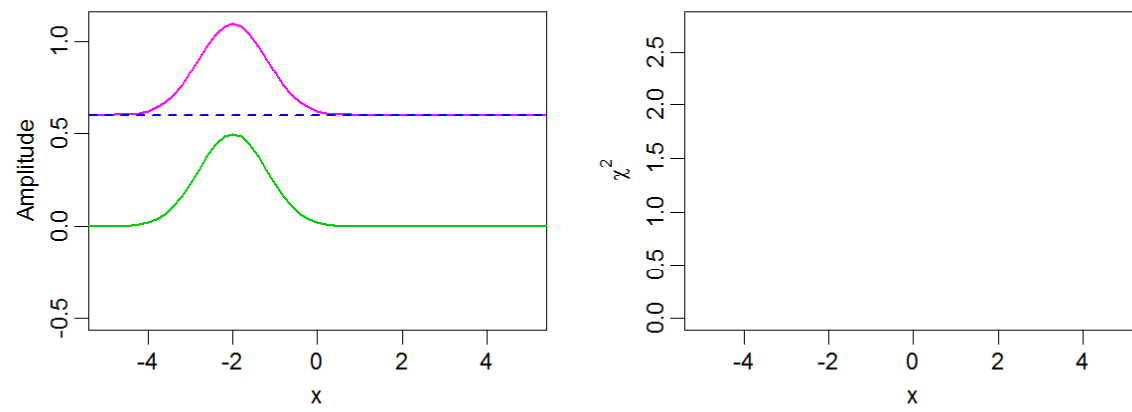
$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

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$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$

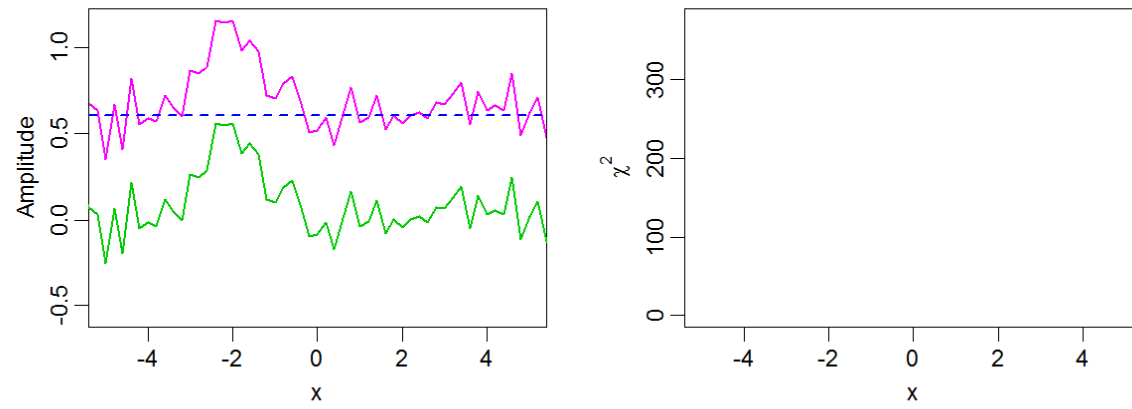


$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2 \quad \text{The full definition}$$

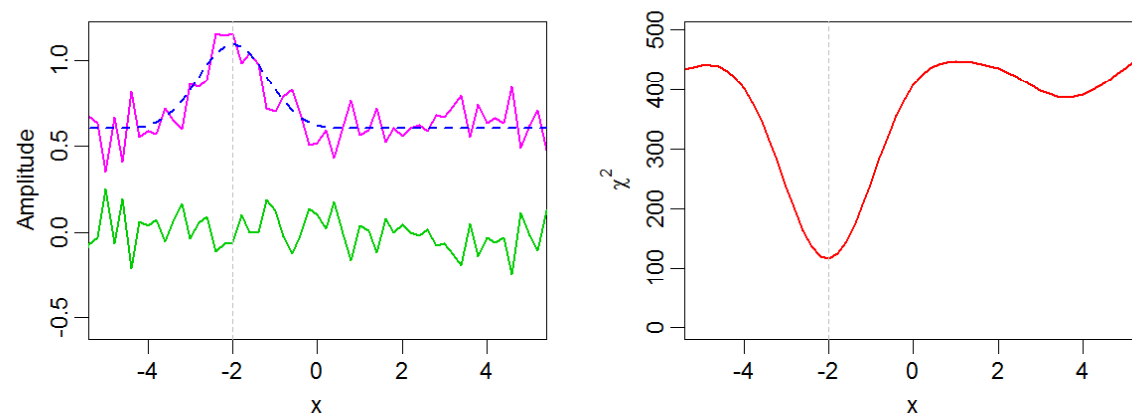
$$\chi^2 \propto \sum_{i=1}^N (x_i - \text{model}_i)^2 \quad \text{An unscaled but appealing version}$$

Add noise to calculation

$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2$$



$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2$$

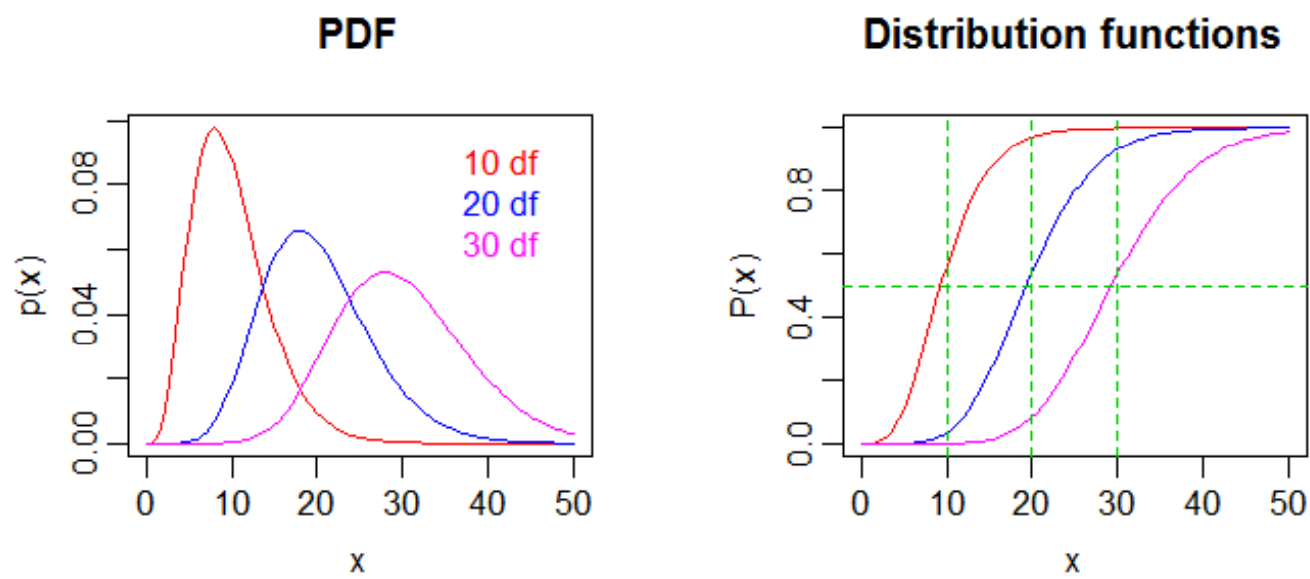


$$\chi^2 = \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2$$

Can't we get more information that just the location of a minimum?



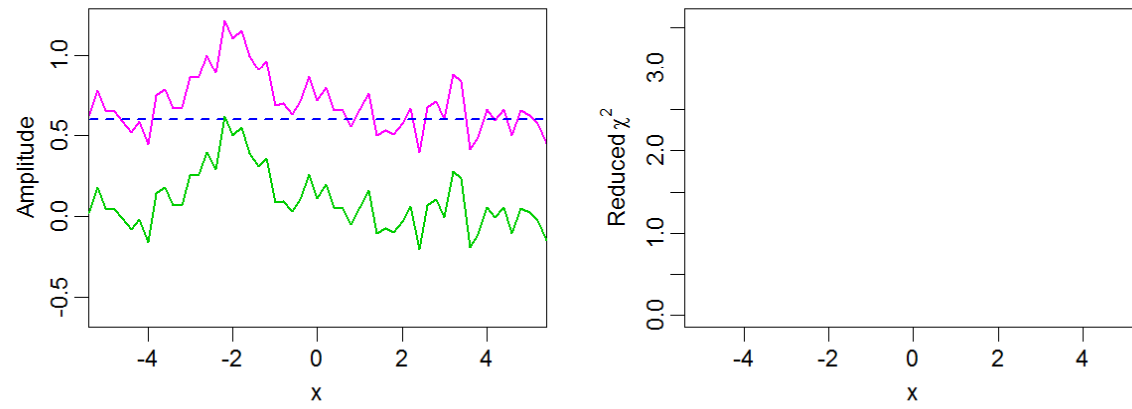
# $\chi^2$ PDF and CDF



## Reduced $\chi^2$ with noise

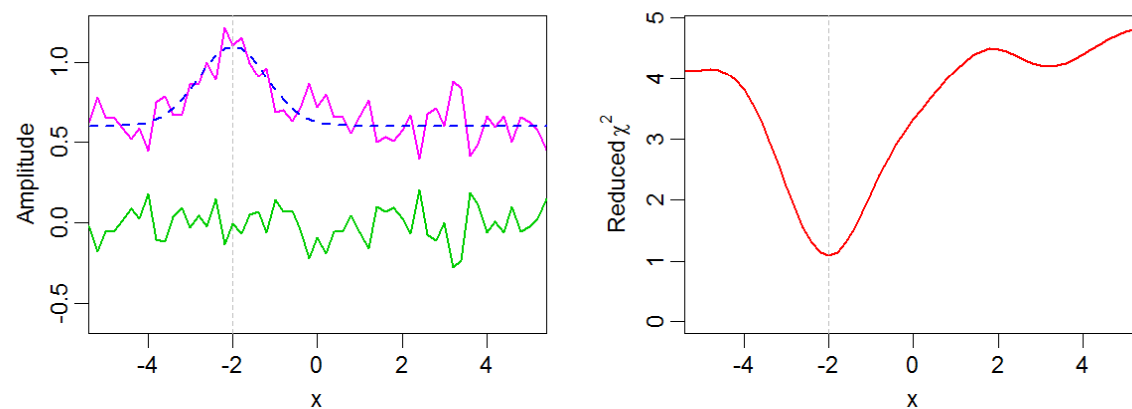
Divide by degrees of freedom for "reduced  $\chi^2$ "

$$\chi_r^2 = \frac{1}{df} \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2, \quad df \leq N$$



Divide by degrees of freedom for "reduced  $\chi^2$ "

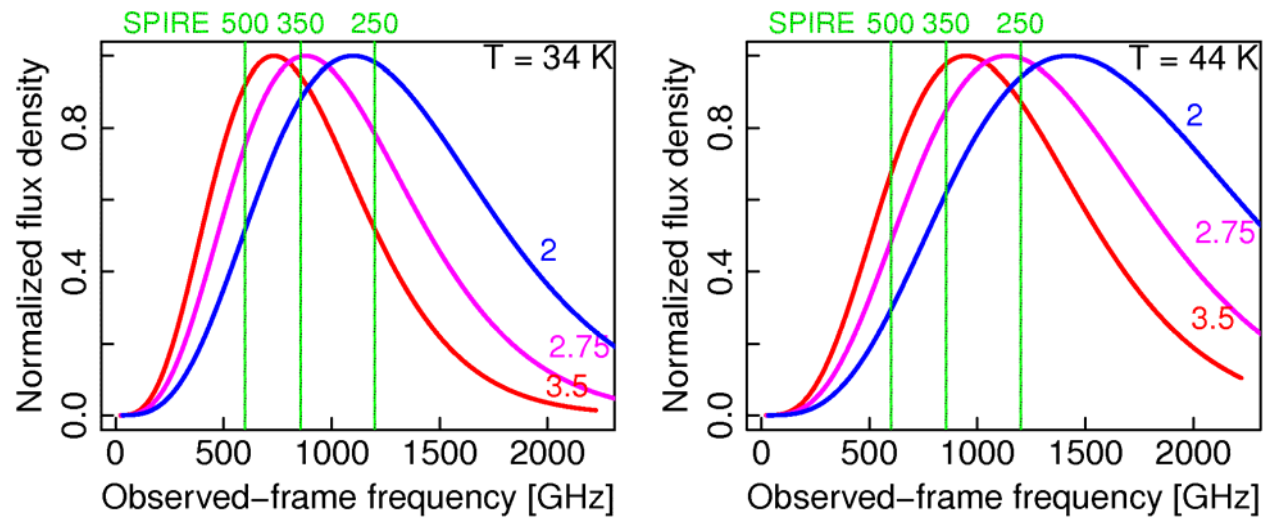
$$\chi_r^2 = \frac{1}{df} \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2, \quad df \leq N$$



Divide by degrees of freedom for "reduced  $\chi^2$ "

$$\chi_r^2 = \frac{1}{df} \sum_{i=1}^N \left( \frac{x_i - \text{model}_i}{\sigma_i} \right)^2, \quad df \leq N$$

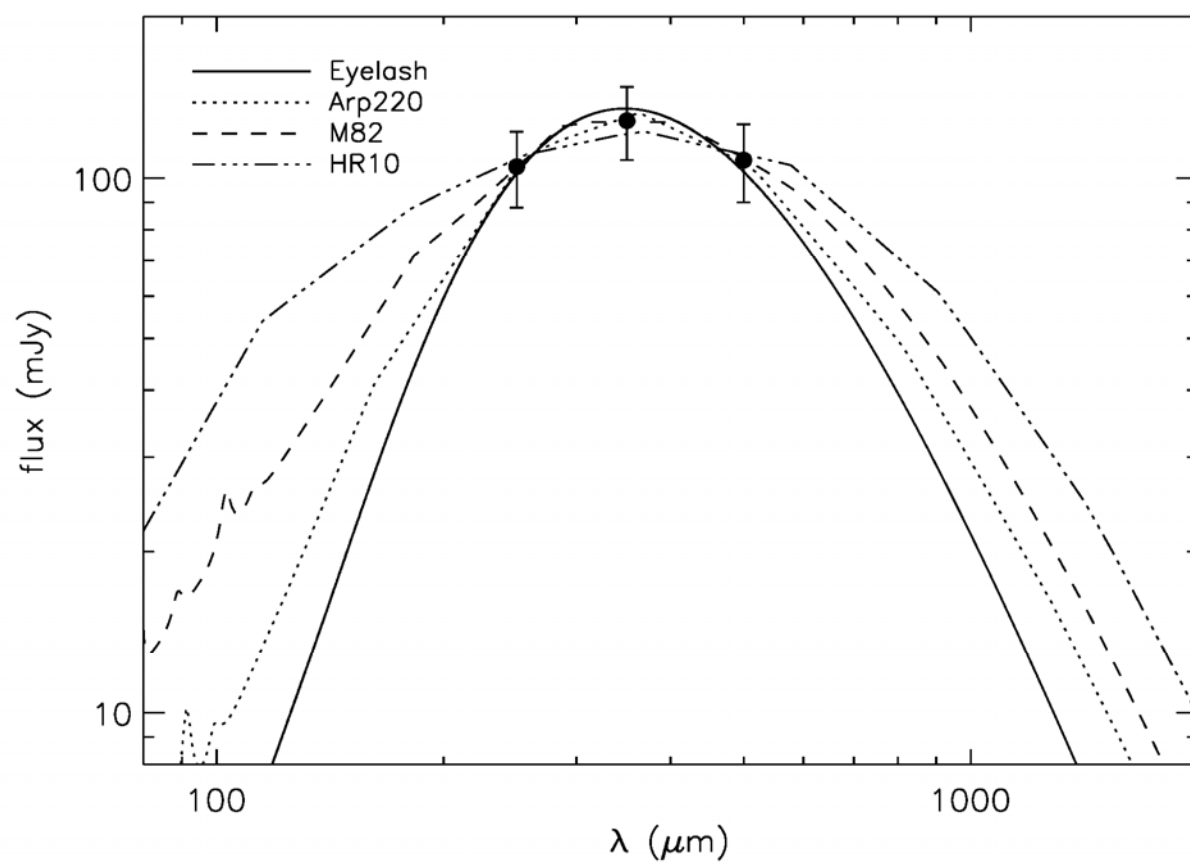
## Photometric redshifts and the T-z degeneracy

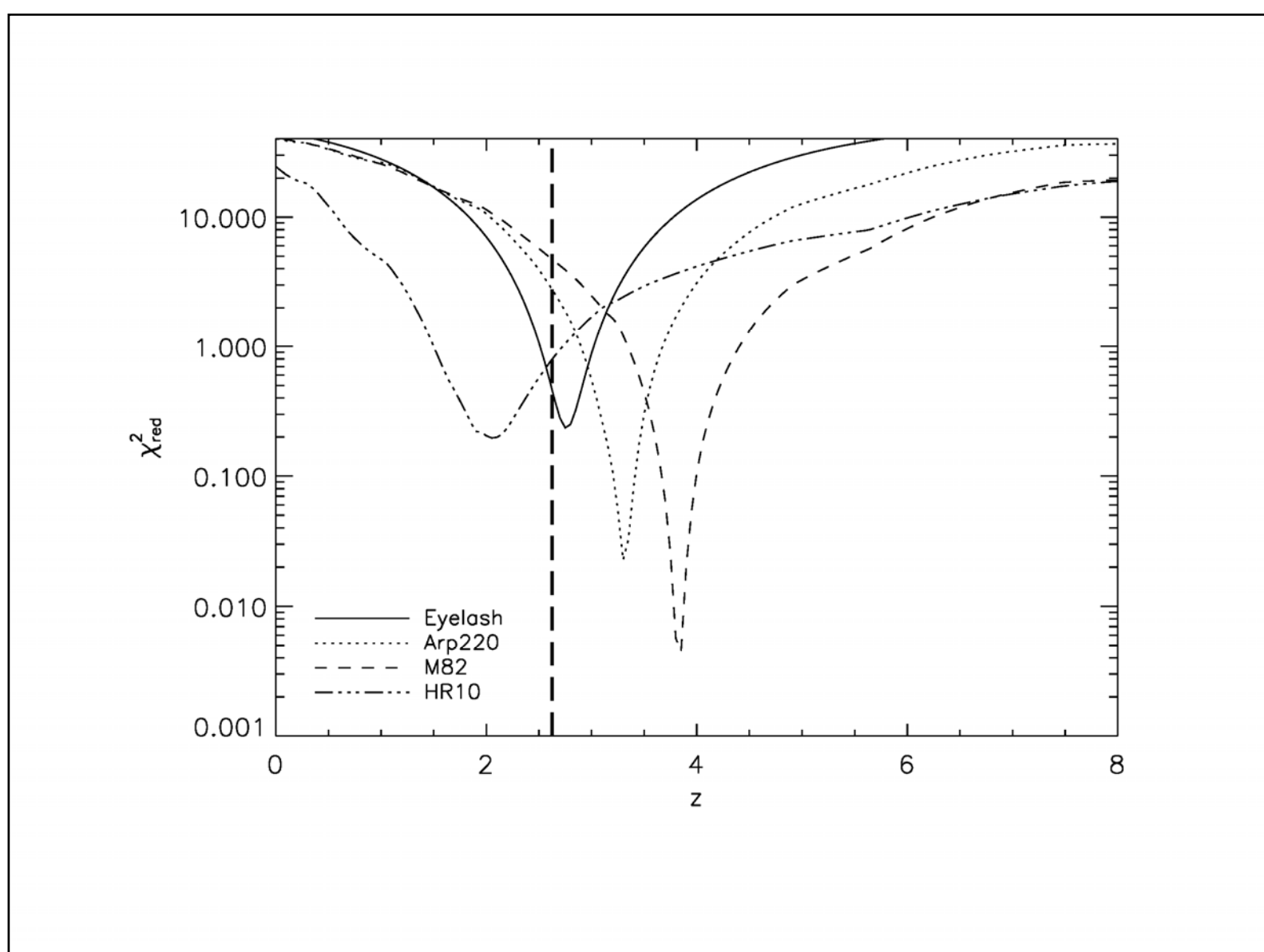
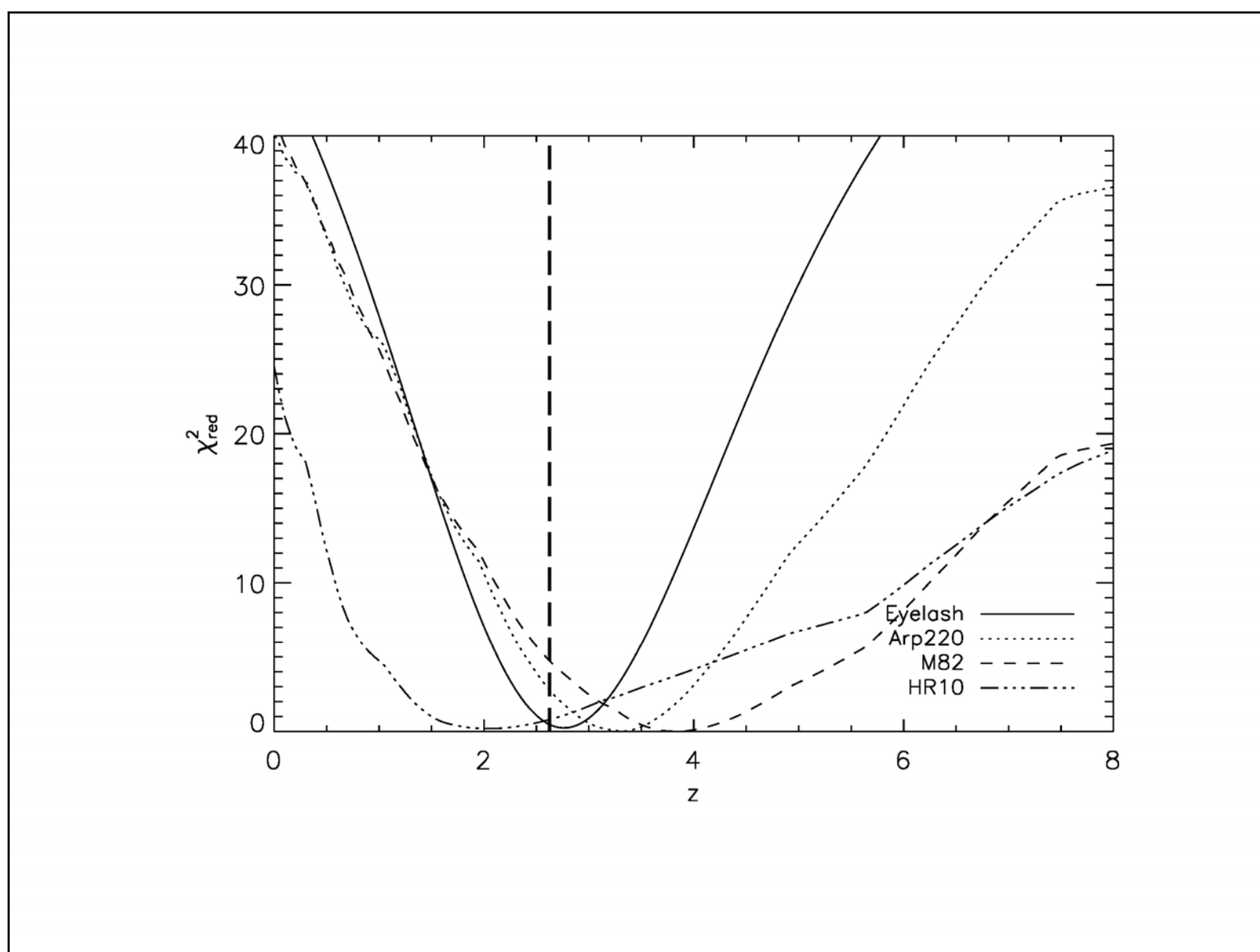


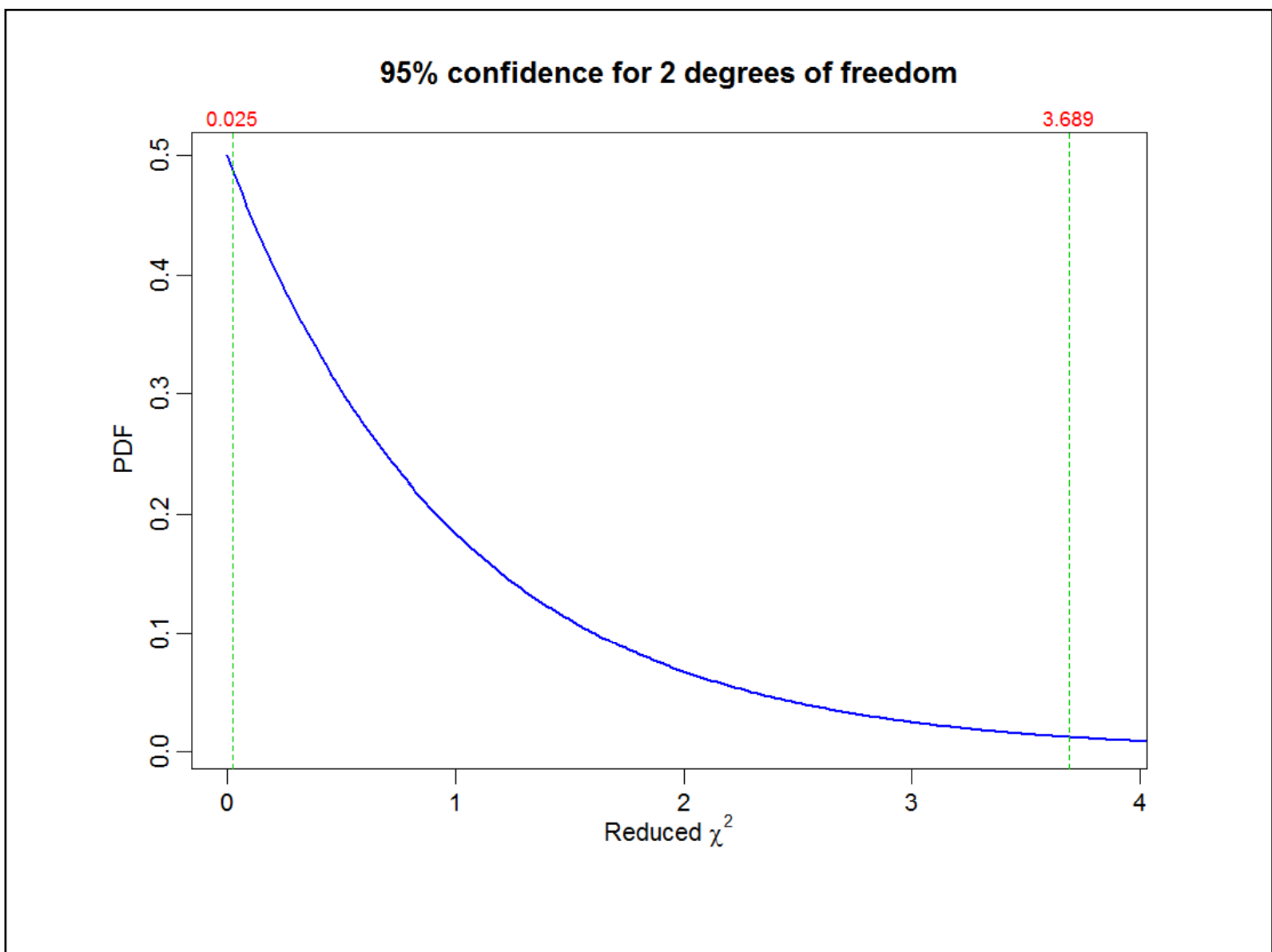
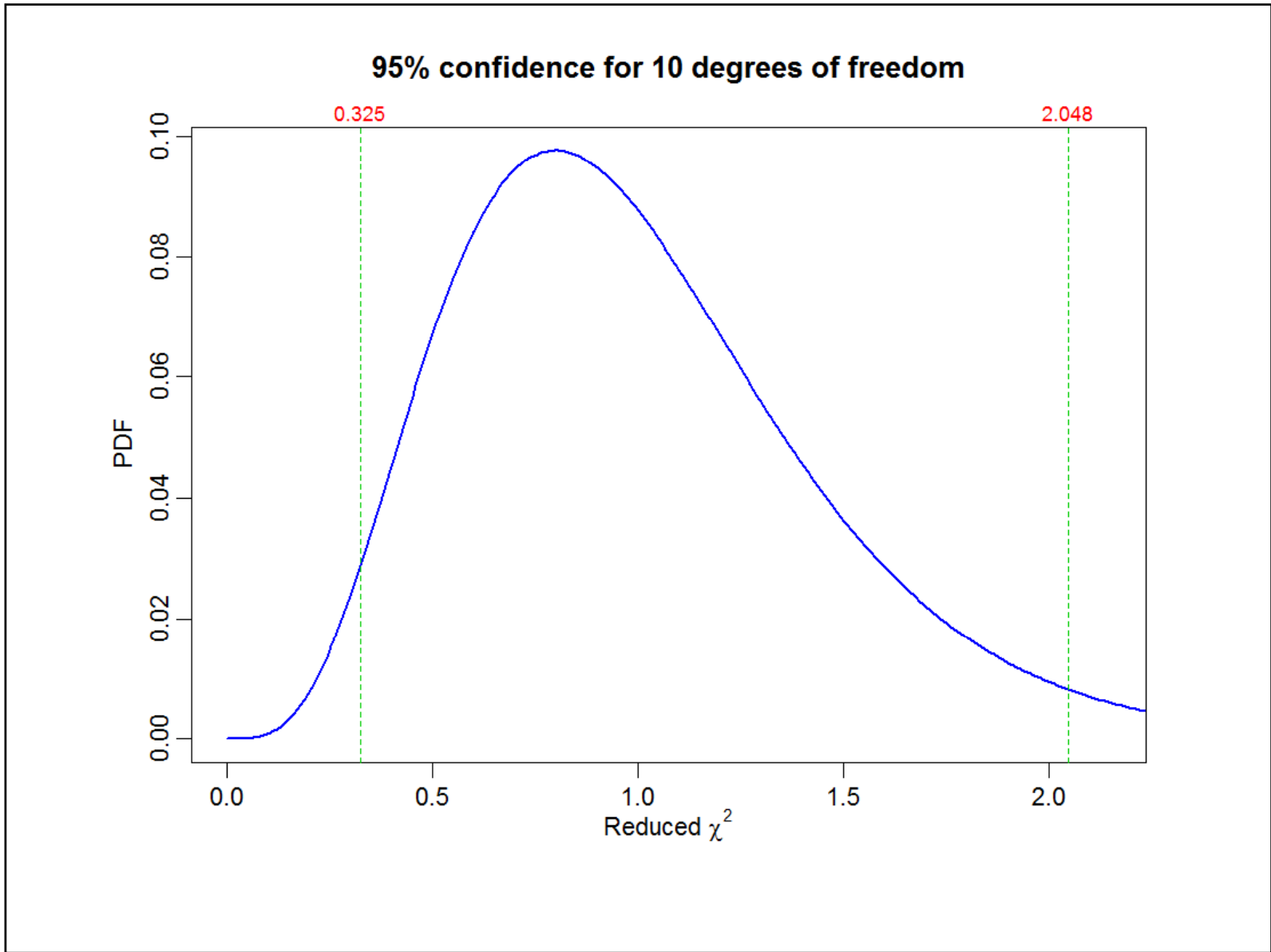
$$\lambda_{\max} T = \text{constant} \quad \text{Wien's displacement law}$$

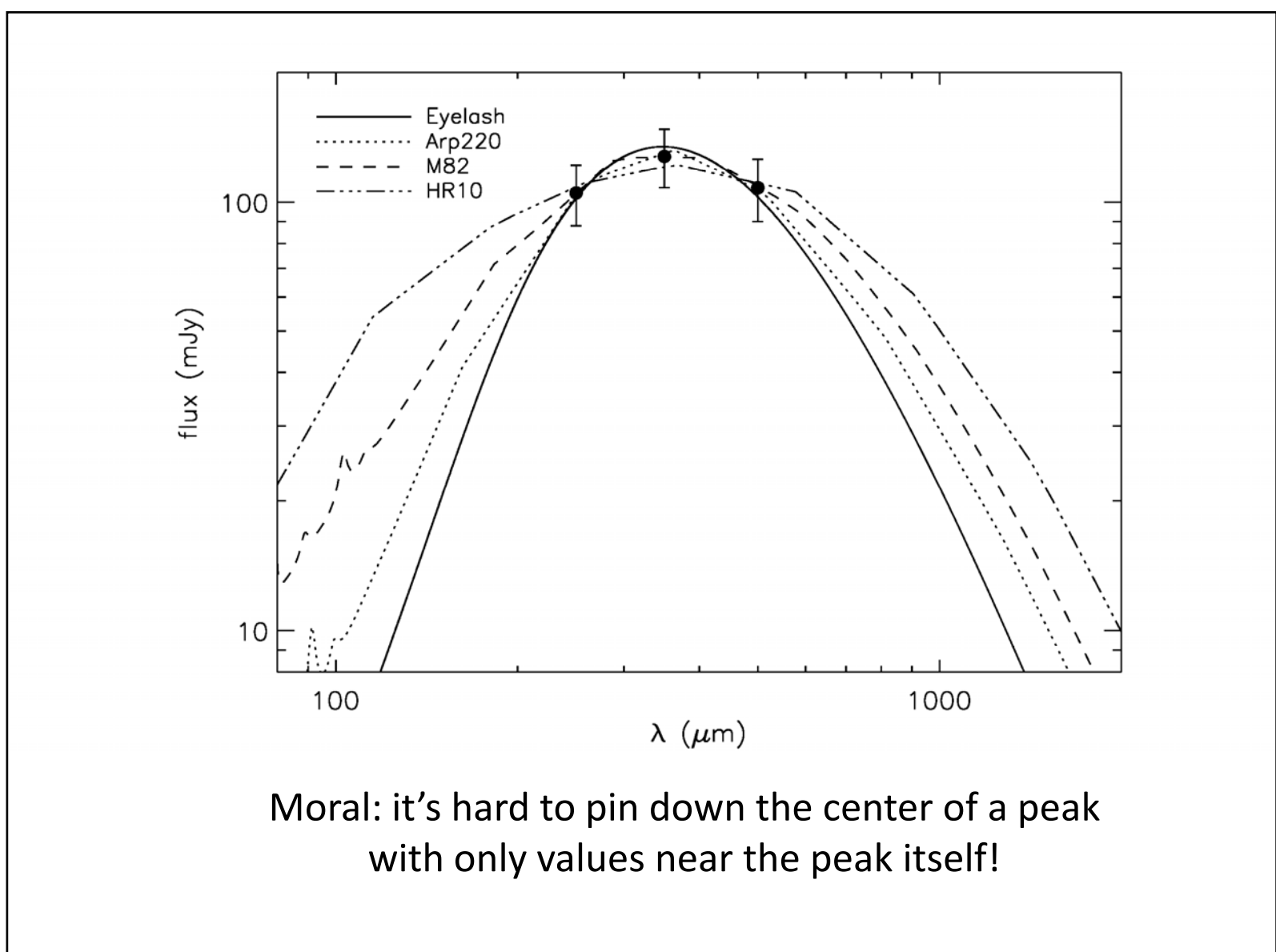
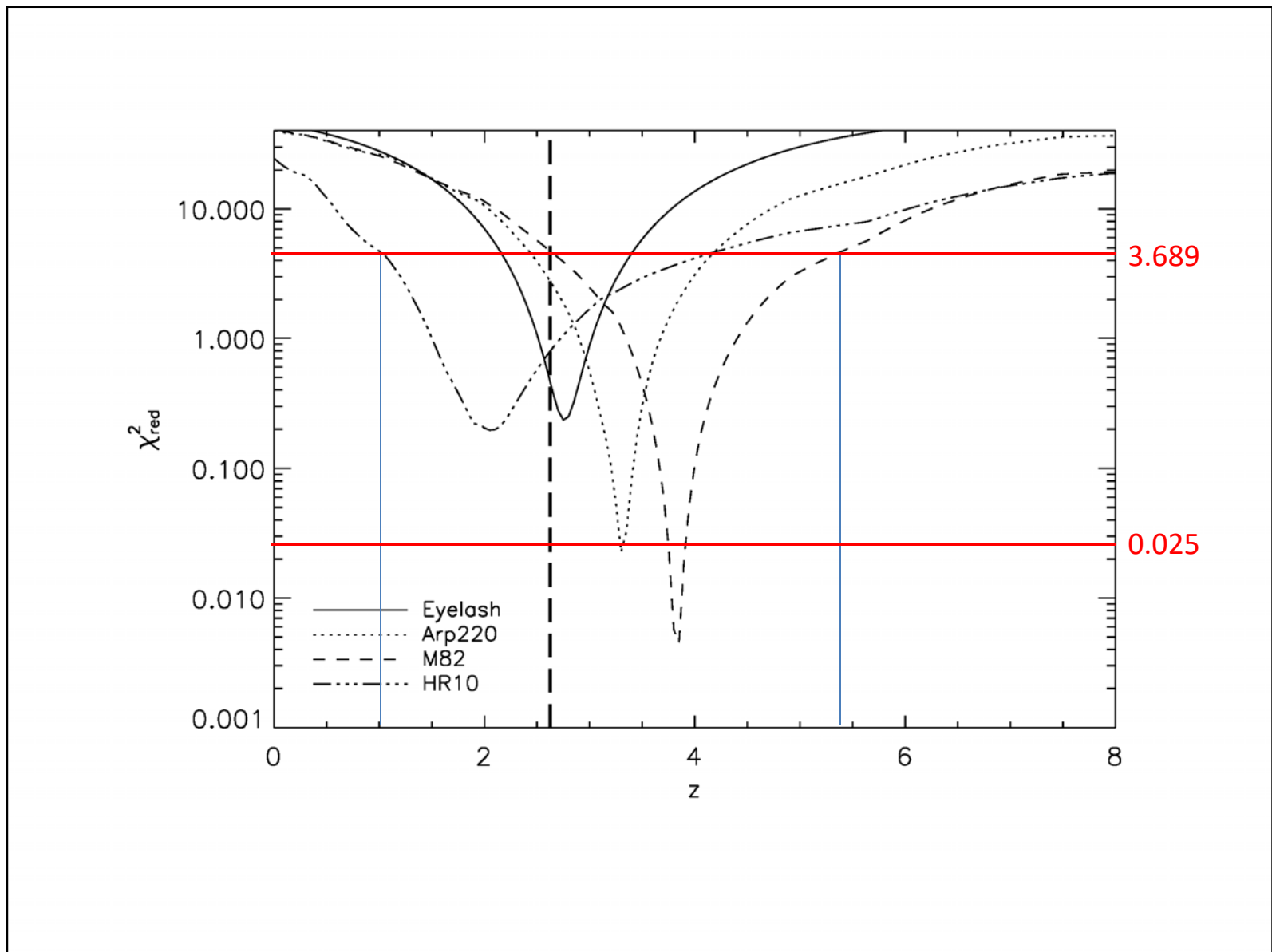
$$\lambda_{\max, \text{observed}} = \frac{\lambda_{\max, \text{rest}}}{1+z} \Rightarrow \lambda_{\max, \text{observed}} (1+z) T = \text{constant}$$

Modified for redshift

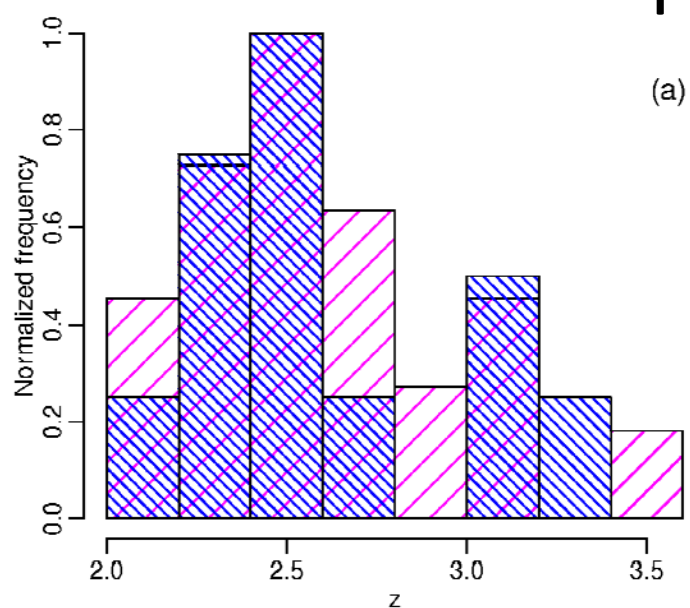








## Redshift distributions from CO 1–0 and optical observations

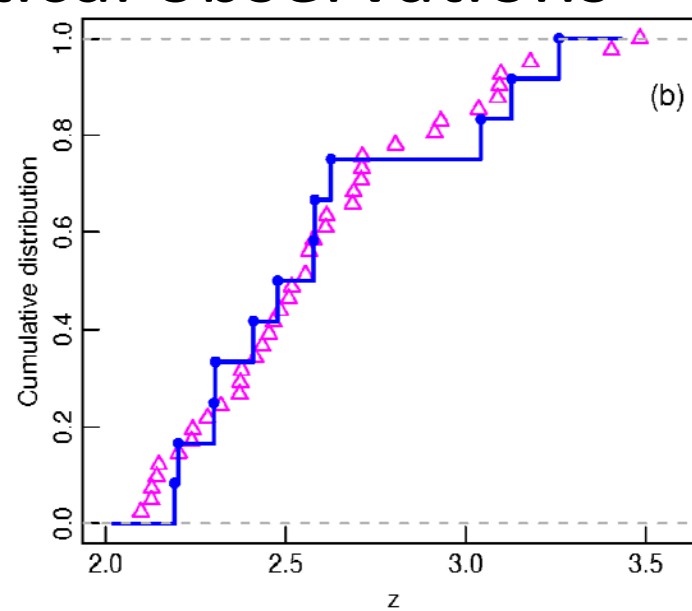
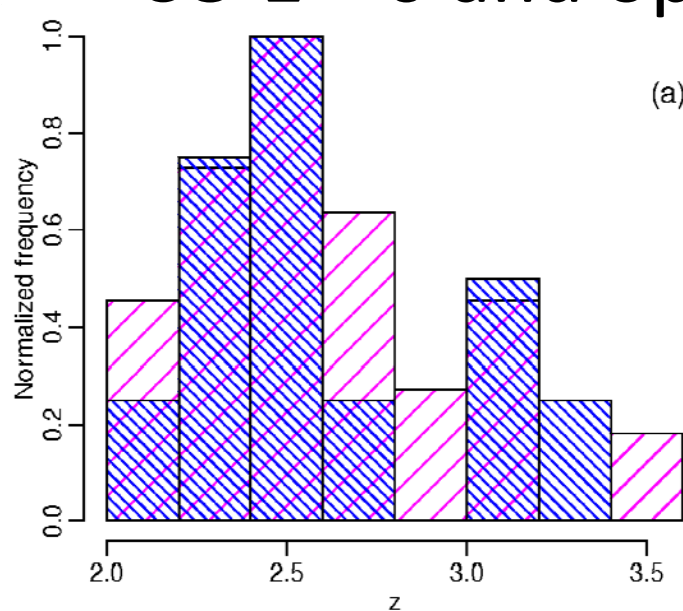


Blue: CO, Harris et al. 2012

Magenta: Optical, Chapman et al. 2005

Harris et al. 2012

## Redshift distributions from CO 1–0 and optical observations



Blue: CO, Harris et al. 2012

Magenta: Optical, Chapman et al. 2005

Harris et al. 2012

# Poisson and Normal PDFs

