

# Fitting and Comparison of Models of Radio Spectra

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# Outline

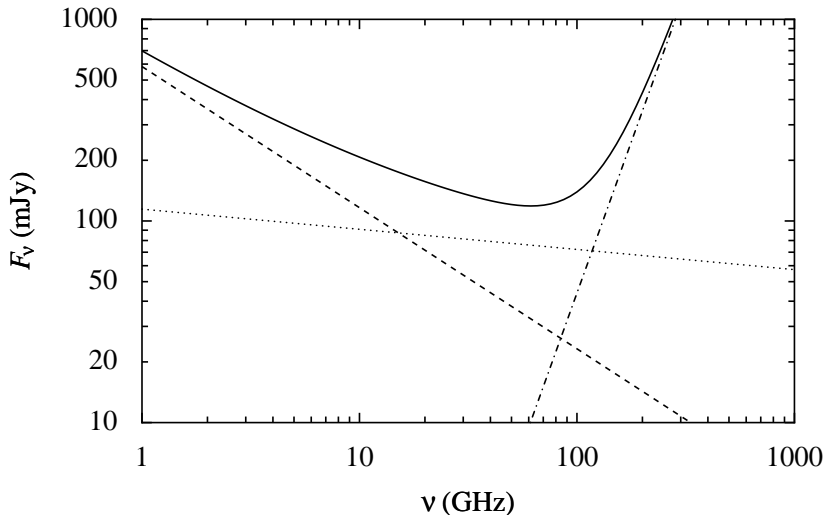
- 1 Introduction
- 2 Method: Bayesian Analysis/Nested sampling
  - Bayesian analysis in a nutshell
  - Implementation
  - Visualisation
- 3 Free-free component in a supernova remnant
- 4 Spinning dust
- 5 Spectra of (U)LIRGs
- 6 Summary

# Introduction

- Motivation:
  - Started as preparation for analysis of some forthcoming observational data (GBT+MUSTANG)
  - Turned out useful in its own right
- Mostly a *spare-parts* project - reused various components I developed for other purposes
- Method paper: Nikolic (2009)
- All of the source code available under GPL license from:  
<http://www.mrao.cam.ac.uk/~bn204/galevol/specs/index.html>

# Schematic radio spectrum of a star-forming galaxy

Maybe add low-frequency turn-over?

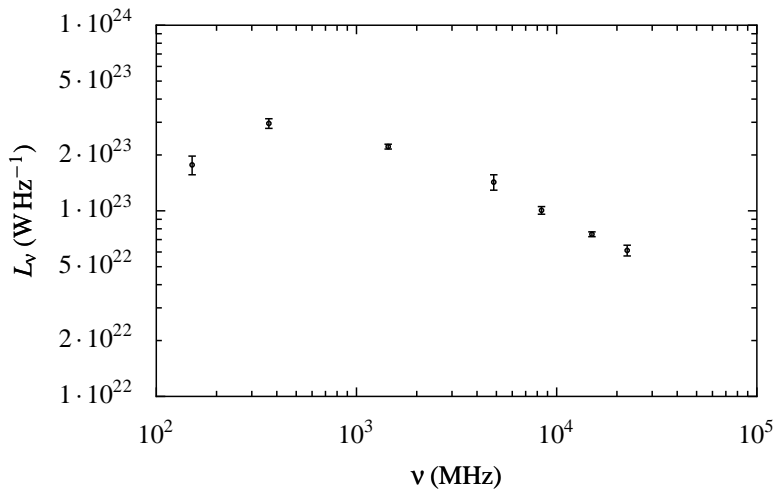


Schematic & **hypothetical** (continuum-only) spectrum of NGC 3627: the dashed line is the synchrotron component; the

# Why analyse radio spectra

- Energetics
  - Reconstruct the total energy balance from few/sparse measurements of the spectrum
  - What physical process is the source of the energy?
- Geometry:
  - Size/filling factor from the low-frequency turnover
- Dynamics:
  - Electron ageing
  - Are **equilibrium** models of star formation sufficient?
- Redshift determination  $\equiv$  radio “photometric” redshifts
  - Currently mostly used for sub-millimetre selected (“SCUBA”) sources
- Physics:
  - Free-free emission
  - Slope of the dust continuum – physics of interstellar dust

# Example observed radio spectrum



Observed spectrum of Arp 220 [data from Clemens et al. (2010)].

# Analysis strategy: model fitting

- Non-linear (free-free absorption, slope of synchrotron, etc.) so simple inversions not in general reliable
- In radio, sub-mm and far-IR, the physics is fairly well understood so can propose **candidate models** easily
- Computationally relatively easy:
  - Synchrotron radiation (analytic or 1-D integral)
  - Thermal free-free (analytic)
  - Modified black-body emission from dust (analytic or 1-D integral)
  - Spinning dust models (analytic)

# Requirements for model fitting

- 1 An objective measure of how well the model fits the observed data
- 2 For all model parameters:
  - 1 Unbiased estimates
  - 2 Error on these estimates
  - 3 Correlations between the errors
  - 4 Full probability distributions if significantly non-Gaussian
- 3 An objective way of **comparing** how well *different* models fit the data
- 4 A mechanism to incorporate already known constraints on model parameters
- 5 Visualisation of the fit in comparison to observations

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# Bayesian analysis in a nutshell

- ① Can handle “nuisance” parameters
  - ① They do not bias estimates of other parameters
  - ② They are correctly taken into account when calculating significance
- ② Fully describes non-Gaussian distributions
  - ① Even multi-modal distributions (although not very efficiently in the present implementation of my program)
- ③ Unbiased
  - ① No fitting into the noise (e.g., “flux-boosting”)
- ④ Objective model (or hypothesis) selection

# Bayes equation & the evidence

$$p(\theta|D, H) = \frac{p(D|\theta, H)p(\theta|H)}{p(D|H)}$$

- $D$ : Observed data  $\rightarrow$  flux density at several frequencies
- $H$ : Hypothesis  $\rightarrow$  model for emission & priors for parameters
- $p(D|\theta, H)$ : Likelihood  $\rightarrow$  given a model *and* its parameters, how likely are the observed data?
- $p(\theta|D, H)$ : Posterior  $\rightarrow$  given a model, what we know about its parameters
- $p(D|H)$ : “Evidence”, objective measure of how good the model is

# Bayes equation & the evidence

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# Calculating the evidence

Evidence is an integral over the likelihood over prior volume

$$p(D|H) = \int d\theta p(D|\theta, H)p(\theta|H)$$

- Evidence is not available from standard Markov Chain Monte Carlo calculations
- I use a new implementation of the **nested sampling** algorithm by Skilling (2006). Compared to MCMC, this algorithm is:
  - Efficient (fewer likelihood function evaluations)
  - Reliable (less chance of getting stuck in local maxima)
  - The output is both the evidence and the posterior distribution
- The algorithm is available **under GPL**

# Inputs/outputs

## Inputs

- 1 The model (as a routine/class)
- 2 Priors: only flat, independent priors supported. I.e., a “prior box”
- 3 Observed data and errors (for the moment assumed Gaussian)

## Outputs

- 1 The evidence value
- 2 Histograms of marginalised distributions of each model parameter
- 3 Two-dimensional histograms of partially marginalised distributions
- 4 Fan-diagram of flux vs frequency
- 5 Maximum likelihood plot of flux vs frequency

# Inputs/outputs

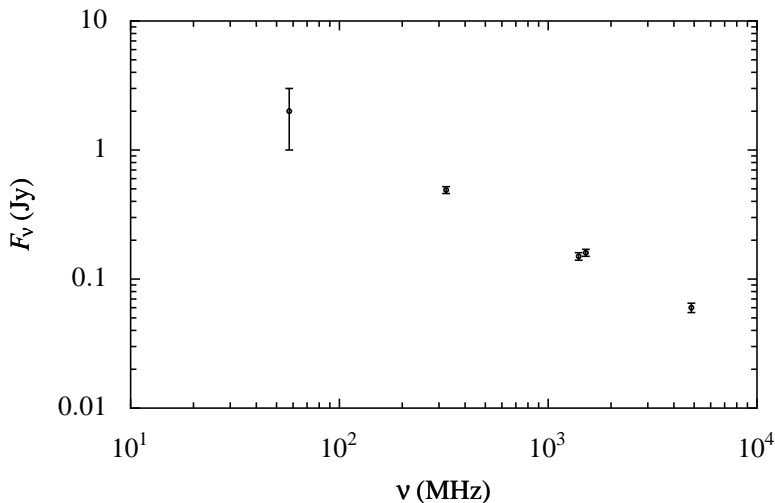
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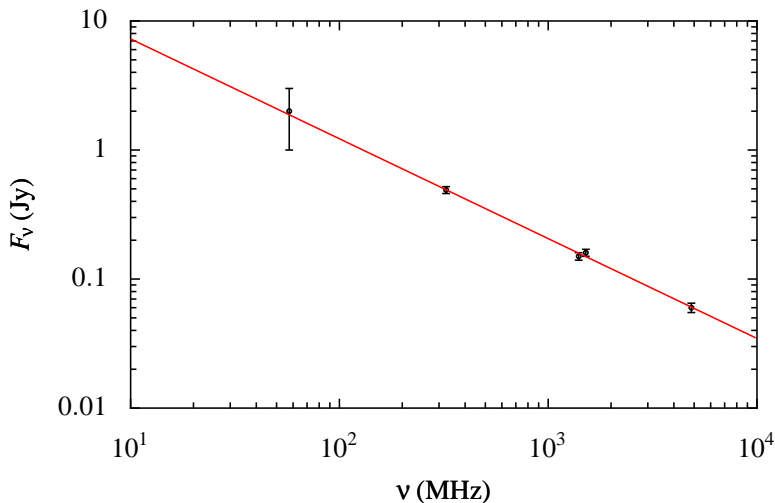
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# NGC 628 observations



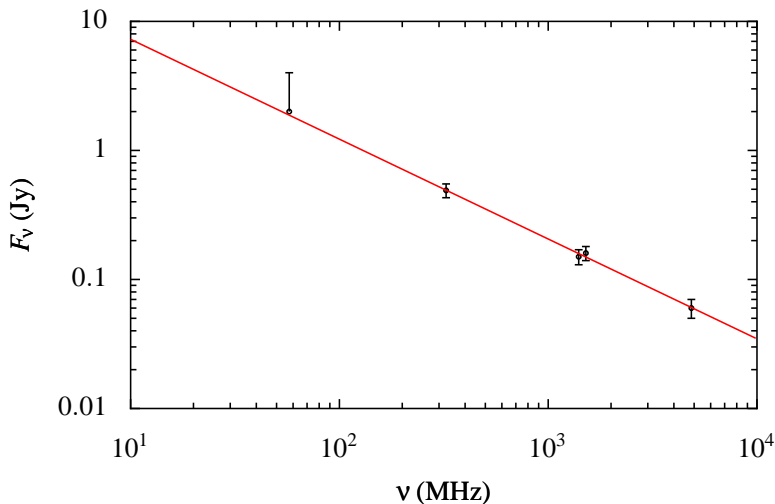
Observations at five frequencies of the near-by galaxy NGC 628 collected by Paladino et al. (2009)

# NGC 628 – max. likelihood synchrotron fit



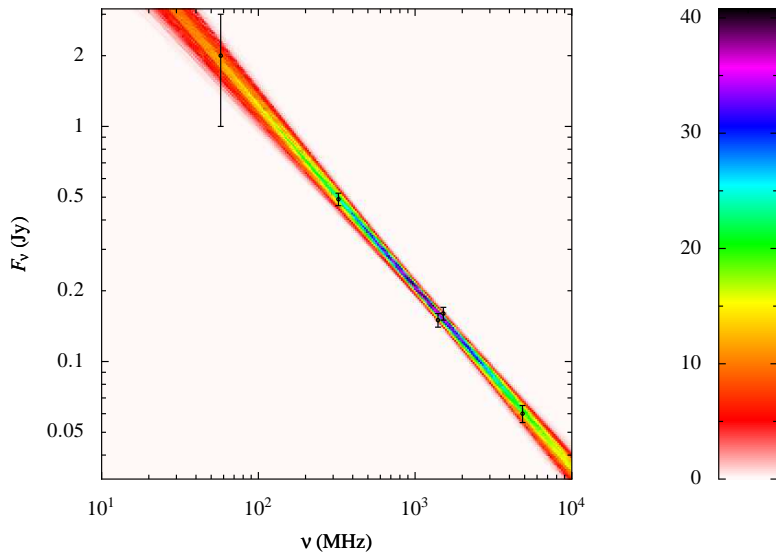
Observations at five frequencies of the near-by galaxy NGC 628 collected by Paladino et al. (2009)

## NGC 628 – doubled errors &amp; max. likelihood fit

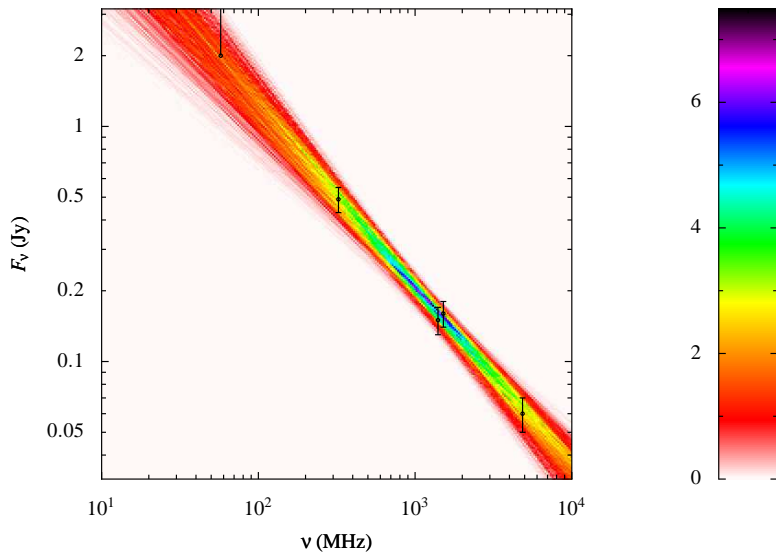


I have scaled up the error estimates by a factor of two

## NGC 628 – original errors &amp; fan-diagram



## NGC 628 – doubled errors &amp; fan-diagram



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# Introduction

Data courtesy of D. A. Green in Cambridge

- Analysis of spectrum of supernova remnant HB3
- Is there evidence for flattening of the spectrum?
  - Could be interpreted as a thermal free-free component due to interaction of a shock with the molecular cloud
- See Urošević et al. (2007), Green (2007), Onić & Urošević (2008)

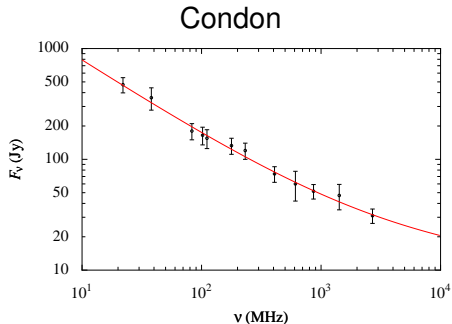
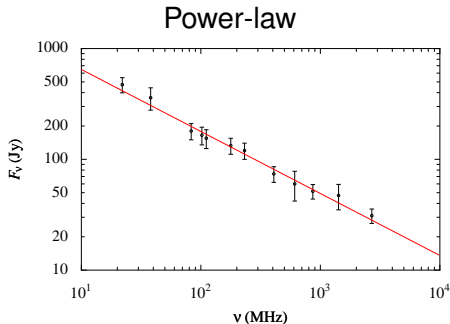
# Condon model

- Single power-law synchrotron with slope ( $\alpha$ ) as a free parameter
- Free-free emission component ( $H$  is the thermal fraction at 1 GHz)
- Thermal free-free absorption at low frequencies ( $\tau$  is the optical depth at 1 GHz)

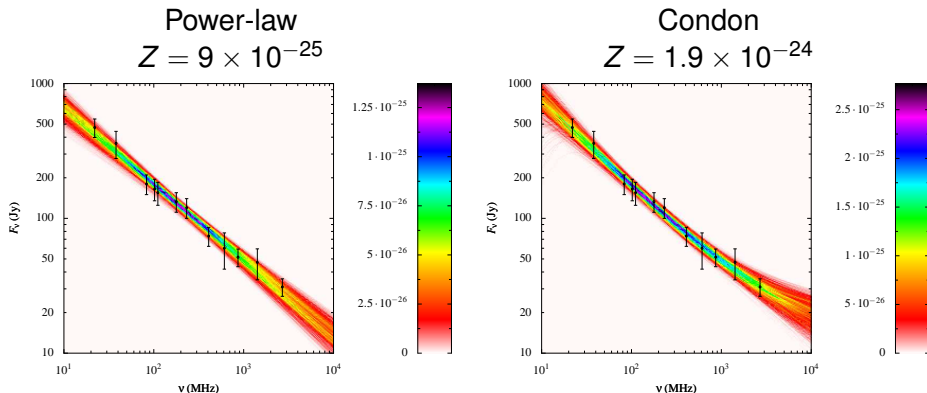
$$A(\nu; \tau^*) = 1 - \exp \left[ -10^{\tau^*} \left( \frac{\nu}{1 \text{ GHz}} \right)^{-2.1} \right]$$

$$F_\nu(\nu; H, \alpha) = \frac{A(\nu)}{A(1 \text{ GHz})} \left( \frac{\nu}{1 \text{ GHz}} \right)^2 \left[ H + (1 - H) \left( \frac{\nu}{1 \text{ GHz}} \right)^{0.1 + \alpha} \right]$$

# Maximum likelihood model fits for HB3



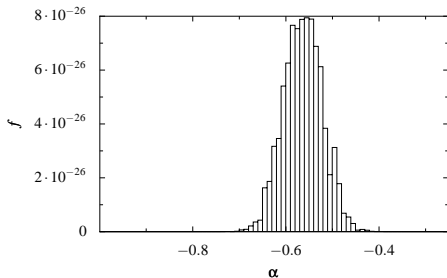
# Fan diagrams of the Bayesian analysis of HB3



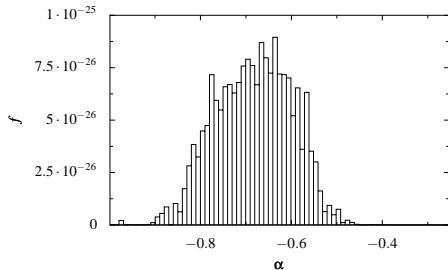
Priors:  $-1 < \alpha < -0.25$   
 $0 < H < 0.6$   
 $-20 < \log \tau < -2$

Marginalised distribution of  $\alpha$ 

Power-law  
 $Z = 9 \times 10^{-25}$



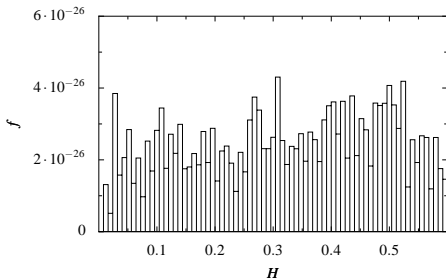
Condon  
 $Z = 1.9 \times 10^{-24}$



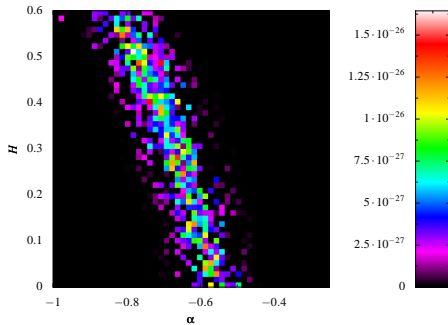
# (Partially-)Marginalised distribution of $H$

Condon model, i.e., synchrotron + thermal emission

PD of thermal fraction at 1 GHz

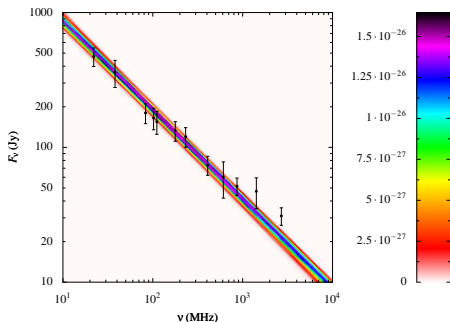


Correlation between thermal fraction and synchrotron slope

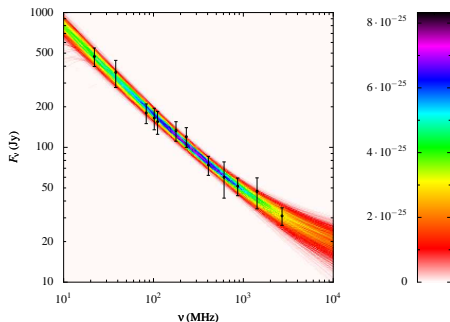


# New prior: constrain $\alpha$ to 0.1 range around -0.7

Power-law  
 $Z = 1.2 \times 10^{-25}$



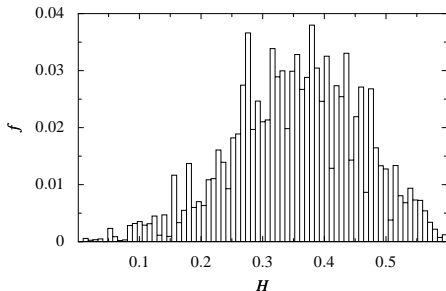
Condon  
 $Z = 5.4 \times 10^{-24}$



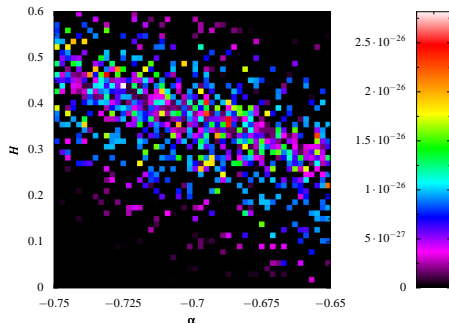
Priors:  $-0.75 < \alpha < -0.65$   
 $0 < H < 0.6$   
 $-20 < \log \tau < -2$

# New prior: distribution of $H$

PD of thermal fraction at 1 GHz



Correlation between thermal fraction and synchrotron slope



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# Introduction I

See Scaife et al. (2010) for details

## Original observation

- Centimetre-wave emission correlated with galactic dust foregrounds (e.g., de Oliveira-Costa et al., 1997) in excess to what is expected from simple models
- Draine & Lazarian (1998) suggest it could be due to spinning dust

## Vigorous debate about the nature the emission...

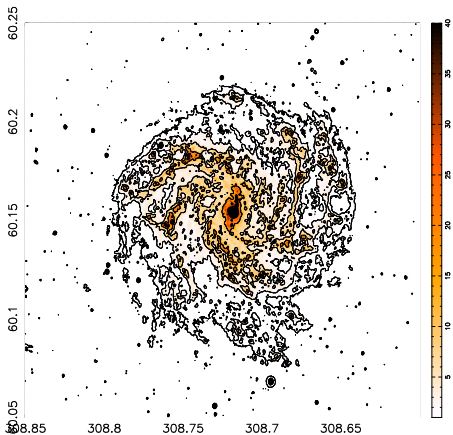
## First detection in an external galaxy

Murphy et al. (2010) – but is it due to spinning dust or a very compact HII region?

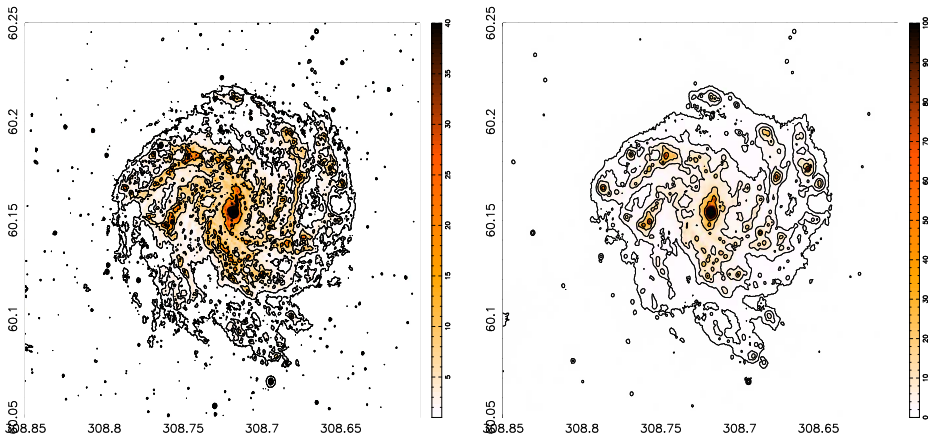
# Introduction II

See Scaife et al. (2010) for details

8 micron



24 micron

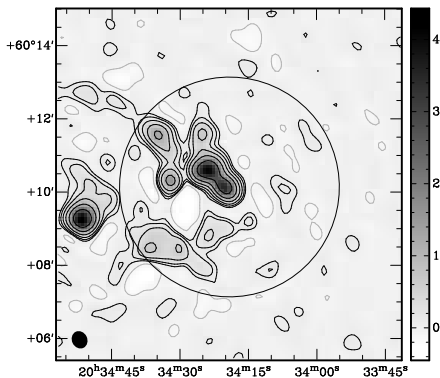


Spitzer/SINGS images of NGC 6946

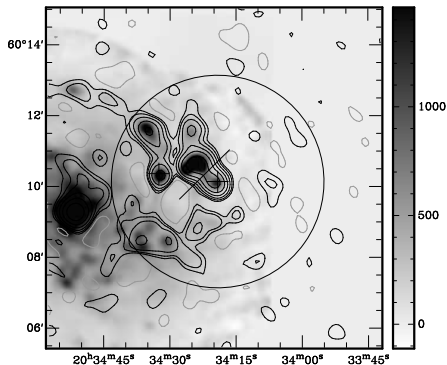
# Introduction III

See Scaife et al. (2010) for details

15 GHz contour + grayscale



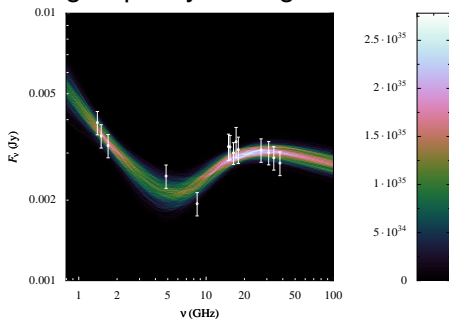
15 GHz contour +  
8.5 GHz grayscale



# Results

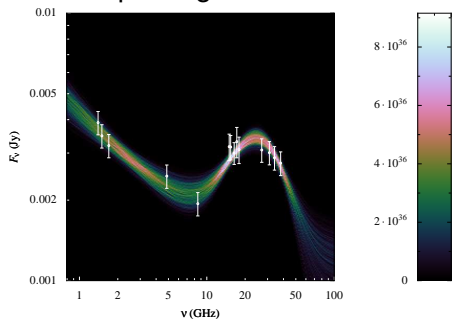
“Region 4” – the anomalous region

## High-opacity HII region model



$$Z = 2.4 \times 10^{36}$$

## Spinning dust model



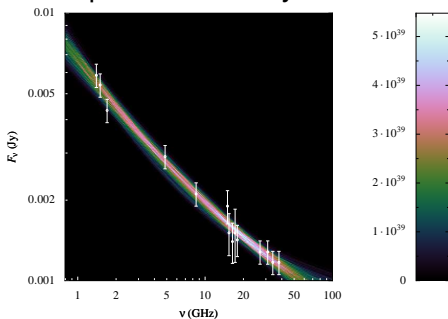
$$Z = 7.2 \times 10^{37}$$

Clearly should prefer this model

# Results

“Region 8” – a normal region

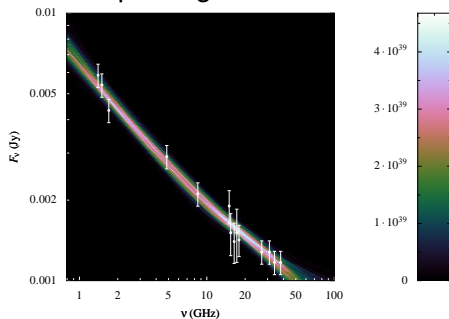
Simple thermal + synchrotron



$$Z = 4.8 \times 10^{40}$$

Simpler model preferred!

Spinning dust model



$$Z = 4.3 \times 10^{40}$$

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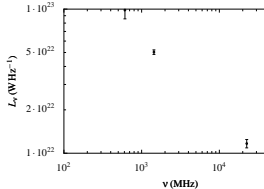
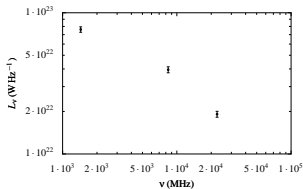
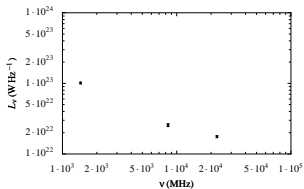
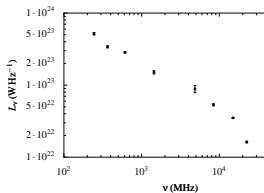
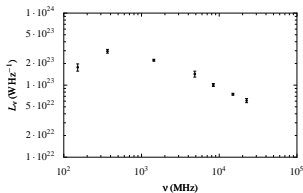
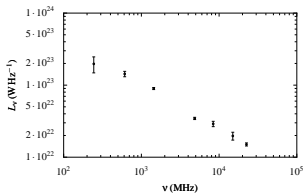
# Spectra of (U)LIRGs

- Work in-progress – collaboration with Marcel Clemens from U. of Padua
- Data are based on Clemens et al. (2010)

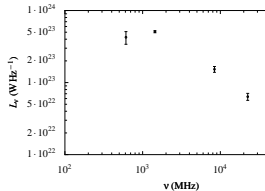
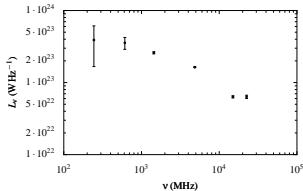
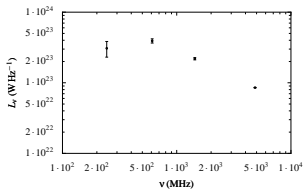
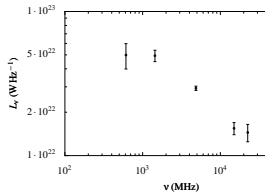
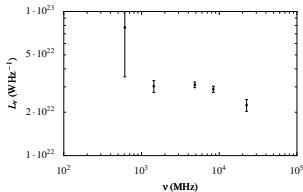
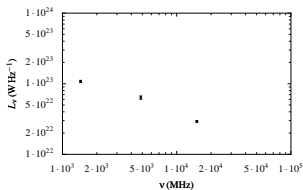
## Goals:

- Constrain the geometry of star-formation from radio to far IR spectra
- Physics of synchrotron emission (ionisation losses, underlying causes for the far IR-radio correlation) and of free-free emission (why don't we see it?)
- Constrain AGN contribution?

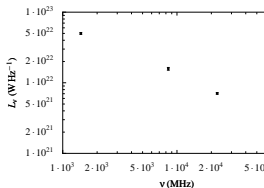
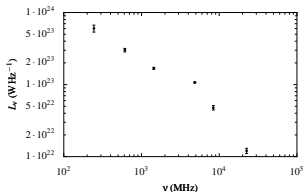
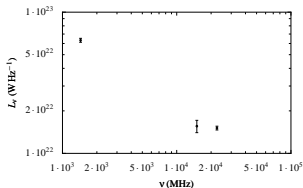
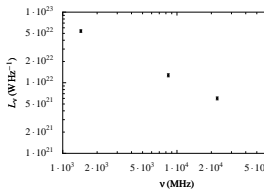
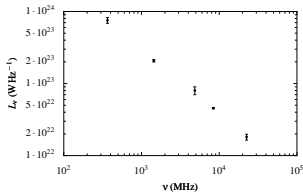
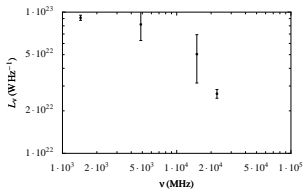
## Simple one-component model I



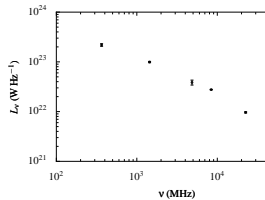
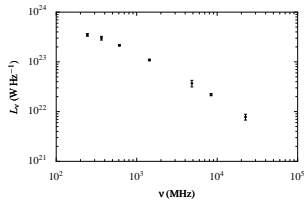
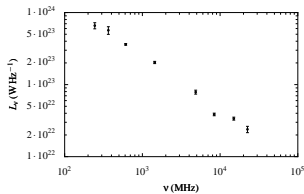
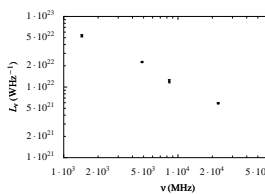
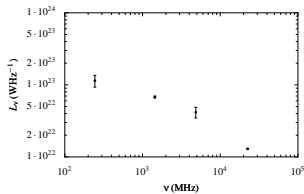
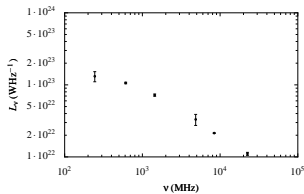
## Simple one-component model II



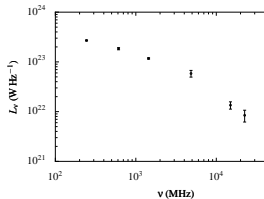
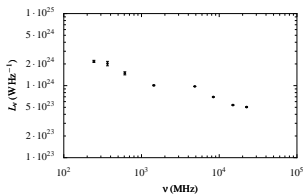
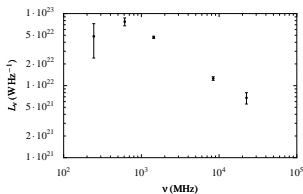
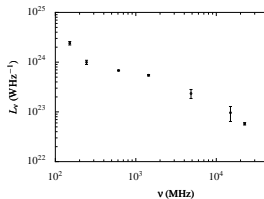
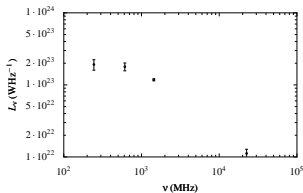
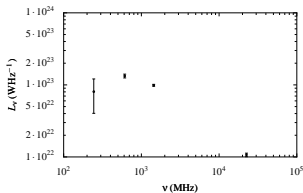
## Simple one-component model III



## Simple one-component model IV



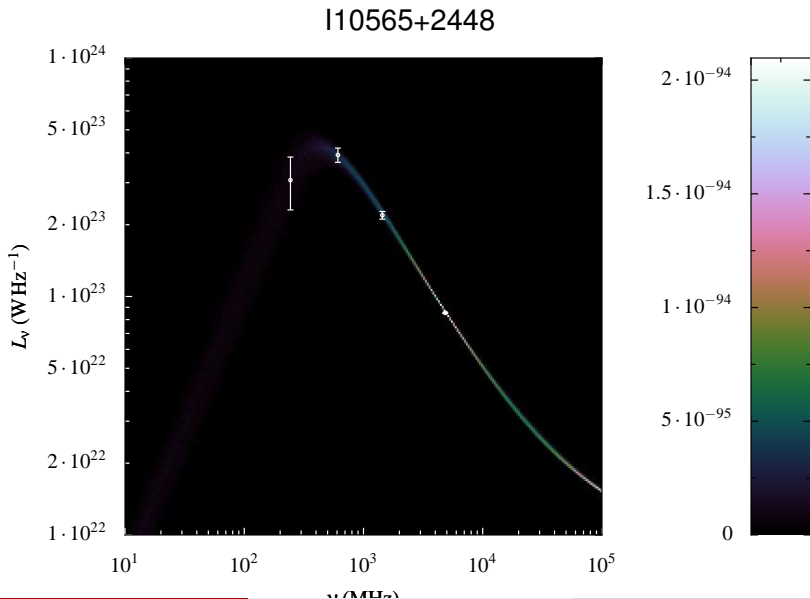
## Simple one-component model V



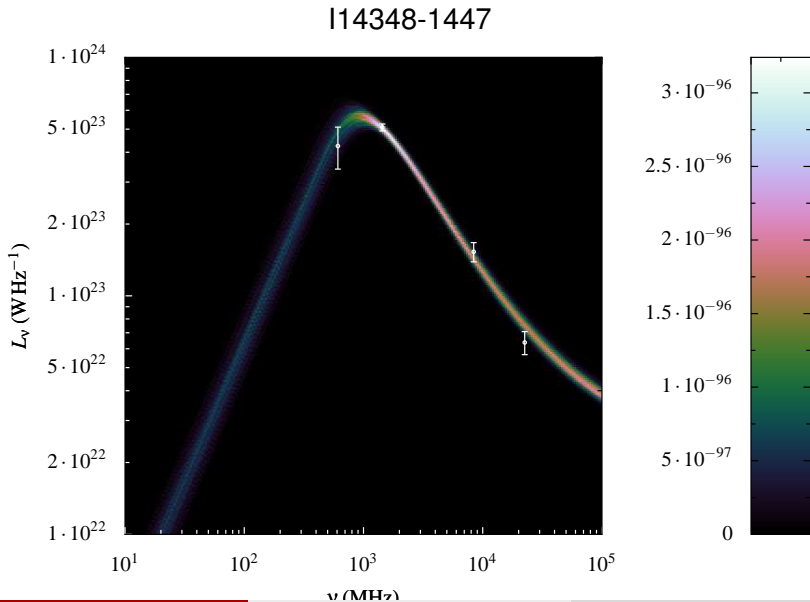
# Simple one-component model

- Spherical, homogeneous, well-mixed (“giant”) star-forming region
- Synchrotron emission based on the supernova rate  
 $\alpha$  a free parameter from -1 to -0.5
- Free emission and absorption based on the ionising photon rate  
 $\log 10 R_g$  free parameter: radius of the region
- Far-infrared emission linearly dependent to SFR and based on the FIR-radio correlation

## (U)LIRGs well fit by one-component models I

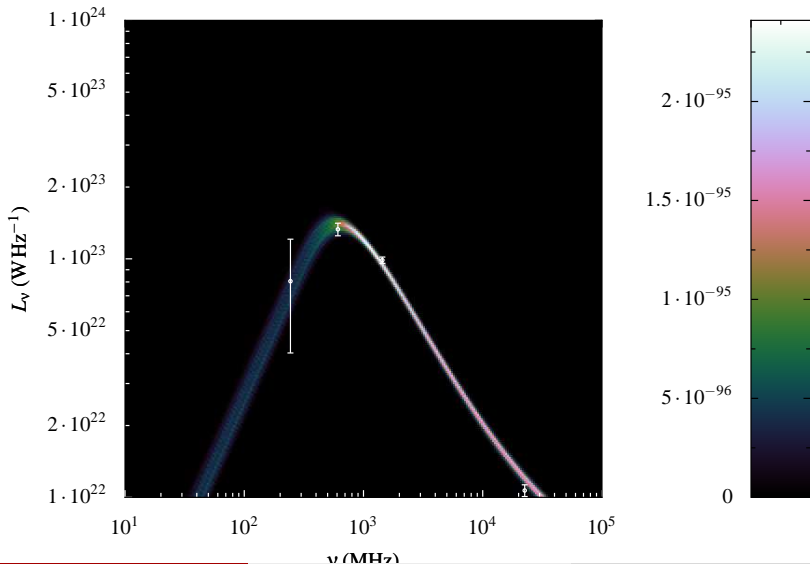


## (U)LIRGs well fit by one-component models II



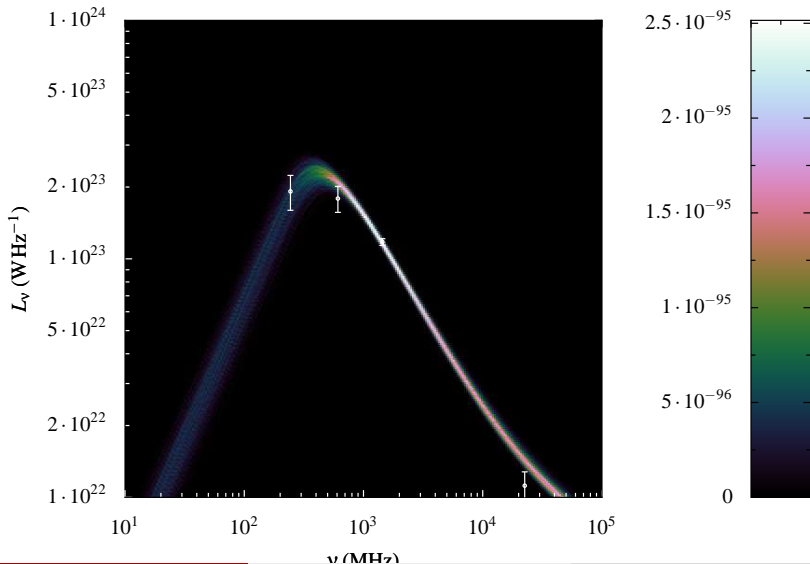
## (U)LIRGs well fit by one-component models III

UGC2369

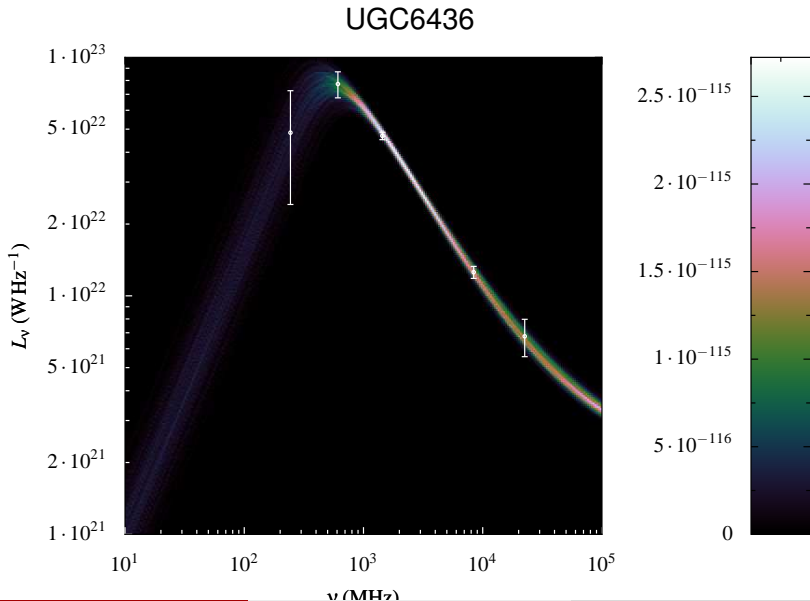


## (U)LIRGs well fit by one-component models IV

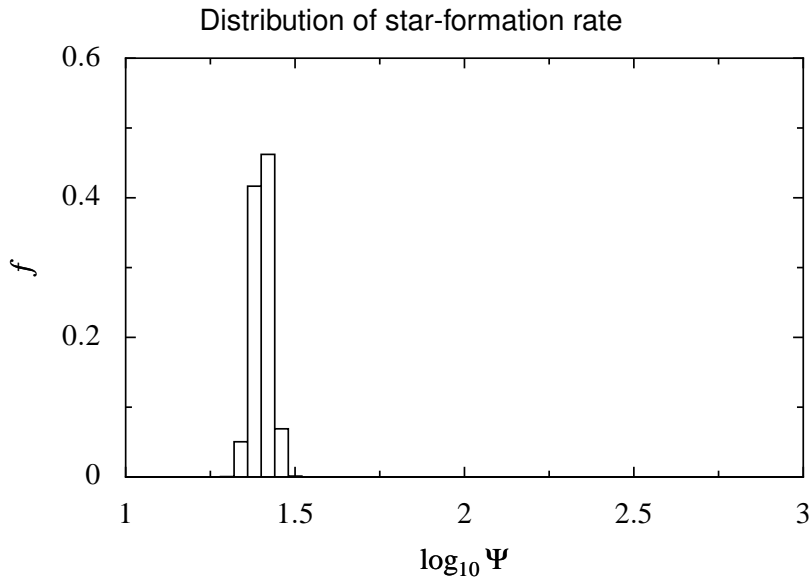
UGC4881



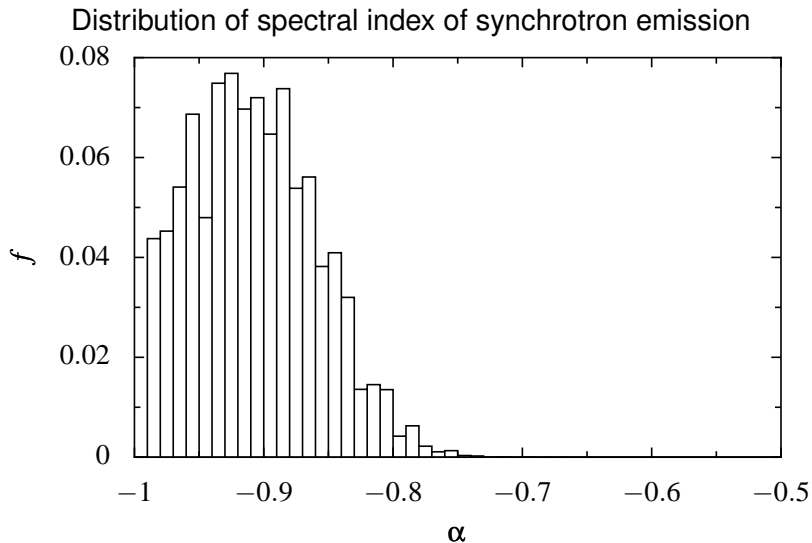
## (U)LIRGs well fit by one-component models V



## UGC 6436 – parameter distributions I

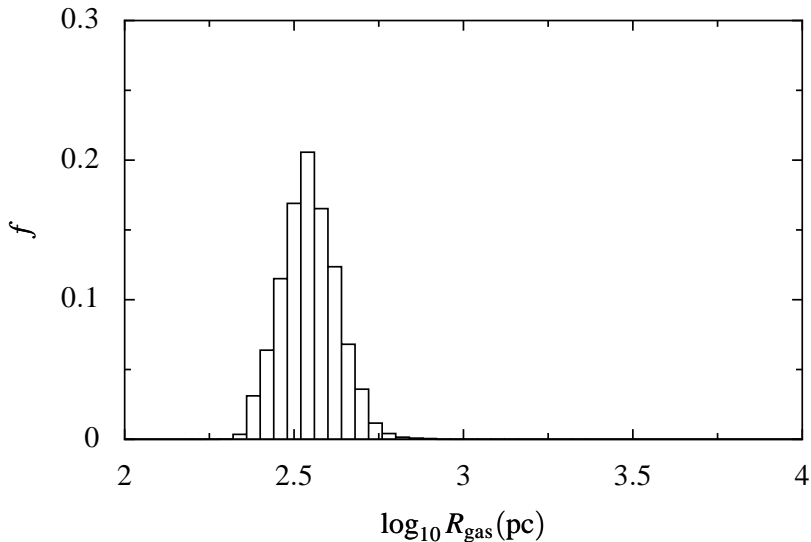


## UGC 6436 – parameter distributions II

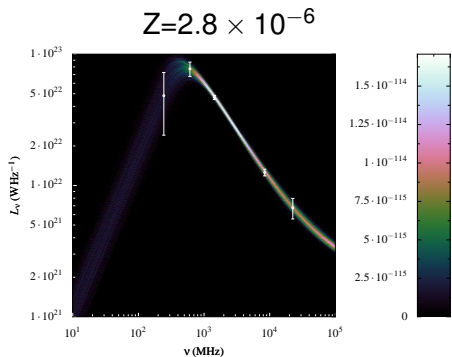
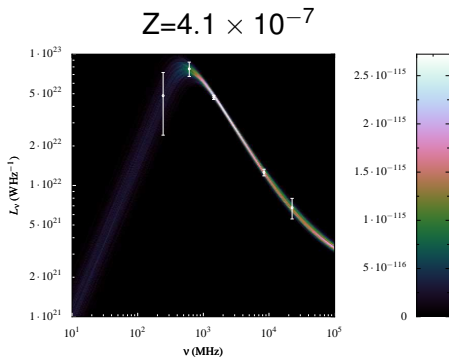


## UGC 6436 – parameter distributions III

Distribution of radius of molecular gas

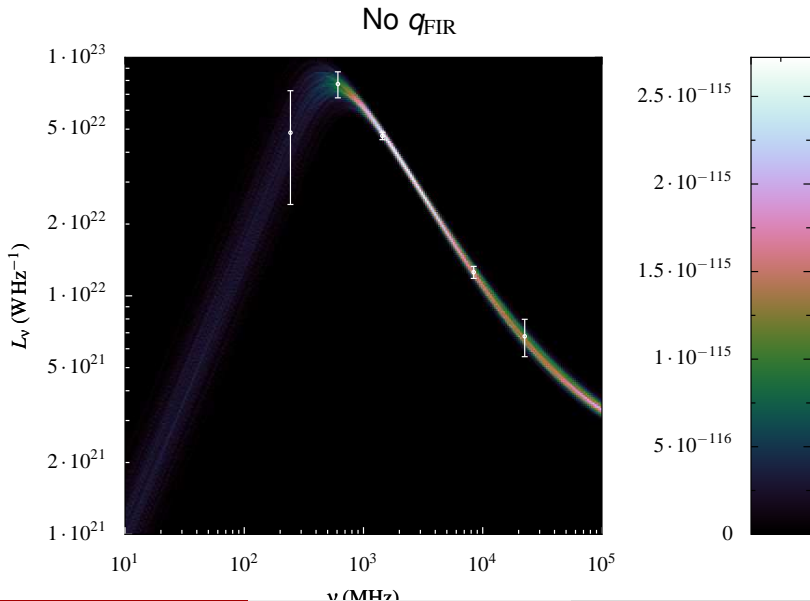


## UGC 6436 – are radio and IR consistent? I

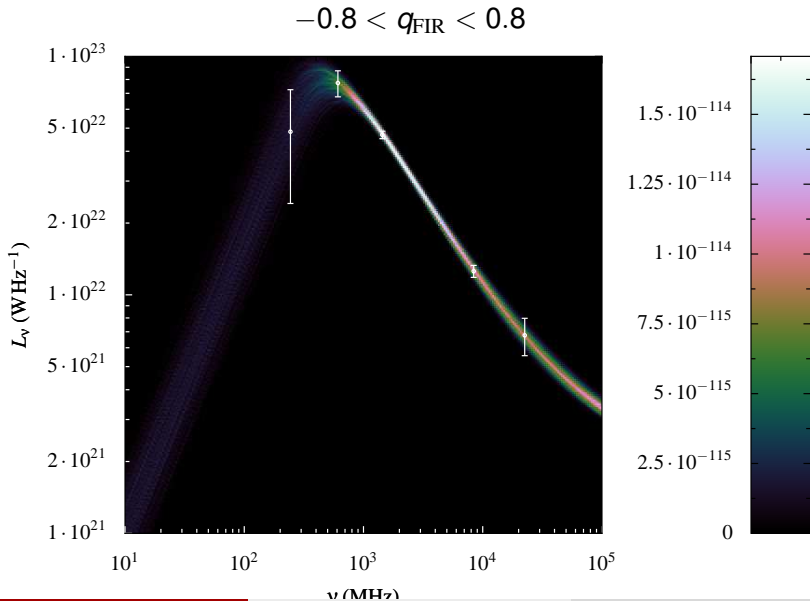


FIR emission per unit SFR  
is a free parameter  
[in fact: ratio to nominal ( $q_{\text{FIR}}$ )]

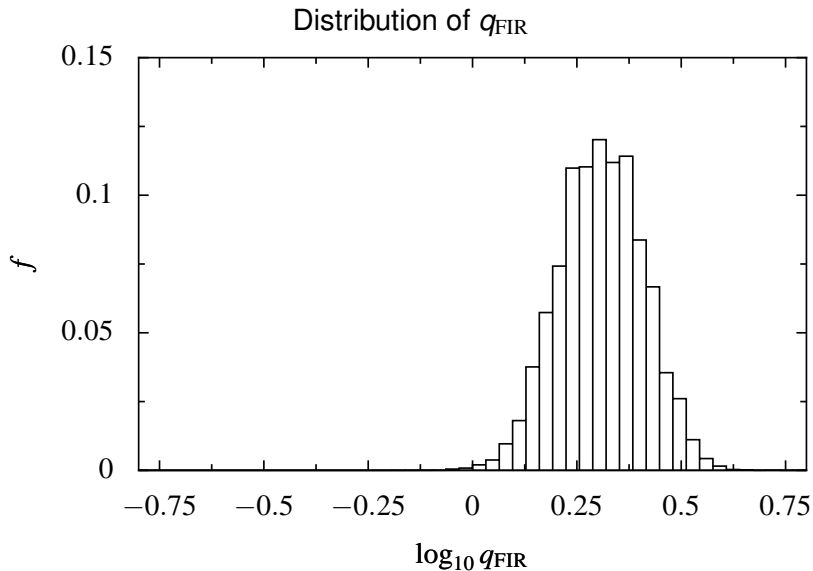
## UGC 6436 – are radio and IR consistent? II



## UGC 6436 – are radio and IR consistent? III



## UGC 6436 – are radio and IR consistent? IV



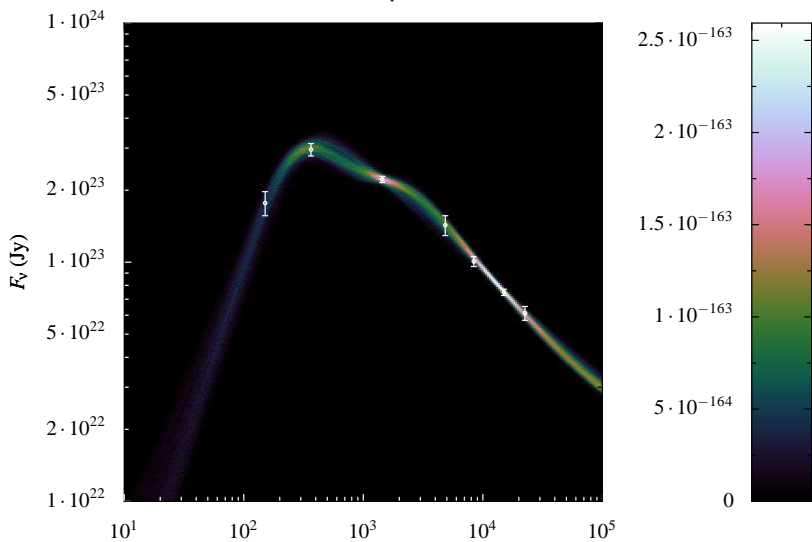
# Simple two-component model

## Two of these!

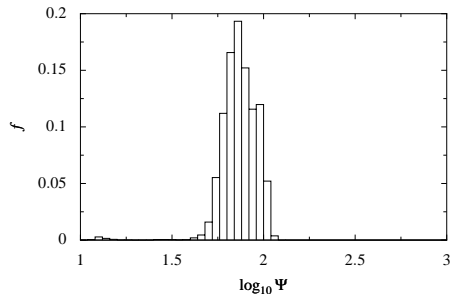
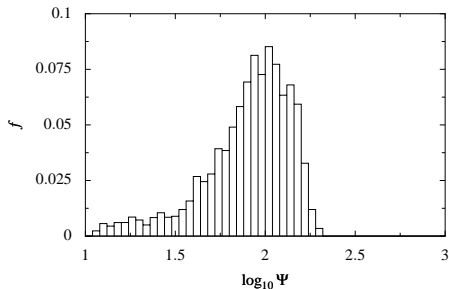
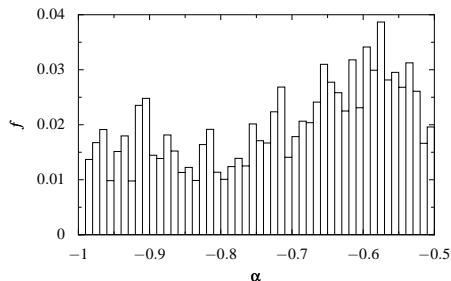
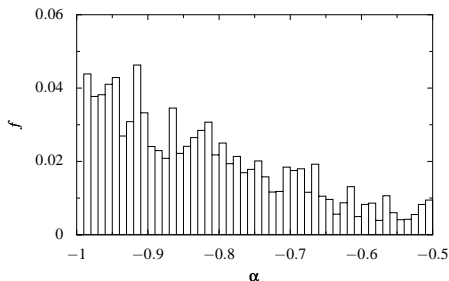
- Spherical, homogeneous, well-mixed (“giant”) star-forming region
- Synchrotron emission based on the supernova rate  
 $\alpha$  a free parameter from -1 to -0.5
- Free emission and absorption based on the ionising photon rate  
 $\log 10R_g$  free parameter: radius of the region
- Far-infrared emission linearly dependent to SFR and based on the FIR-radio correlation

## Arp 220 I

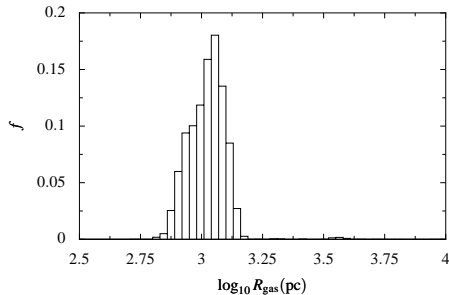
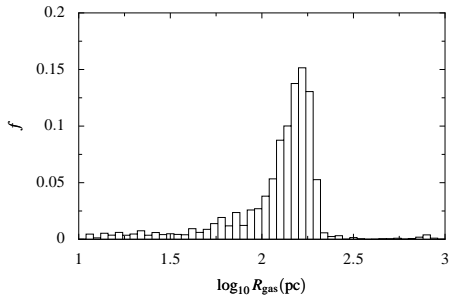
## Arp 220



## Arp 220 – parameter distributions I



## Arp 220 – parameter distributions II



# Future directions

- ... complete the present analysis!
  - More **physical** model:
    - Directly model synchrotron from the  $e^-$  spectrum
    - Kennicutt-Schmidt law for the star-formation rate
    - $B$  by hydrostatic support? Minimum energy?
  - Analysis across sample, i.e., all the galaxies together
- 
- Volunteers?

# Outline

- 1 Introduction
- 2 Method: Bayesian Analysis/Nested sampling
  - Bayesian analysis in a nutshell
  - Implementation
  - Visualisation
- 3 Free-free component in a supernova remnant
- 4 Spinning dust
- 5 Spectra of (U)LIRGs
- 6 Summary

# Summary

Bayesian analysis of radio spectra:

- Objective model selection
- Full probability distributions, parameter correlations
- Natural way to introduce physical constraints through priors
- Easily handle nuisance parameters
- (Rigorous theoretical framework)

End result:

Fewer rules of thumb. Less fiddling with parameters.

Get the full picture and extract maximum information.

Let the data guide you to which experiments you should do next

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