

Searches in Astrophysics and Cosmology



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Outline

- Methodology and recent trends in Astro-statistics
- Five real-life examples:
 1. Cosmological parameters
 2. Neutrino mass
 3. Photometric redshifts
 4. Galaxy shapes and classification
 5. Exo-planets

modern Astro-Statistics books

- Lyons (1991)
- Press et al. (1992)
- Lupton (1993)
- Babu & Feigelson (1996)
- Sivia (1996)
- Cowan (1998)
- Starck & Murtagh (2002)
- Martinez & Saar (2002)
- Wall & Jenkins (2003)
- Saha (2003)
- Gregory (2005)
- Hobson et al. (2009)

Recent trends

- Astro-Statistics is more 'respectable'.
- In Cosmology, Bayesian approaches are more popular (since 90s) than Frequentist methods.
- More awareness of model selection methods (e.g. Evidence, AIC, BIC, ...).
- Computer intensive methods (e.g. MCMC) are more common, and free packages available.

Example 1: cosmological parameters

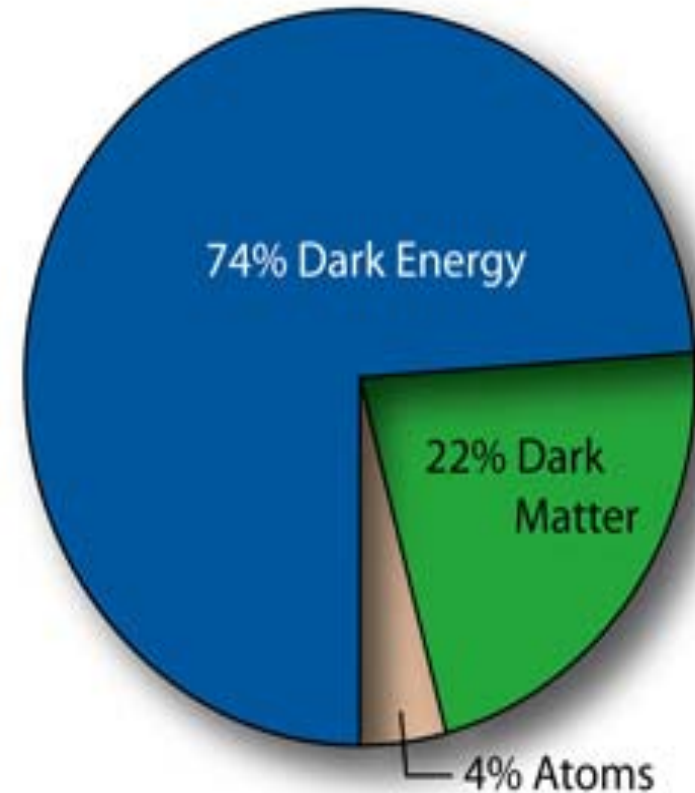
“Evidence” for the Dark Universe

Observational data

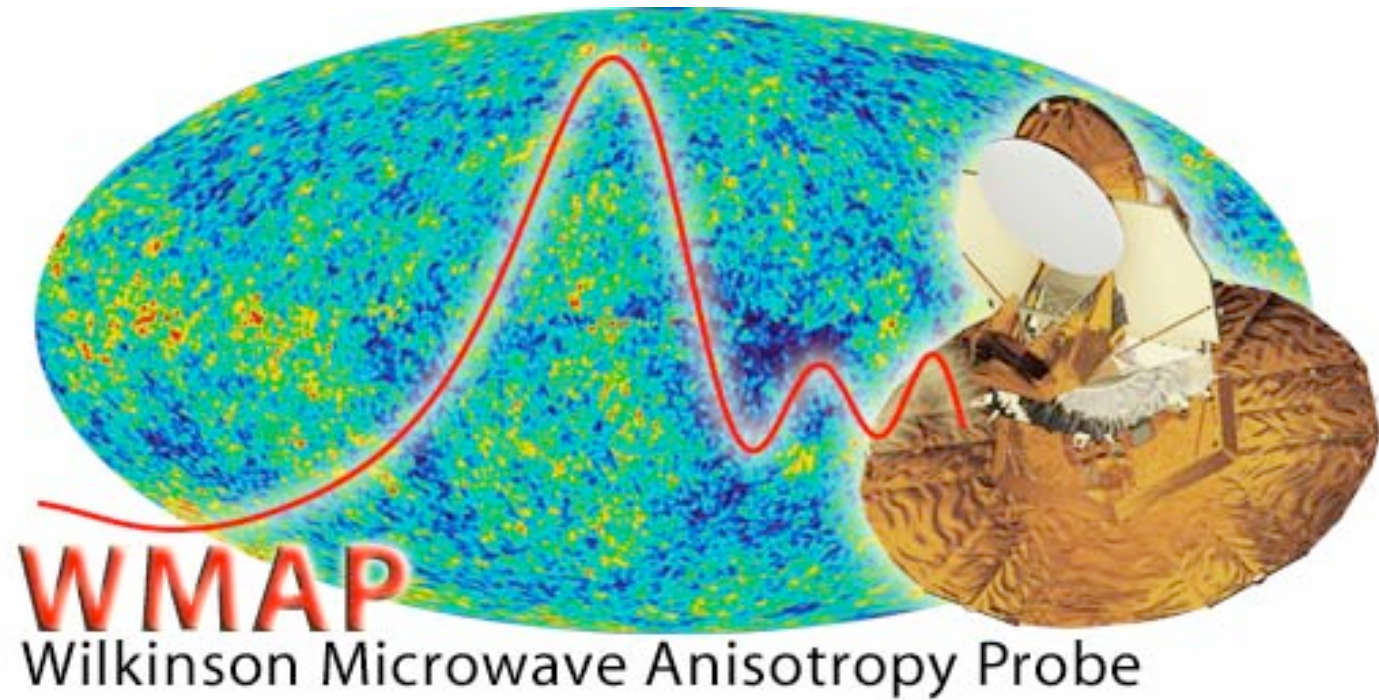
- ◆ Type Ia Supernovae
- ◆ Galaxy Clusters
- ◆ Cosmic Microwave Background
- ◆ Large Scale Structure
- ◆ Gravitational Lensing

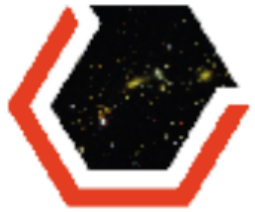
Physical effects:

- ◆ Geometry
- ◆ Growth of Structure



“WMAP ++ Cosmology”





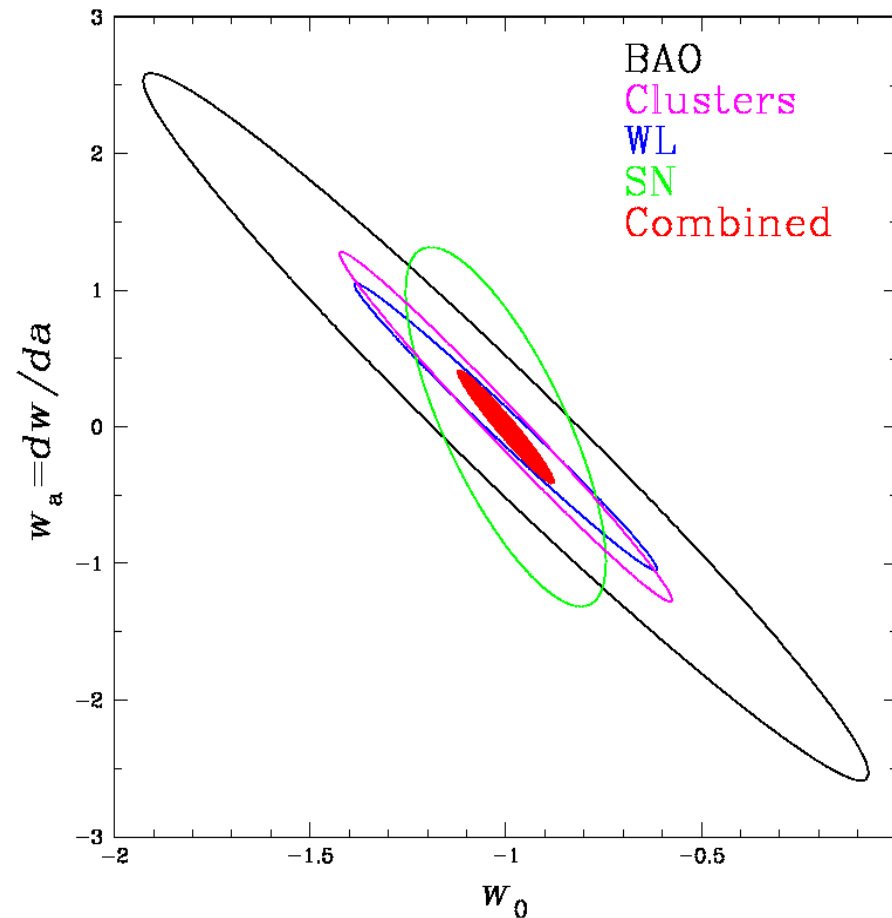
DARK ENERGY
SURVEY

The Dark Energy <http://www.darkenergysurvey.org>



Fisher Matrix FoM as a tool for decision making by funding agencies

$$w(z) = w_0 + w_a(1-a) \quad 68\% \text{ CL}$$



Example: FoM for Dark Energy Survey is a factor **4.6** tighter compared to near term projects

Sources of uncertainties

- Cosmological (parameters and priors)
- Astrophysical (e.g. cluster M-T, biasing)
- Instrumental (e.g. PSF)

Use and Abuse of

- Priors
- Marginalization
- Evidence

How to choose a prior?

(e.g. on the curvature)

- * Theoretical prejudice
(e.g. “according to Inflation the universe must be flat”)
- * Previous observations
(e.g. “we know from WMAP the universe is flat to within 2%”)
- * Parameterized ignorance (e.g. “uniform prior, Jeffrey’s prior, or Entropy prior?”)

Can we rule out $w = -1$?

WMAP 7-year Cosmological Interpretation

TABLE 4

SUMMARY OF THE 68% LIMITS ON DARK ENERGY PROPERTIES FROM WMAP COMBINED WITH OTHER DATA SETS

| Section | Curvature | Parameter | +BAO+ H_0 | +BAO+ H_0 + $D_{\Delta t}$ ^a | +BAO+SN ^b |
|-------------|-------------------|--------------|-------------------------------|---|----------------------------------|
| Section 5.1 | $\Omega_k = 0$ | Constant w | -1.10 ± 0.14 | -1.08 ± 0.13 | -0.980 ± 0.053 |
| Section 5.2 | $\Omega_k \neq 0$ | Constant w | -1.44 ± 0.27 | -1.39 ± 0.25 | $-0.999^{+0.057}_{-0.056}$ |
| | | Ω_k | $-0.0125^{+0.0064}_{-0.0067}$ | $-0.0111^{+0.0060}_{-0.0063}$ | $-0.0057^{+0.0067}_{-0.0068}$ |
| | | | + H_0 +SN | +BAO+ H_0 +SN | +BAO+ H_0 + $D_{\Delta t}$ +SN |
| Section 5.3 | $\Omega_k = 0$ | w_0 | -0.83 ± 0.16 | -0.93 ± 0.13 | -0.93 ± 0.12 |
| | | w_a | $-0.80^{+0.84}_{-0.83}$ | $-0.41^{+0.72}_{-0.71}$ | $-0.38^{+0.66}_{-0.65}$ |

Bayesian Evidence: To BE or not to BE?

$$p(d|\mathcal{M}) \equiv \int_{\Omega_{\mathcal{M}}} p(d|\theta, \mathcal{M})p(\theta|\mathcal{M})d\theta \quad (\text{Bayesian evidence}).$$

$$B_{01} \equiv \frac{p(d|\mathcal{M}_0)}{p(d|\mathcal{M}_1)} \quad (\text{Bayes factor}).$$

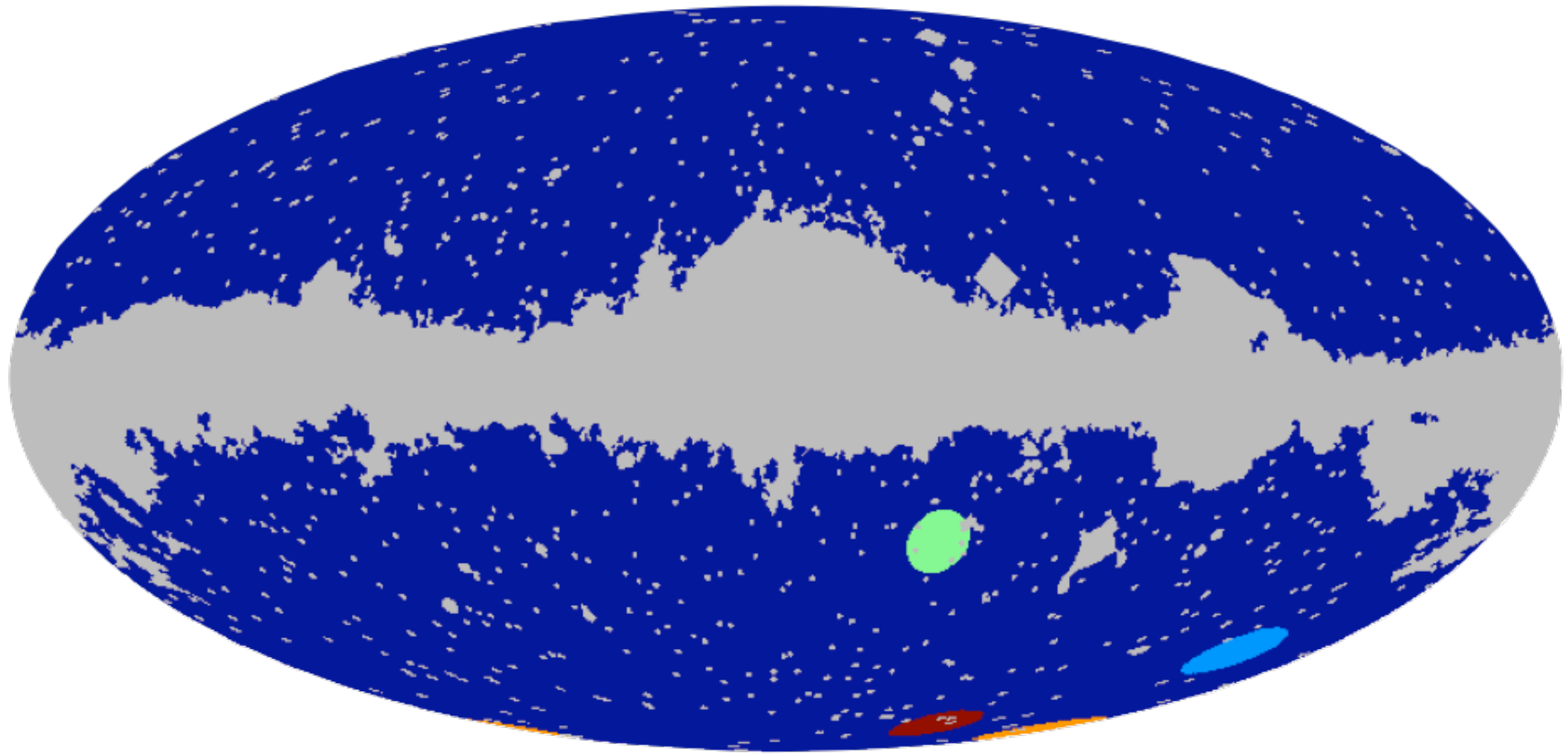
e.g. B=1 weak; B=5 strong

But how sensitive to assumed priors?

Variations: AIC, BIC, DIC,...

e.g. Liddle (2007), Trotta (2008), Efstathiou (2009)

WMAP 7-Year: circles in the CMB sky?

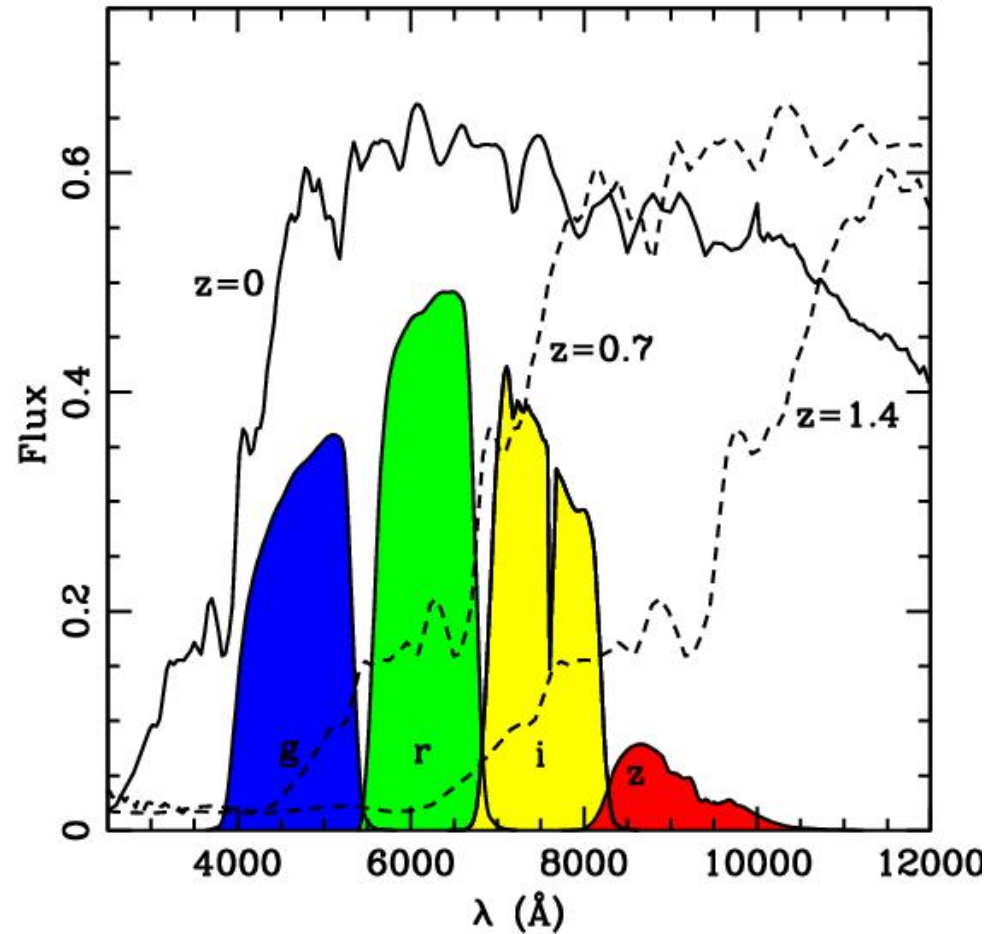


Fenney, Peiris, Johnson & Mortlock (2010)

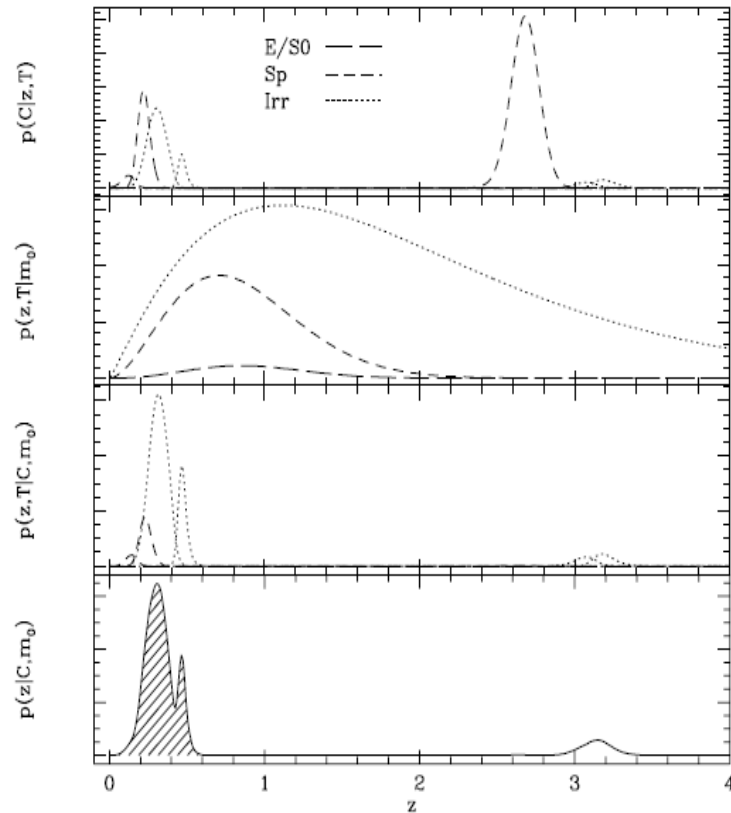
Example 2: Photometric Redshifts

Photometric redshift

- Probe strong spectral features (4000 break)
- Difference in flux through filters as the galaxy is redshifted.



Bayesian Photo-z



$$p(C|z, T)$$

likelihood

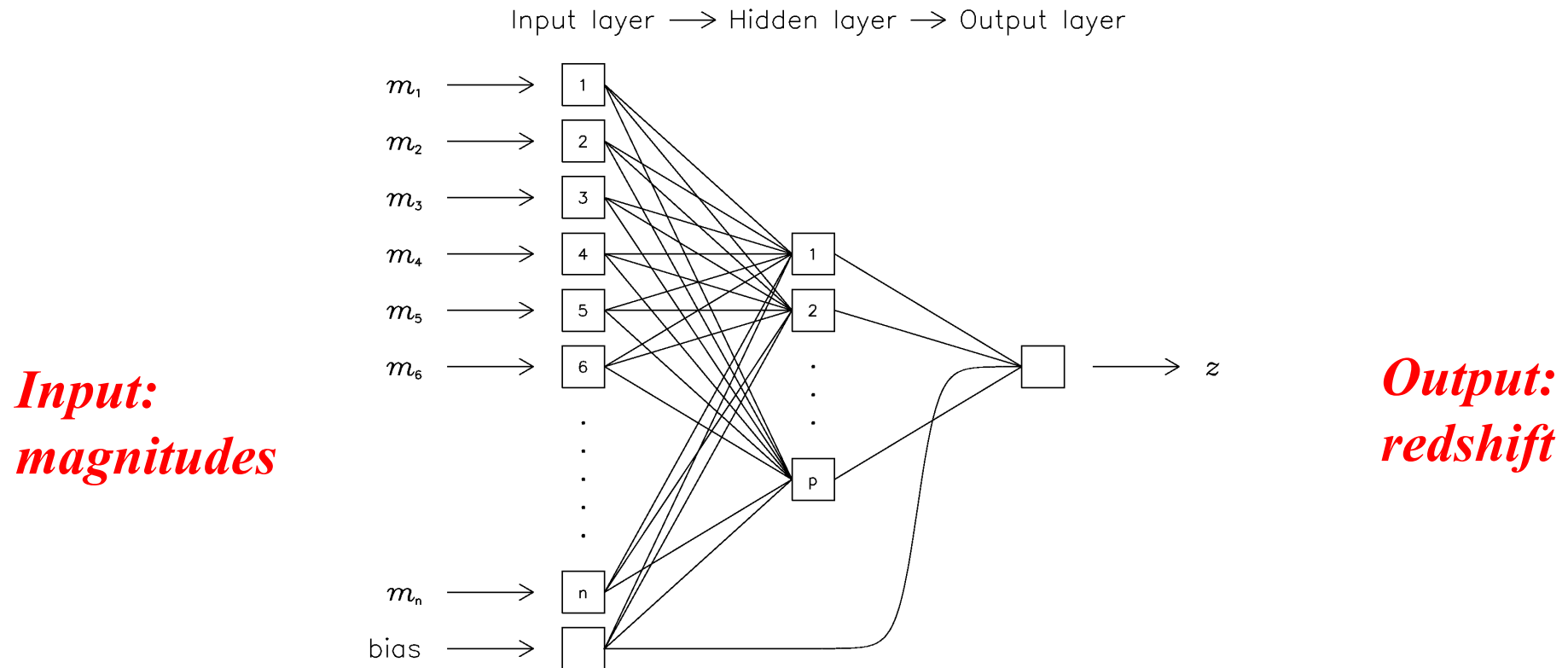
$$p(z, T|m_0)$$

prior

Redshift z

Benitez 2000 (BPZ)

ANNz - Artificial Neural Network

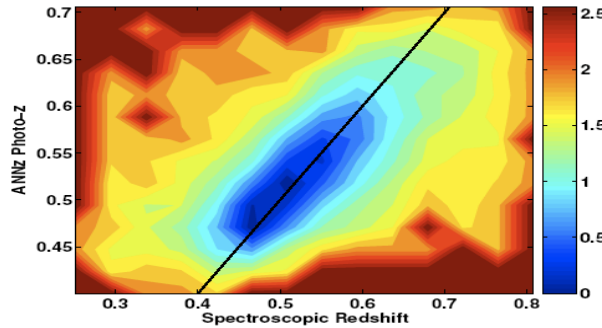


Collister & Lahav 2004

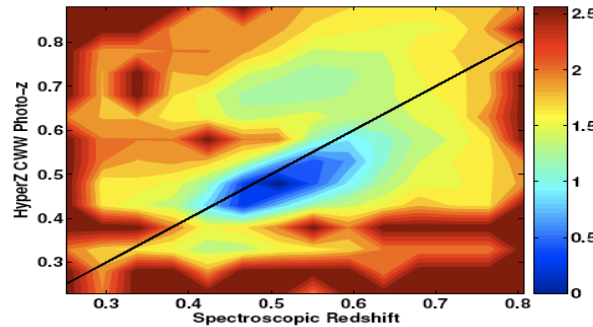
<http://www.star.ucl.ac.uk/~lahav/annz.html>

1.5M LRGs (“MegaZ”) photo-z code comparison

ANNz

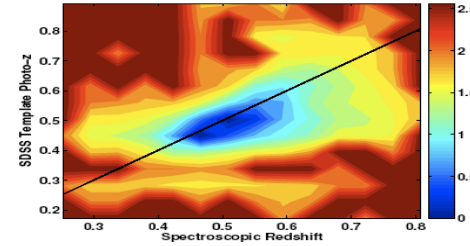


HpZ+WWC

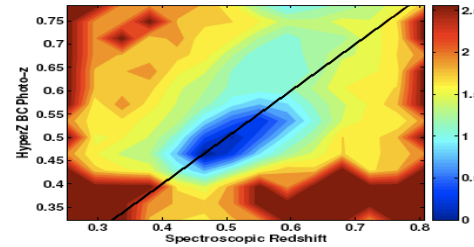


Abdalla, Banerji, Lahav & Rashkov

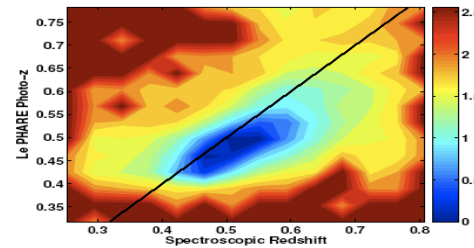
Cf. PHAT (Hildebrandt et al. 2010)



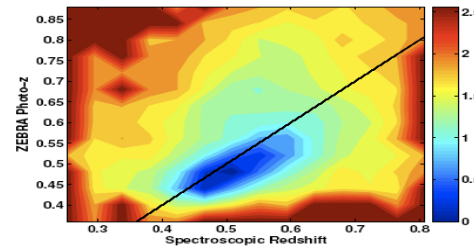
SDSS



HpZ+BC



Le PHARE



Zebra

Example 3: Neutrino mass

KATRIN

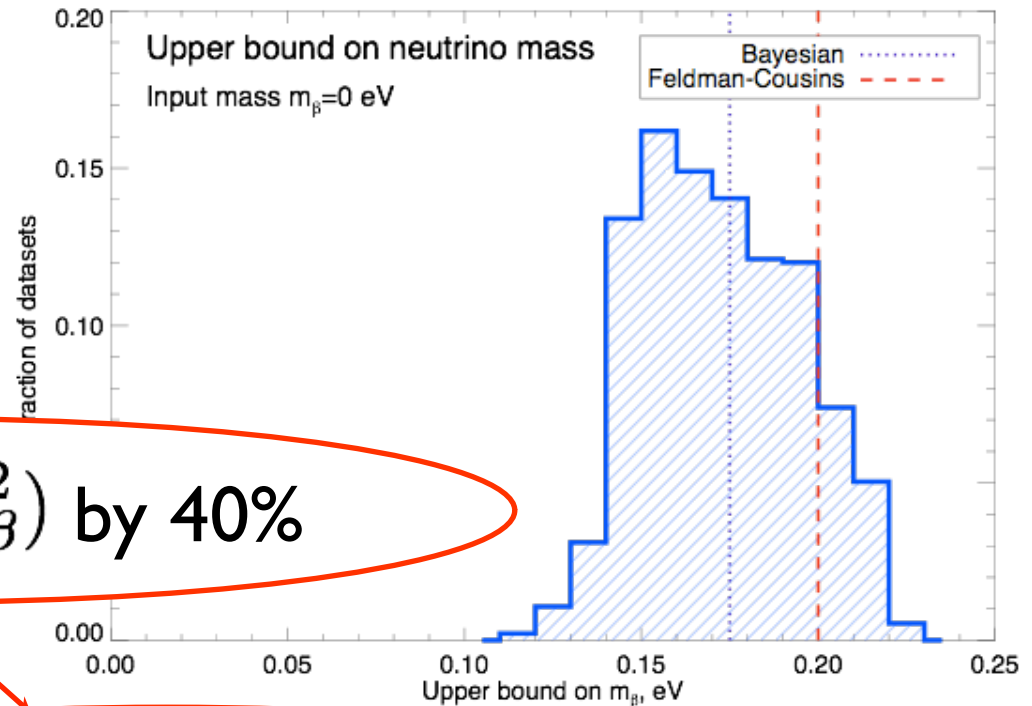
- Next generation tritium beta decay spectrometer
- Sensitivity
 $m_{\beta} < 0.2 \text{ eV}$ at 90% C.L.
- 5σ detection threshold at
 $m_{\beta} = 0.35 \text{ eV}$
- KATRIN Design Report Angrik et al. (2005)

Frequentist analysis!



1) Null result

- Assume zero mass
- Frequentist reference:
 $m_\beta < 0.20$ eV, 90%C.L.



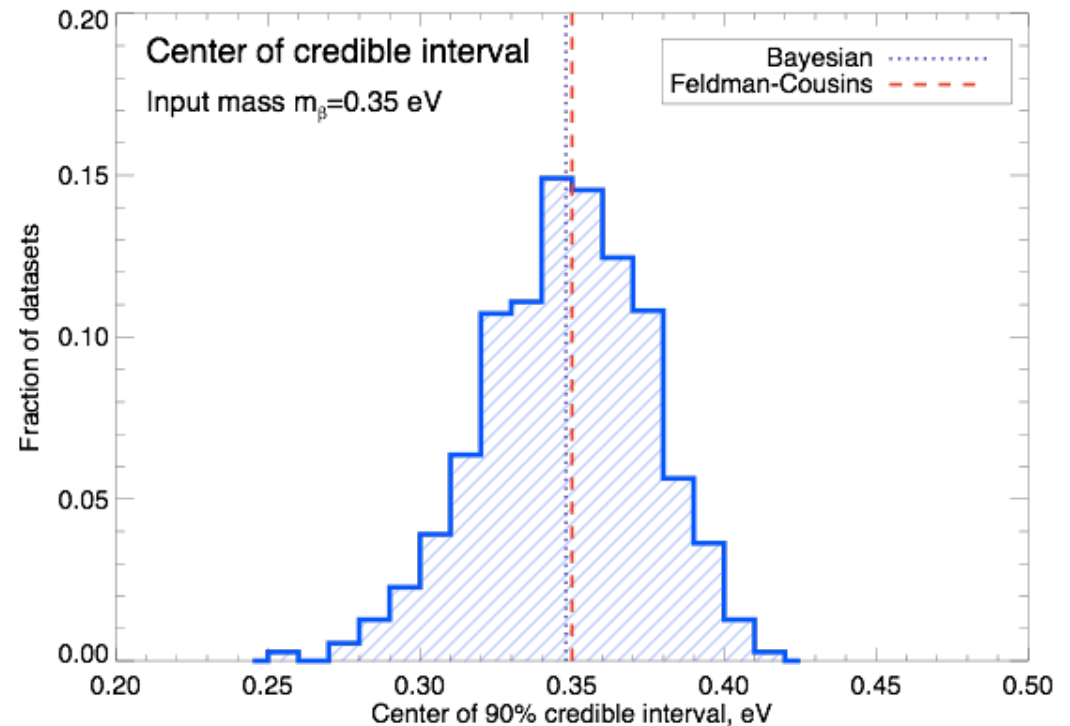
Reducing $\sigma(m_\beta^2)$ by 40%

- Find upper limit in each

$\rightarrow m_\beta < 0.17$ eV, 90%C.L.

2) Discovery potential

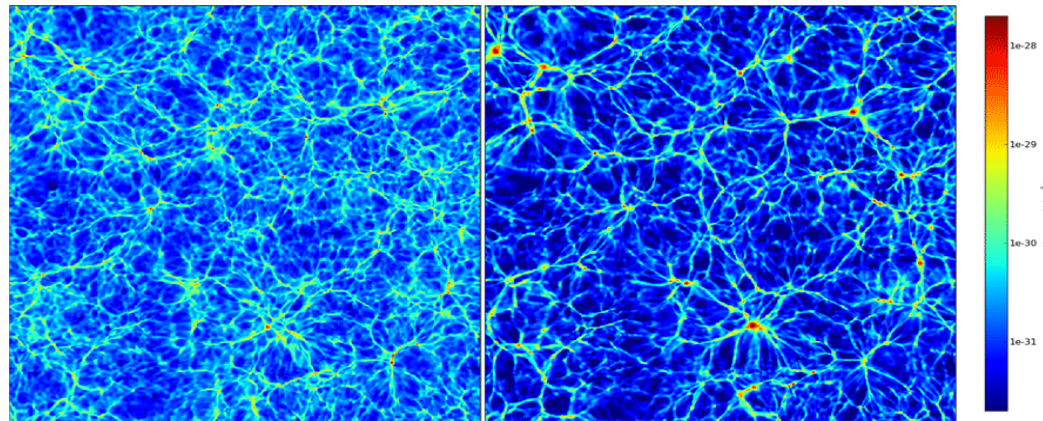
- Assume $m_\beta = 0.35$ eV
- 1000 posteriors
- Recover input regardless of analysis



Neutrino mass from galaxy surveys

700,000 galaxies with ANNz photo-z within 3.3 (Gpc/h)^3

$0.05 \text{ eV} < \text{Total neutrino mass} < 0.28 \text{ eV}$ (95% CL)



BBC News

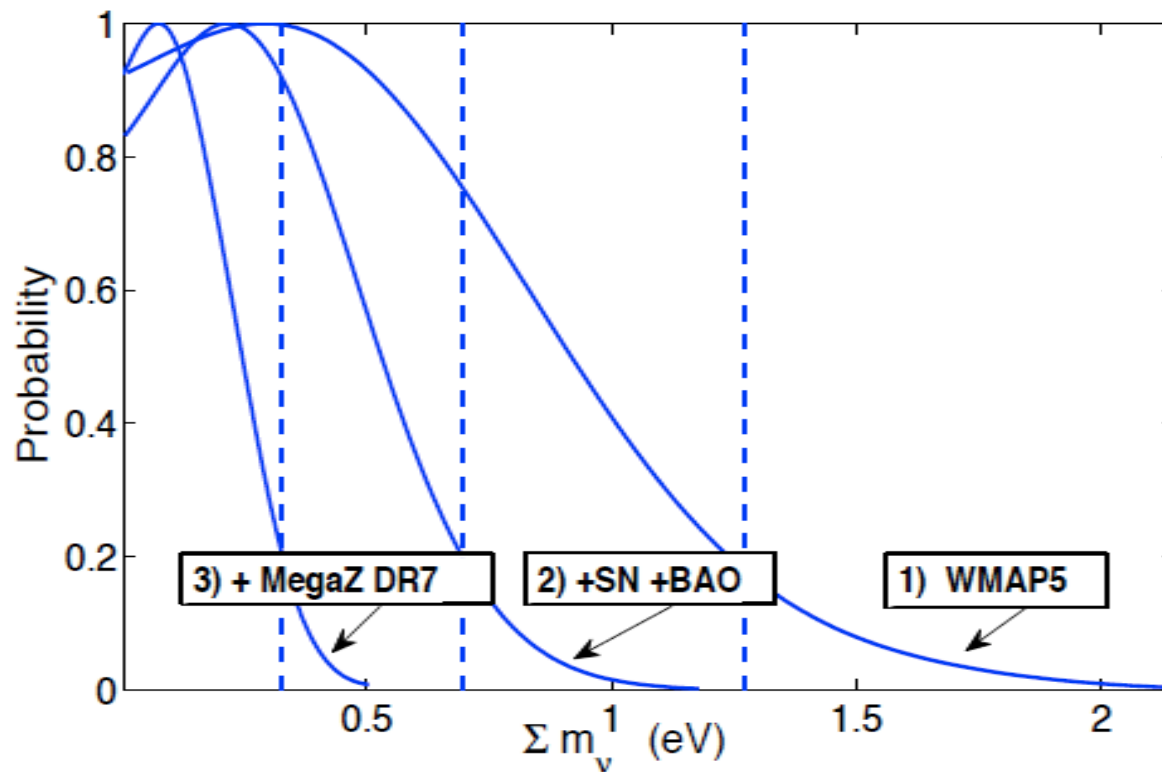
Neutrino 'ghost particle' sized up by astronomers

11:48 GMT, Tuesday, 22 June 2010 12:48 UK

Thomas, Abdalla & Lahav, PRL (2010)

Upper limits on total Neutrino mass

Total mass < 0.28 eV (95% CL)



Thomas, Abdalla & Lahav, PRL (2010) 0911.5291

Example 4:
galaxy shapes and classification

- One Million galaxies classified by 100,000 people!

Is the galaxy simply smooth and rounded, with no sign of a disk?



Smooth



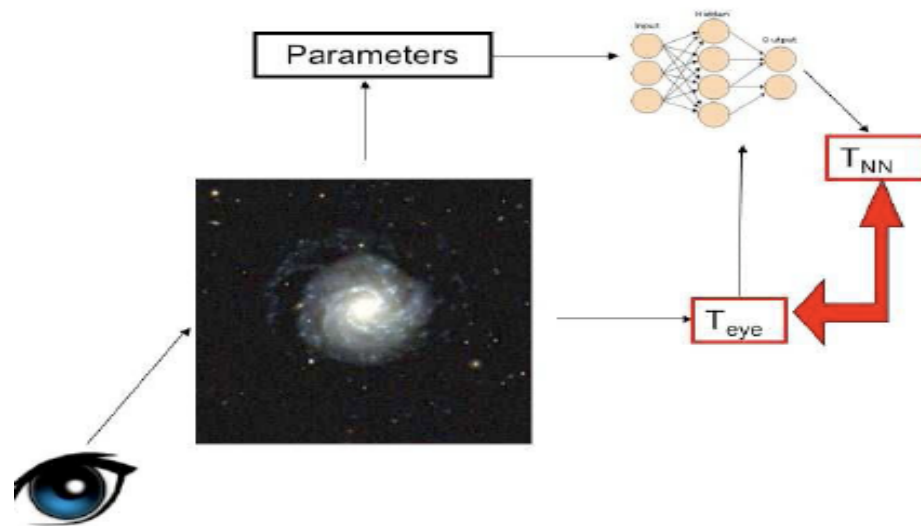
Features or disk



Star or artifact

Need help? 

Galaxy zoo and machine learning

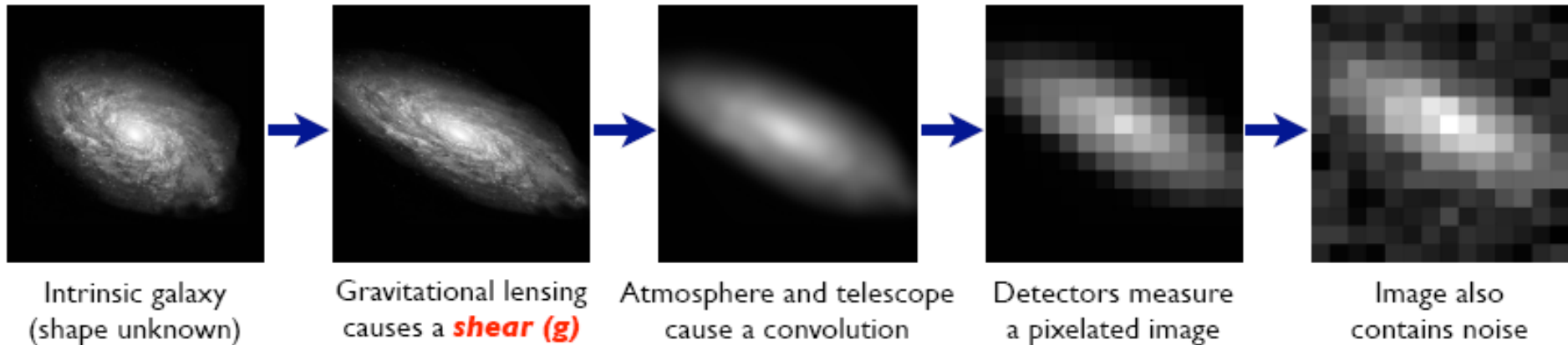


| | | GALAXY ZOO | | |
|---|------------|------------|--------|------------|
| | | Elliptical | Spiral | Star/Other |
| A | ELLIPTICAL | 91% | 0.08% | 0.5% |
| N | SPIRAL | 0.1% | 93% | 0.2% |
| N | STAR/OTHER | 0.3% | 0.3% | 96% |

Cosmic shear measurement

The Forward Process.

Galaxies: Intrinsic galaxy shapes to measured image:



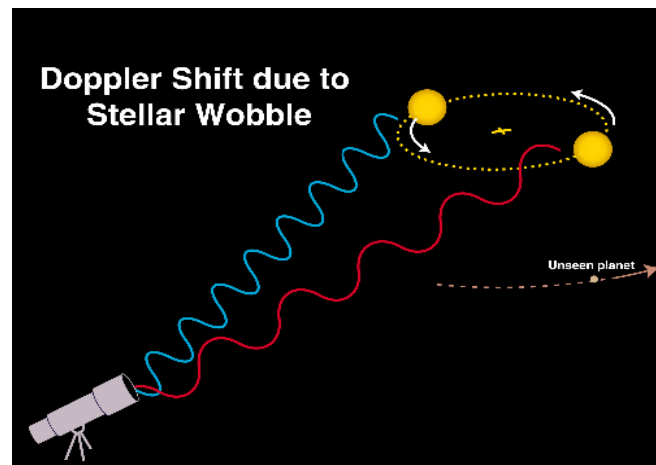
GREAT08 (Bridle et al.); GREAT10 (Kitching et al.)

Example 5: Exo-planets

Modelling orbits of planets

- (i) Given the wobble of the host star,
is there evidence for 0,1,2,3,... planets?

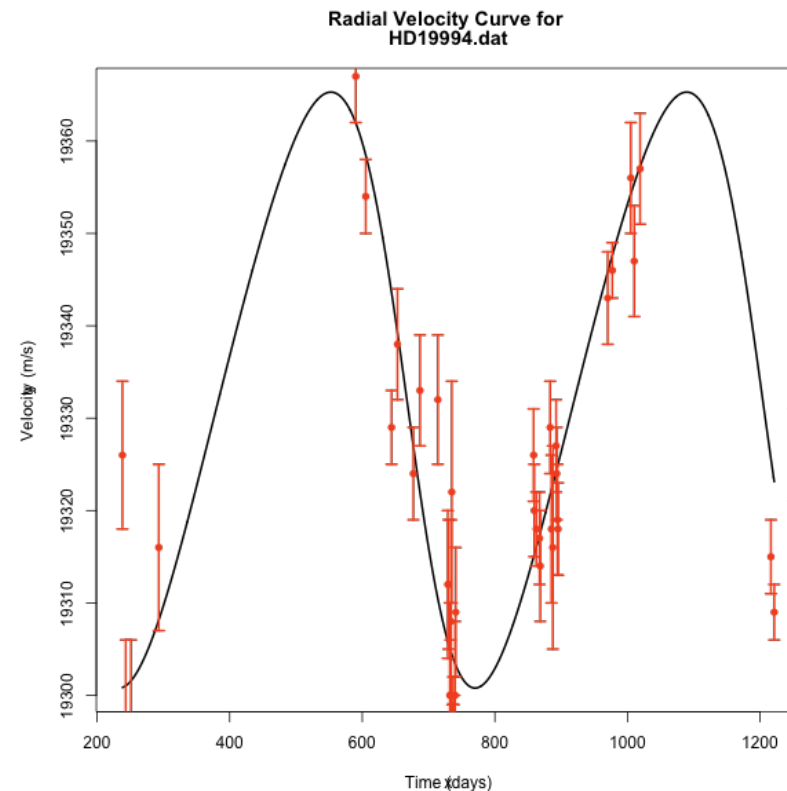
- (ii) What are the parameters and associated errors of
each detected planet?



ExoFit: orbital parameters from radial velocity data

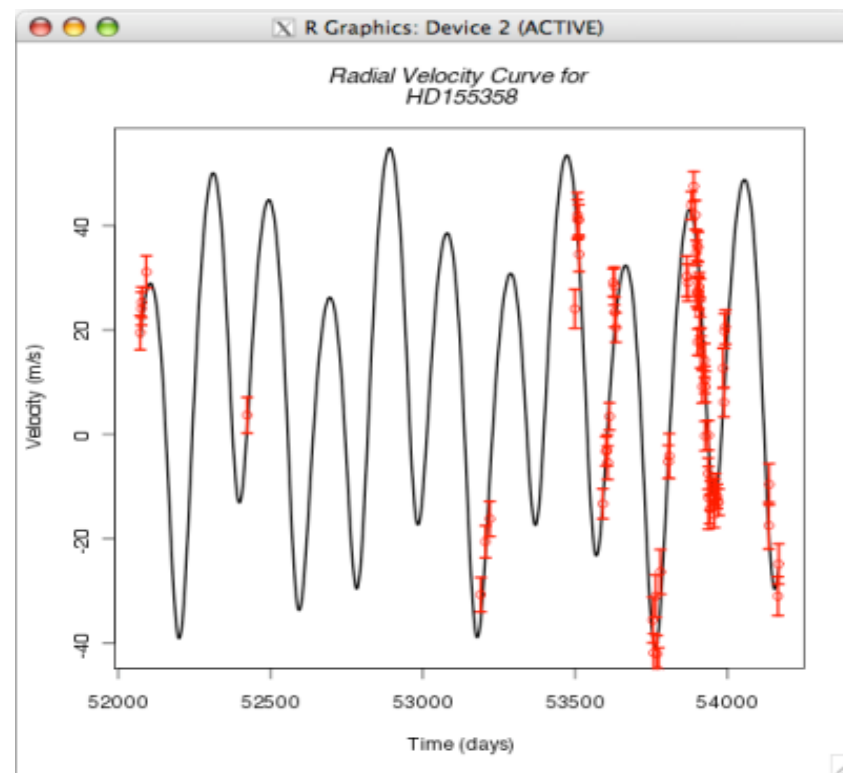
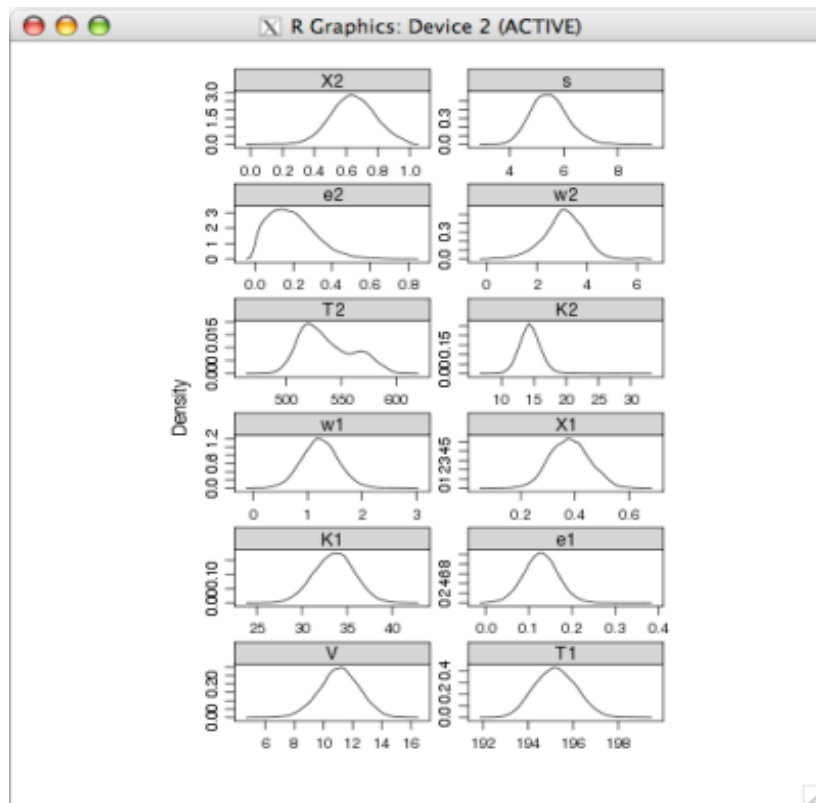
- Publically available
- Easy to use and fast
- Bayesian approach, with emphasis on priors, implemented by Markov Chain Monte Carlo (MCMC).

Balan & Lahav (MNRAS, 2009)



www.star.ucl.ac.uk/~lahav/exofit.html

HD155358: 2-planet search with ExoFit

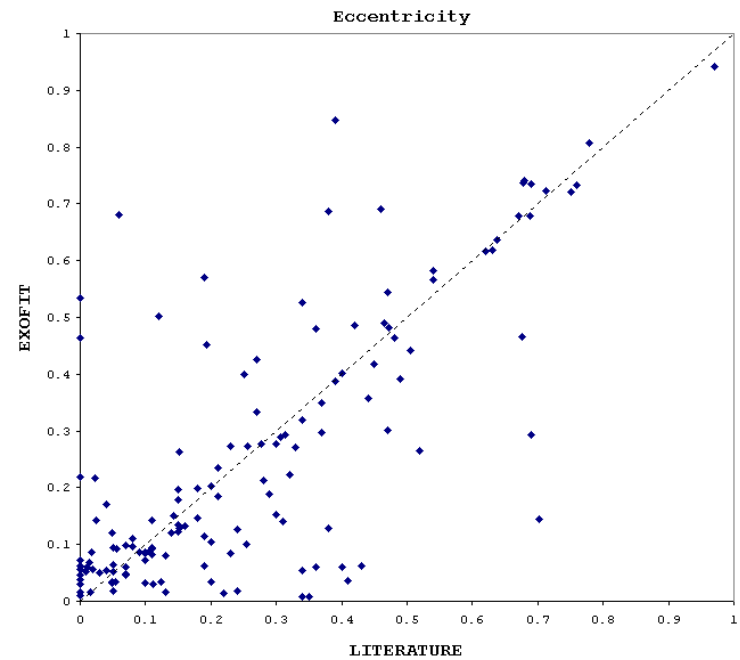
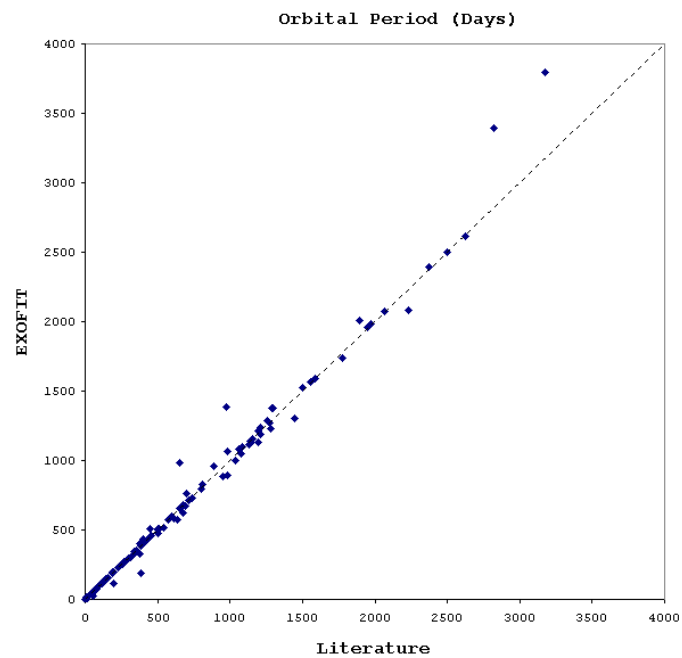


Priors for 2-planet system

| Para. | Prior | Mathematical Form | Min | Max |
|-----------------|---------------|--|-------|--------|
| $V(m s^{-1})$ | Uniform | $\frac{1}{V_{max} - V_{min}}$ | -2000 | 2000 |
| $T_1(days)$ | Jeffreys | $\frac{1}{T_1 \ln\left(\frac{T_1_{max}}{T_1_{min}}\right)}$ | 0.2 | 15000 |
| $K_1(m s^{-1})$ | Mod. Jeffreys | $\frac{(K_1 + K_{10})^{-1}}{\ln\left(\frac{K_{10} + K_{1max}}{K_{10}}\right)}$ | 0.0 | 2000 |
| e_1 | Uniform | 1 | 0 | 1 |
| ϖ_1 | Uniform | $\frac{1}{2\pi}$ | 0 | 2π |
| χ_1 | Uniform | 1 | 0 | 1 |
| $T_2(days)$ | Jeffreys | $\frac{1}{T_2 \ln\left(\frac{T_2_{max}}{T_2_{min}}\right)}$ | 0.2 | 15000 |
| $K_2(m s^{-1})$ | Mod. Jeffreys | $\frac{(K_2 + K_{20})^{-1}}{\ln\left(\frac{K_{20} + K_{2max}}{K_{20}}\right)}$ | 0.0 | 2000 |
| e_2 | Uniform | 1 | 0 | 1 |
| ϖ_2 | Uniform | $\frac{1}{2\pi}$ | 0 | 2π |
| χ_2 | Uniform | 1 | 0 | 1 |
| $s(m s^{-1})$ | Mod. Jeffreys | $\frac{(s + s_0)^{-1}}{\ln\left(\frac{s_0 + s_{max}}{s_0}\right)}$ | 0 | 2000 |

Following Gregory 05

Comparison of ExoFit vs. Literature



From a new uniformly-derived catalogue of 200 exo-planets
(Balan, Lever, Lahav, in preparation)

Summary

- Five real-life examples:

Cosmological parameters

Neutrino mass

Photometric redshifts

Galaxy shapes and classification

Exo-planets

- Future Challenges in Astro-Statistics

Further input much needed from statistics

- Model selection methodology
- MCMC machinery and extensions
- Detection of non-Gaussianity and shape finders
- Blind de-convolution (eg. PSF)
- Object classification
- Comparing simulations with data
- Visualisation
- VO technology