Constraining parameters of a disk-wind model of the BLR

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Outline

- •The Disk-wind model
- Methods
 - Light curves
 - Bayesian Formalism

Results

- Inclination angle
- Responsivity parameter
- Inclination angle & responsivity parameter
- Summary & Conclusions
- •Improvements & Future work

Disk-wind model

- A thin Keplerian disk model with a purely radial wind component (Murray et al. 1995, Chiang & Murray 1996).
- Gas velocity:

$$\boldsymbol{v} = v_r \hat{\boldsymbol{r}} + v_{\phi} \hat{\boldsymbol{\phi}}$$

where;

$$v_r \ll v_{\phi}$$

and

$$\frac{dv_r}{dr} \gtrsim \frac{dv_{\phi}}{dr}$$

$$v_{proj} = \sin i (v_r \sin \phi + v_{\phi} \cos \phi)$$



• $M_{\rm BH} \simeq 10^7 {
m M}_{\odot}$, $v_r \simeq 10^7 {
m cms}^{-1}$, $v_{igoplus} \simeq 10^8 - 10^9 {
m cms}^{-1}$

Echo images



Light curves

- Continuum light curve C(t) :
 A simulated damped random walk (DRW).
- Line light curve L(v, t): Obtained by convolving 2D echo function $\Psi(v, \tau)$ with C(t).

$$L(v,t) = \int_{0} \Psi(v,\tau)C(t-\tau)d\tau$$

Integrated line light curve L(t) :
 Obtained by summing over all velocities.



Bayesian Formalism

• From Bayes' theorem,

 $p(\theta|D, I) p(D|I) = p(D|\theta, I) p(\theta|I)$

Posterior x Evidence = Likelihood x Prior

 $p(\theta|I)$ – prior probability distribution of the model parameters θ , before observational data D.

 $p(D \mid \theta, I)$ – probability of observing data D given the model with parameters θ .

 $p(\theta|D, I)$ – posterior probability distribution of model having the parameters θ , given observed evidence D.

p(D|I) – 'model evidence' or the 'marginal likelihood'.

• Likelihood function:

$$p(D|\theta, I) = \prod_{i} \frac{1}{\sigma\sqrt{2\pi}} e^{\left[-\frac{1}{2\sigma^2} \left(D_i - m_i(\theta)\right)^2\right]}$$

- •*cos i*_{obs} : 0.05, 0.15, ..., 0.95 (total 10)
- •cos *i_{mod}* : evenly spaced in [0.00, 1.00) (total 150)
- •No overlap between *i*_{obs} and *i*_{mod}.
- •An observation i_{obs} compared with a set of 150 models (i.e. 150 i_{mod} values).
- •Repeated for 10 different *i*_{obs} values.



 Without adding error to 'observation'.



- With uncertainty added to the 'observation'.
- σ~50% of the mean value of the *'observation'*.



- •Responsivity A(r) \propto r $^{\eta}$
- • η_{obs} : -1.5, -0.5, 0.0, 0.5, 1.5 (Total 5)
- • η_{mod} : evenly spaced in [-2.0, 3.0]. (Total 50)
- •An observation η_{obs} compared with a set of 50 models (i.e. 50 η_{mod} values).
- •Repeated for 5 different η_{obs} values.





- With uncertainty added to the 'observation'.
- σ~50% of the mean value of the *'observation'*.



Results: inclination angle i & responsivity parameter η

•A grid of 143 *i* x 51 η models.

•One of the models (with uncertainty added) was selected as *'observation'* and compared with the rest of the models.





-0.5

90

40⁵⁰ i

20³⁰

3

-1

0

1

Summary & Conclusions

- •Explored the feasibility of constraining parameters of a disk-wind model of the BLR using a Bayesian statistical analysis.
- •Three different cases:
 - Inclination angle *i*.
 - Responsivity parameter η.
 - Inclination angle *i* & responsivity parameter η .
- •In all three cases we were able to recover the parameter/s of the 'observation'.

Improvements & Future work

- •Increasing the number of parameters explored.
- •For higher degree parameter space exploration, a more efficient method will be needed.
 - E.g. Nested sampling
- •Applying this method to real observational data.