

# CHASING LAMBDA

Aleksandra Kurek \*, Marek Szydlowski \* °

\* Astronomical Observatory, Jagiellonian University, Poland

°Complex System Research Centre, Jagiellonian University,  
Poland

# MODELS WITH DARK ENERGY

- |   |   |   |
|---|---|---|
| 1 | <p><math>\Lambda</math>CDM model<br/> <math>p_X = -\rho_X</math> (Weinberg 1989)</p>  | $H^2 = H_0^2 \{ \Omega_{m,0}(1+z)^3 + (1 - \Omega_{m,0}) \}$  |
| 2 | <p>Model with generalized Chaplygin gas<br/> <math>p_X = -\frac{A}{\rho_X^\alpha}</math> <math>A &gt; 0</math> and <math>\alpha = \text{const}</math></p>   | $H^2 = H_0^2 \left\{ \Omega_{m,0}(1+z)^3 + (1 - \Omega_{m,0}) [A_S + (1 - A_S)(1+z)^{3(1+\alpha)}]^{1+\alpha} \right\}$         |
| 3 | <p>Model with phantom dark energy<br/> <math>p_X = w_X \rho_X</math> <math>w_X &lt; -1</math> (Caldwell 2002)</p>   | $H^2 = H_0^2 \{ \Omega_{m,0}(1+z)^3 + (1 - \Omega_{m,0})(1+z)^{3(1+w_X)} \}$  |
| 4 | <p>Model with dynamical E.Q.S<br/> <math>w(a) = w_0 + w_1(1-a)</math> (Chevallier &amp; Polarski 2001)</p>  | $H^2 = H_0^2 \left\{ \Omega_{m,0}(1+z)^3 + (1 - \Omega_{m,0})(1+z)^{3(w_0+w_1+1)} \exp\left[-\frac{3w_1z}{1+z}\right] \right\}$ |
| 5 | <p>Quintessence model<br/> <math>\rho_X = \rho_{X0} a^{-3(1+\bar{w}_X(a))}</math> (Peebles &amp; Ratra 1988)<br/> <math>\bar{w}_X(a) = \frac{\int w_X(a) d \ln a}{\int d(\ln a)} \equiv w_0 a^\alpha</math> (Rahvar &amp; Movahed 2007)</p> | $H^2 = H_0^2 \left\{ \Omega_{m,0}(1+z)^3 + (1 - \Omega_{m,0})(1+z)^{3(1+w_0(1+z)^{-\alpha})} \right\}$                          |

mean of the coefficient of the EQS  
in the log scale factor

# MODELS WITH MODIFIED THEORY OF GRAVITY

- 6 DGP model  $H^2 = H_0^2 \left\{ \left[ \sqrt{\Omega_{m,0}(1+z)^3 + \Omega_{rc,0}} + \sqrt{\Omega_{rc,0}} \right]^2 \right\}$   
(Dvali et al. 2000)  $\Omega_{rc,0} = \frac{(1-\Omega_{m,0})^2}{4}$
- 7  $\Lambda$ CDM model  $H^2 = H_0^2 \{ \Omega_{m,0}(1+z)^3 - \Omega_{n,0}(1+z)^n + 1 - \Omega_{m,0} + \Omega_{n,0} \}$  (Singh & Vandersloot 2005)
- 8 interacting model with  $\Lambda$   $H^2 = H_0^2 \{ \Omega_{m,0}(1+z)^3 + \Omega_{int,0}(1+z)^n + 1 - \Omega_{m,0} - \Omega_{int,0} \}$  (Szydlowski et al. 2006)
- 9 Cardassian model  $H^2 = H_0^2 \left\{ \Omega_{r,0}(1+z)^4 + \Omega_{m,0}(1+z)^4 \left[ \frac{1}{1+z} + (1+z)^{-4+4n} \left( \frac{1-\Omega_{r,0}-\Omega_{m,0}}{\Omega_{m,0}} \right) \left( \frac{\frac{1}{1+z} + \frac{\Omega_{r,0}}{\Omega_{m,0}}}{1 + \frac{\Omega_{r,0}}{\Omega_{m,0}}} \right)^n \right] \right\}$   
 $\Omega_{r,0} = 10^{-4}$ ;  $3H^2 = \rho + B\rho^n$  (Freese & Lewis 2002)
- 10 Sahni-Shtanov brane I model  $H^2 = H_0^2 \left\{ \Omega_{m,0}(1+z)^3 + \Omega_{\sigma,0} + 2\Omega_{l,0} - 2\sqrt{\Omega_{l,0}}\sqrt{\Omega_{m,0}(1+z)^3 + \Omega_{\sigma,0} + \Omega_{l,0} + \Omega_{\Lambda b,0}} \right\}$   
(Shtanov 2000)  $\Omega_{\sigma,0} = 1 - \Omega_{m,0} + 2\sqrt{\Omega_{l,0}}\sqrt{1 + \Omega_{\Lambda b,0}}$

# BAYESIAN FRAMEWORK OF MODEL SELECTION

The best one - the greatest value of the POSTERIOR PROBABILITY defined as

$$P(M_i|D) = \frac{P(D|M_i)P(M_i)}{P(D)}$$

$P(D)$  - normalization constant:

$$\sum_{i=1}^K P(M_i|D) = 1 \longrightarrow P(D) = \sum_{i=1}^K P(D|M_i)P(M_i)$$

- conclusions depend on the set of models under consideration

$P(M_i)$  - prior probability for i-th model

- value depends on our previous information (knowledge) about the models

# BAYESIAN FRAMEWORK OF MODEL SELECTION

$P(D|M_i)$  - marginal likelihood (also called evidence):

$$P(D|M_i) = \int L(\bar{\theta}_i|D, M_i) P(\bar{\theta}_i|M_i) d\bar{\theta}_i \equiv E_i$$

likelihood of the model      vector of model parameters      prior probability for model parameters

approximation to  $-2 \ln E$        $\longrightarrow$

$$BIC = -2 \ln \mathcal{L} + d \ln N$$

maximum likelihood      number of model parameters      number of data points

assumptions       $\longrightarrow$

1. iid
2.  $f(x_i|\bar{\theta}) = \exp \left[ \sum_{k=1}^S w_k(\bar{\theta}) t_k(x_i) + b(\bar{\theta}) \right]$
3. prior for model parameters  $\neq 0$  in the maximum likelihood
4. sample size large with respect to the number of model parameters

Schwarz 1978

1. prior for model parameters  $\neq 0$  in the neighborhood of the maximum likelihood
2. prior is bound in the whole parameter space
3. sample size large with respect to the number of model parameters

Cavanaugh & Neath 1999

# BAYESIAN FRAMEWORK OF MODEL SELECTION

POSTERIOR ODDS

$$O_{12} \equiv \frac{P(M_1|D)}{P(M_2|D)} = \frac{P(M_1)}{P(M_2)} \frac{P(D|M_1)}{P(D|M_2)}$$

BAYES FACTOR -  $B_{12}$

<b>2 ln <math>B_{12}</math></b>	<b>evidence in favor model 1</b>
<0,2)	not worth than a bare mention
<2,6)	positive
<6,10)	strong
>10	very strong

(Kass & Raftery 1995)

# APPLICATION TO COSMOLOGICAL MODELS COMPARISON

1. SN Ia:  $N = 192$  - 60 ESSENCE; 57 SNLS; 30 HST; 45 local sample  
(Riess et al. 2007; Wood-Vasey et al. 2007; Davis et al. 2007)

$$L_{\text{SN}} \propto \exp \left[ -\frac{1}{2} \left( \sum_{i=1}^N \frac{(\mu_i^{\text{theor}} - \mu_i^{\text{obs}})^2}{\sigma_i^2} \right) \right],$$

$$\mu_i^{\text{obs}} = m_i - M;$$

$$\mu_i^{\text{theor}} = 5 \log_{10} D_{Li} - 5 \log_{10} H_0 + 25;$$

$$D_{Li} = H_0 d_{Li} = H_0 (1 + z_i) c \int_0^{z_i} \frac{dz'}{H(z')}.$$

2. CMB

$$L_R \propto \exp \left[ -\frac{(R^{\text{theor}} - R^{\text{obs}})^2}{2\sigma_R^2} \right],$$

$$R^{\text{theor}} = \sqrt{\Omega_{m,0}} \int_0^{z_{\text{dec}}} \frac{H_0}{H(z)} dz,$$

$$R^{\text{obs}} = 1.70 \pm 0.03 \quad z_{\text{dec}} = 1089$$

(Spergel et al. 2006; Wang & Mukherjee 2006)

# APPLICATION TO COSMOLOGICAL MODELS COMPARISON

## 3. BAO

$$L_A \propto \exp \left[ -\frac{(A^{\text{theor}} - A^{\text{obs}})^2}{2\sigma_A^2} \right],$$

$$A^{\text{theor}} = \sqrt{\Omega_{m,0}} \left( \frac{H(z)}{H_0} \right)^{-\frac{1}{3}} \left[ \frac{1}{z_A} \int_0^{z_A} \frac{H_0}{H(z)} \right]^{\frac{2}{3}},$$

$$A^{\text{obs}} = 0.469 \pm 0.017 \quad z_A = 0.35$$

(Eisenstein et al. 2005)

## 4. $H(z)$ : $N=9$ (differential ages $\frac{dt}{dz}$ of the passively evolving galaxies)

$$L_H \propto \exp \left( -\frac{1}{2} \left[ \sum_{i=1}^N \frac{(H(z_i) - H_i(z_i))^2}{\sigma_i^2} \right] \right).$$

$$H(z) \equiv \frac{\dot{a}}{a} = -\frac{1}{1+z} \frac{dz}{dt}.$$

(Simon et al. 2005)

$$L \propto L_{\text{sn}} L_R L_A L_H$$

$$N=192+1+1+9$$



# POSTERIOR PROBABILITIES FOR MODELS WITH DARK ENERGY

1	$\Lambda$ CDM model	<b>0.84</b>
2	Model with generalized Chaplygin gas	0.02
3	Model with phantom dark energy	0.06
4	Model with dynamical E.Q.S	0.04
5	Quintessence model	0.04

# POSTERIOR PROBABILITIES FOR MODELS WITH MODIFIED GRAVITY

6	DGP model	0.07
7	$\Lambda$ CDM model	0.03
8	interacting model with $\Lambda$	0.13
9	Cardassian model $\Omega_{r,0} = 10^{-4}$	<b>0.74</b>
10	Sahni-Shtanov brane I model	0.03

# POSTERIOR PROBABILITIES FOR ALL MODELS

1	$\Lambda$ CDM model	<b>0.74</b>
2	Model with generalized Chaplygin gas	0.02
3	Model with phantom dark energy	0.05
4	Model with dynamical E.Q.S	0.04
5	Quintessence model	0.03
6	DGP model	0.01
7	B $\Lambda$ CDM model	0.005
8	interacting model with $\Lambda$	0.01
9	Cardassian model $\Omega_{r,0} = 10^{-4}$	0.09
10	Sahni-Shtanov brane I model	0.005

# CONCLUSIONS

In the light of data used in analysis:

- $\Lambda$ CDM model is the best one from the set of models with dark energy as well as the best one from the set of all models considered
- Cardassian model is the best one from the set of models with modified gravity