

2011 Cross Strait Meeting on Particle Physics and Cosmology

# Observational Constraints on Exponential Gravity

Phys. Rev. D 82, 103515 (2010)

Ling-Wei Luo

Collaborator: Louis Yang, Chung-Chi Lee, Chao-Qiang Geng



*Department of Physics, National Tsing Hua University*

April 1st, 2011

# Outline

- 1 Introduction
- 2 Observational Constraints
- 3 Results
- 4 Summary



# Outline

**1** Introduction

2 Observational Constraints

3 Results

4 Summary



- Alternative dark energy models
  - Modified matter
    - Quintessence
    - K-essence
    - Perfect fluid models
  - Modified gravity
    - $f(R)$  gravity (non-linear Lagrangian density in terms of  $R$ )
    - Scalar-tensor theories ( $R$  couples to  $\phi$  with the form:  $F(\phi)R$ )
    - Braneworld models
- Others



- $f(R)$  models:

$$S_{\text{E-H}} = \frac{1}{16\pi G} \int d^4x \sqrt{-g} R \rightarrow S = \frac{1}{16\pi G} \int d^4x \sqrt{-g} R + f(R).$$

- Hu and Sawicki

$$f(R) = -\mu R_c \frac{(R/R_c)^{2n}}{(R/R_c)^{2n} + 1} \text{ with } n > 0 \text{ and } R_c > 0,$$

- Starobinsky

$$f(R) = -\mu R_c [1 - (1 + R^2/R_c^2)^{-n}] \text{ with } n > 0 \text{ and } R_c > 0,$$

- Tsujikawa

$$f(R) = -\mu R_c \tanh(R/R_c) \text{ with } R_c > 0.$$

- Exponential gravity

$$f(R) = -\beta R_s (1 - e^{-R/R_s})$$

Goal: we will test these models with the the observational data (SNe Ia, BAO and CMB).



- Exponential gravity

- The action is

$$S = \frac{1}{2\kappa^2} \int d^4x \sqrt{-g} [R + f(R)] + S_m,$$

where  $\kappa^2 \equiv 8\pi G$  and

$$f(R) = -\beta R_s (1 - e^{-R/R_s}).$$

- The modified Friedmann equation is

$$H^2 = \frac{\kappa^2 \rho_M}{3} + \frac{1}{6} (f_R R - f) - H^2 (f_R + f_{RR} R'),$$

where a subscript R denotes the derivative with respect to R, a prime represents  $d/d \ln a$ , and  $\rho_M = \rho_m + \rho_r$ .

- In flat spacetime, the Ricci scalar is

$$R = 12H^2 + 6HH'.$$



## ■ Exponential gravity

- Following **Hu and Sawicki's parameterization**

$$y_H \equiv \frac{\rho_{DE}}{\rho_m^0} = \frac{H^2}{m^2} - a^{-3} \text{ and } y_R \equiv \frac{R}{m^2} - 3a^{-3},$$

with  $m^2 \equiv \kappa^2 \rho_m^0 / 3$ , we rewrite the modified Friedmann equation and Ricci scalar into two coupled differential equations

$$y'_H = \frac{y_R}{3} - 4y_H$$

and

$$y'_R = 9a^{-3} - \frac{1}{H^2 f_{RR}} \left[ y_H + f_R \left( \frac{H^2}{m^2} - \frac{R}{6m^2} \right) + \frac{f}{6m^2} \right],$$

- Leading to a second order differential equation of  $y_H$  in the form

$$y''_H + J_1 y'_H + J_2 y_H + J_3 = 0.$$

- And the effective dark energy equation of state  $w_{DE}$  is given by

$$w_{DE} = -1 - \frac{y'_H}{3y_H}.$$



# Outline

1 Introduction

**2 Observational Constraints**

3 Results

4 Summary





- Type Ia Supernovae (SNe Ia)

- The distance modulus

$$\mu_{th}(z_i) \equiv 5 \log_{10} D_L(z_i) + \mu_0,$$

where  $\mu_0 \equiv 42.38 - 5 \log_{10} h$  with  $H_0 = h \cdot 100 \text{ km/s/Mpc}$ .

- The Hubble-free luminosity distance for the flat universe

$$D_L(z) = (1+z) \int_0^z \frac{dz'}{E(z')},$$

where  $E(z) = H(z)/H_0$ .

- The  $\chi^2$  for the SNe Ia data is

$$\chi_{SN}^2 = \sum_i \frac{[\mu_{obs}(z_i) - \mu_{th}(z_i)]^2}{\sigma_i^2}.$$



## ■ Type Ia Supernovae (SNe Ia)

- Since the absolute magnitude of SNe Ia is unknown, we should minimize  $\chi_{SN}^2$  with respect to  $\mu_0$ , which relates to the absolute magnitude, and expand it to be

$$\chi_{SN}^2 = A - 2\mu_0 B + \mu_0^2 C,$$

where

$$A = \sum_i \frac{[\mu_{obs}(z_i) - \mu_{th}(z_i; \mu_0 = 0)]^2}{\sigma_i^2},$$
$$B = \sum_i \frac{\mu_{obs}(z_i) - \mu_{th}(z_i; \mu_0 = 0)}{\sigma_i^2}, \quad C = \sum_i \frac{1}{\sigma_i^2}.$$

The minimum of  $\chi_{SN}^2$  with respect to  $\mu_0$  is

$$\boxed{\tilde{\chi}_{SN}^2 = A - \frac{B^2}{C}}.$$



## ■ Baryon Acoustic Oscillations (BAO)

- The distance ratio

$$d_z \equiv r_s(z_d)/D_V(z) \quad (z_d \text{ is redshift at the drag epoch.})$$

- The volume-averaged distance

$$D_V(z) \equiv \left[ (1+z)^2 D_A^2(z) \frac{z}{H(z)} \right]^{1/3}.$$

- $D_A(z)$  is the proper angular diameter distance

$$D_A(z) = \frac{1}{1+z} \int_0^z \frac{dz'}{H(z')}, \quad (\text{for flat universe}).$$

- The comoving sound horizon

$$r_s(z) = \frac{1}{\sqrt{3}} \int_0^{1/(1+z)} \frac{da}{a^2 H(z'=\frac{1}{a}-1) \sqrt{1 + (3\Omega_b^0/4\Omega_\gamma^0)a}},$$

here  $\Omega_b^0 = 0.022765h^{-2}$  and  $\Omega_\gamma^0 = 2.469 \times 10^{-5}h^{-2}$ .



## ■ Baryon Acoustic Oscillations (BAO)

- $z_d$  is the redshift at the drag epoch

$$z_d = \frac{1291(\Omega_m^0 h^2)^{0.251}}{1 + 0.659(\Omega_m^0 h^2)^{0.828}} [1 + b_1(\Omega_b^0 h^2)^{b_2}],$$

with

$$b_1 = 0.313(\Omega_m^0 h^2)^{-0.419} [1 + 0.607(\Omega_m^0 h^2)^{0.674}],$$

$$b_2 = 0.238(\Omega_m^0 h^2)^{0.223}.$$

- The measured distance ratios  $d_{z=0.2}^{obs} = 0.1905 \pm 0.0061$  and  $d_{z=0.35}^{obs} = 0.1097 \pm 0.0036$  with the inverse covariance matrix:

$$C_{BAO}^{-1} = \begin{pmatrix} 30124 & -17227 \\ -17227 & 86977 \end{pmatrix}.$$

- The  $\chi^2$  for the BAO data is

$$\chi_{BAO}^2 = (x_{i,BAO}^{th} - x_{i,BAO}^{obs})(C_{BAO}^{-1})_{ij}(x_{j,BAO}^{th} - x_{j,BAO}^{obs}),$$

where  $x_{i,BAO} \equiv (d_{0.2}, d_{0.35})$ .



## ■ Cosmic Microwave Background (CMB)

- The acoustic scale

$$l_A(z_*) \equiv (1 + z_*) \frac{\pi D_A(z_*)}{r_S(z_*)}.$$

- The shift parameter

$$R(z_*) \equiv \sqrt{\Omega_m^0} H_0 (1 + z_*) D_A(z_*).$$

- The decoupling epoch

$$z_* = 1048 [1 + 0.00124(\Omega_b^0 h^2)^{-0.738}] [1 + g_1(\Omega_m^0 h^2)^{g_2}],$$

where

$$g_1 = \frac{0.0783(\Omega_b^0 h^2)^{-0.238}}{1 + 39.5(\Omega_b^0 h^2)^{0.763}}, \quad g_2 = \frac{0.560}{1 + 21.1(\Omega_b^0 h^2)^{1.81}}.$$



## ■ Cosmic Microwave Background (CMB)

- $l_A(z_*) = 302.09$ ,  $R(z_*) = 1.725$  and  $z_* = 1091.3$  with the inverse covariance matrix:

$$C_{CMB}^{-1} = \begin{pmatrix} 2.305 & 29.698 & -1.333 \\ 29.698 & 6825.27 & -113.180 \\ -1.333 & -113.180 & 3.414 \end{pmatrix}.$$

- The  $\chi^2$  for the CMB data is

$$\chi_{CMB}^2 = (x_{i,CMB}^{th} - x_{i,CMB}^{obs})(C_{CMB}^{-1})_{ij}(x_{j,CMB}^{th} - x_{j,CMB}^{obs}),$$

where  $x_{i,CMB} \equiv (l_A(z_*), R(z_*), z_*)$ .



# Outline

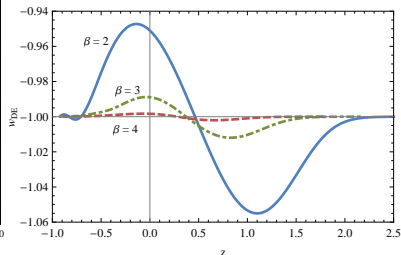
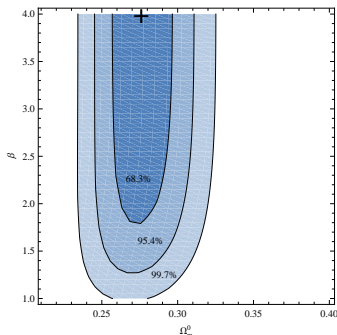
- 1 Introduction
- 2 Observational Constraints
- 3 Results**
- 4 Summary



■ The  $\chi^2 = \tilde{\chi}_{SN}^2 + \chi_{BAO}^2 + \chi_{CMB}^2$

Exponential Gravity:  $R + f(R) = R - \beta R_s(1 - e^{-R/R_s})$

Model		$\Omega_m^0$	$\chi^2$
Exponential Gravity	$\beta = 2$	$0.274^{+0.014}_{-0.013}$	546.7136
	$\beta = 3$	$0.276^{+0.014}_{-0.013}$	545.3836
	$\beta = 4$	$0.276^{+0.014}_{-0.013}$	545.1721
$\Lambda$ CDM		$0.276^{+0.014}_{-0.013}$	545.1522





- We have a lower bound of parameter  $\beta$  at 1.27 but **no upper limit**
- $\Omega_m^0$  is constrained to the concordance value
- $\beta \rightarrow \infty$  corresponds to the  **$\Lambda$ CDM model**
- From the plot of effective dark energy equation of state  $w_{DE}$ , the deviation from cosmological constant phase ( $w_{DE} = -1$ ) become smaller for larger value of  $\beta$



# Outline

- 1 Introduction
- 2 Observational Constraints
- 3 Results
- 4 Summary**



- We have studied the modified gravity, especially interested on the exponential gravity
- We have done the data fitting by using the SNe Ia, BAO and CMB data
- In the low redshift regime, we follow Hu and Sawicki's parameterization to form the differential equation for the exponential gravity and solve it numerically
- In the high redshift regime, we take advantage of the asymptotic behavior of the exponential gravity toward an effective cosmological constant
- Current observational data can **not distinguish** between the  $\Lambda$ CDM and exponential gravity models



# End

- Thank you!



# Outline

## **5** Backup slides: Statistical Methods



- The method of maximum likelihood - likelihood function
  - A set of  $N$  measure quantities  $\mathbf{x} = (x_1, \dots, x_N)$  describe by a joint p.d.f.  $f(\mathbf{x}; \boldsymbol{\theta})$ , where  $\boldsymbol{\theta} = (\theta_1, \dots, \theta_n)$  is a set of  $n$  parameters whose value are unknown.
  - The *likelihood function*  $\mathcal{L}(\boldsymbol{\theta}) \equiv f(\mathbf{x}; \boldsymbol{\theta})$ .
  - If the measurements  $x_i$  are statistically independent and each follow the p.d.f.  $f(x_i; \boldsymbol{\theta})$ , then the joint p.d.f. for  $x$  factorizes and the likelihood function is

$$\mathcal{L}(\boldsymbol{\theta}) = \prod_{i=1}^N f(x_i; \boldsymbol{\theta}) .$$

- It is usually easier to work with  $\ln \mathcal{L}$ , and since both are maximized for the same parameter value  $\boldsymbol{\theta}$ , the maximum likelihood (ML) estimators can be found by solving the *likelihood equations*,

$$\frac{\partial \ln \mathcal{L}}{\partial \theta_i} = 0 , \quad i = 1, \dots, n .$$



- The method of least squares -  $\chi^2$  function
  - The *method of least squares* (LS) coincides with the method of maximum likelihood in the following special case.
  - Consider a set of  $N$  independent measurements  $y_i$  at known points  $x_i$ . The measurement  $y_i$  is assumed to be Gaussian distributed with mean  $F(x_i; \theta)$  and known variance  $\sigma_i^2$ .
  - The goal is to construct estimators for the unknown parameters  $\theta$ ,

$$\chi^2(\theta) = -2 \ln \mathcal{L}(\theta) + \text{const} = \sum_{i=1}^N \frac{(y_i - F(x_i; \theta))^2}{\sigma_i^2} + \text{const} .$$

- The set of parameters  $\theta$  which maximize  $\mathcal{L}$  is the same as those which minimize  $\chi^2$ .



- The method of least squares -  $\chi^2$  function
  - In general, the measurements  $y_i$  are not Gaussian distributed as long as they are not independent, If they are not independent but rather have a covariance matrix  $V_{ij} = \text{cov}[y_i, y_j]$ , then the LS estimators are determined by the minimum of

$$\chi^2(\boldsymbol{\theta}) = (\mathbf{y} - \mathbf{F}(\boldsymbol{\theta}))^T \mathbf{V}^{-1} (\mathbf{y} - \mathbf{F}(\boldsymbol{\theta})) .$$

where  $\mathbf{y} = (y_1, \dots, y_N)$  is the vector of measurements,  $\mathbf{F}(\boldsymbol{\theta})$  is the corresponding vector of predicted values.

- Best-fit
 

Small value of  $\chi^2$  indicate a good fit. The parameters  $\boldsymbol{\theta}^*$  that minimize  $\chi^2$  are called the best-fit parameters.





## ■ Confidence Level

**Table 32.2:**  $\Delta\chi^2$  or  $2\Delta\ln L$  corresponding to a coverage probability  $1 - \alpha$  in the large data sample limit, for joint estimation of  $m$  parameters.

$(1 - \alpha)$ (%)	$m = 1$	$m = 2$	$m = 3$
68.27	1.00	2.30	3.53
90.	2.71	4.61	6.25
95.	3.84	5.99	7.82
95.45	4.00	6.18	8.03
99.	6.63	9.21	11.34
99.73	9.00	11.83	14.16

PDG2008

