Modified gravity simulations

On the road to build reliable emulators for modified gravity models

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05.05.22, Marseille

ICC, Durham University







European Research Council Established by the European Commission



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Modified gravity models: Effect on the matter power spectrum



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Modified Gravity Simulation Codes

Classic codes	Fast / approximate codes	Emulators
ECOSMOG f(R), nDGP, Galileon	MG-GLAM f(R), nDGP,	FORGE f(R)
^{ISIS} f(R), Symmetron	MG-HALOFIT/CAMB $f(R)$	REACT f(R), nDGP
MG-GADGET $f(R)(+\nu)$	^{MG-COLA} f(R), DGP, Symmetron, JBD	bridge nDGP
AREPO f(R), nDGP (+ ν)	MG-EVOLUTION f(R), nDGP	

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- slow and computationally expensive if many runs are required

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Hassani&Lombriser(2020)

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BRIDGE (preliminary)



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$$S = \int \mathrm{d}^4 x \sqrt{-g} \left[\frac{R + f(R)}{16\pi G} + \mathcal{L}_m \right]$$

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Newtonian limit:

 $\nabla^2 \Phi = \frac{4}{3} \times 4\pi G \delta \rho - \frac{1}{6} \delta R \qquad \nabla^2 f_R = \frac{1}{3} \left(\delta R - 8\pi G \delta \rho \right)$

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$$f(R) = -m^2 \frac{c_1 \left(\frac{R}{m^2}\right)^n}{c_2 \left(\frac{R}{m^2}\right)^n + 1} \qquad m^2 \equiv H_0^2 \Omega_m$$

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3D matter power spectrum emulator



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Emulator cross validation



Emulator cross validation



Making a prediction for independent simulations and comparing to other emulators



 BRIDGE - nDGP emulator (Cesar Hernandez-Aguayo, Carol Cuesta-Lazaro)

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- Larger simulation suites for training: MG-GLAM



Fast full N-body simulations of generic modified gravity: conformal coupling models

Cheng-Zong Ruan,^a César Hernández-Aguayo,^{b,c} Baojiu Li,^a Christian Arnold,^a Carlton M. Baugh,^a Anatoly Klypin,^d and Francisco Prada^e

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- f(R) run with 2048³ particles in a 1 Gpc/h box with a 4096³ grid takes ≈ 27 h on one cosma8 node (128 cores)

MG-GLAM - accuracy

 $\label{eq:compare} \begin{array}{l} \mbox{Compare} \approx 500 \mbox{Mpc}/h \mbox{ boxes with } 1024^3 \mbox{ particles of } \mbox{MG-GLAM} \\ \mbox{(2048^3 grid) and } \mbox{AREPO} \end{array}$



MG-GLAM - power spectrum



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