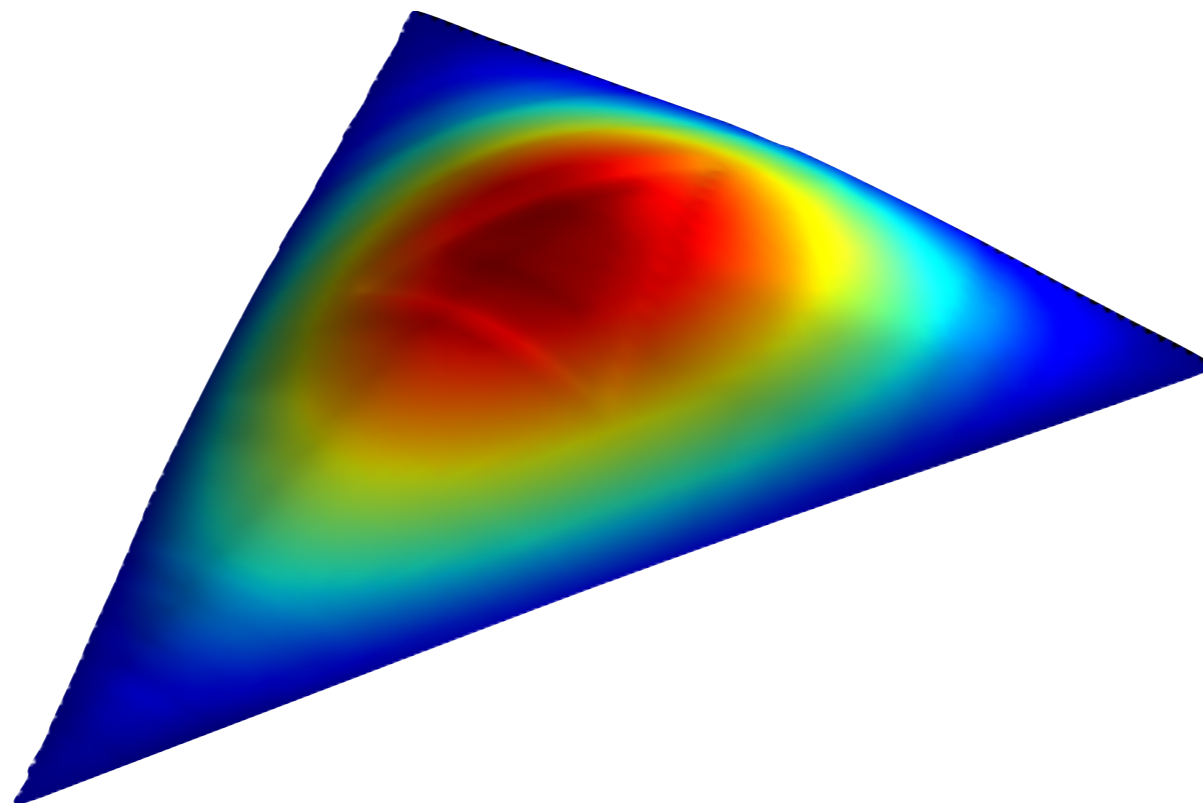


Numerically calculating observables from inflation and reheating: PyTransport and beyond

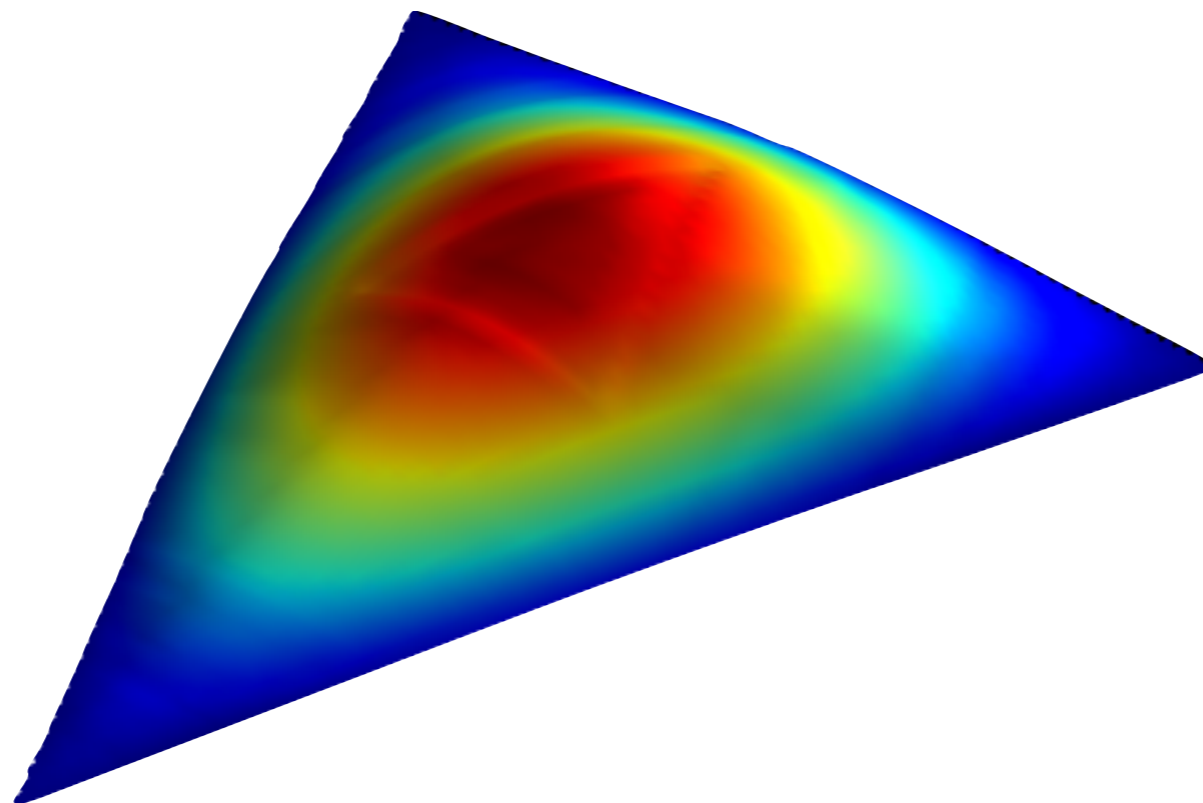


David Mulryne

Queen Mary University of London



Numerically calculating observables from inflation and reheating: PyTransport and beyond



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Collaborators

Transport collaborators:

D Seery, M Dias, J Frazer, J Ronayne

arXiv:1609.00379; arXiv:1708.07130 + ongoing

Visit TransportMethod.com for more information

Non-perturbative reheating collaborations:

S Imrith, A Rajantie

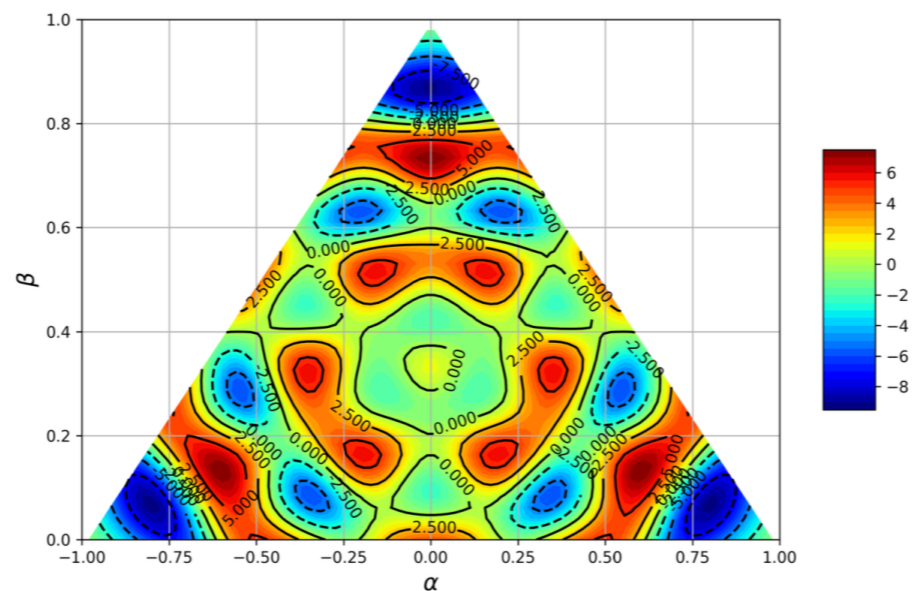
arxiv:1801.02600; arXiv:1903.07487

Motivation

- Many many models of inflation
- New effects in (higher order) correlation functions could potentially allow us to detect new fields
- Models can be complicated, for example with curved field space metric
- In many systems the large N limit has interesting properties. To probe this limit for inflation, however, numerics are essential
- Without numerics, theory error even for simple models can be greater than observational uncertainty
- At very least we should be able to take any model of inflation and confront with (improving) observations

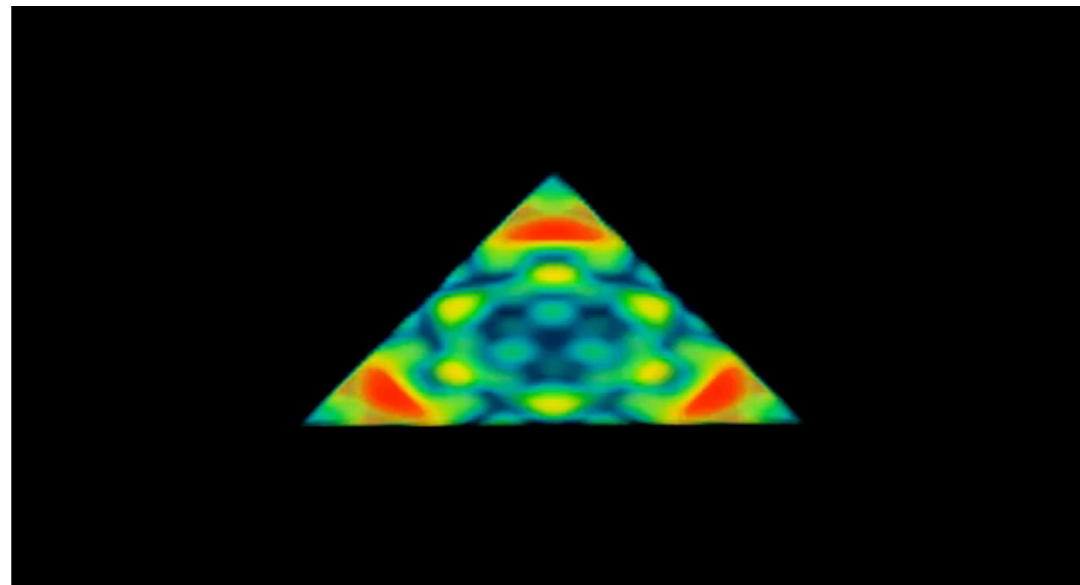
PyTransport

- PyTransport and sibling code CppTransport (developed by David Seery) solves transport equations for inflationary perturbations to produce full power spectrum and bispectrum
- Deals with models with arbitrary numbers of scalar fields, a curved field space metric, perturbative reheating (unreleased)
- Includes all tree-level effects on sub and super-horizon scales
- Publicly available and automated in sense user need only provide potential (and field space metric) — users welcome!



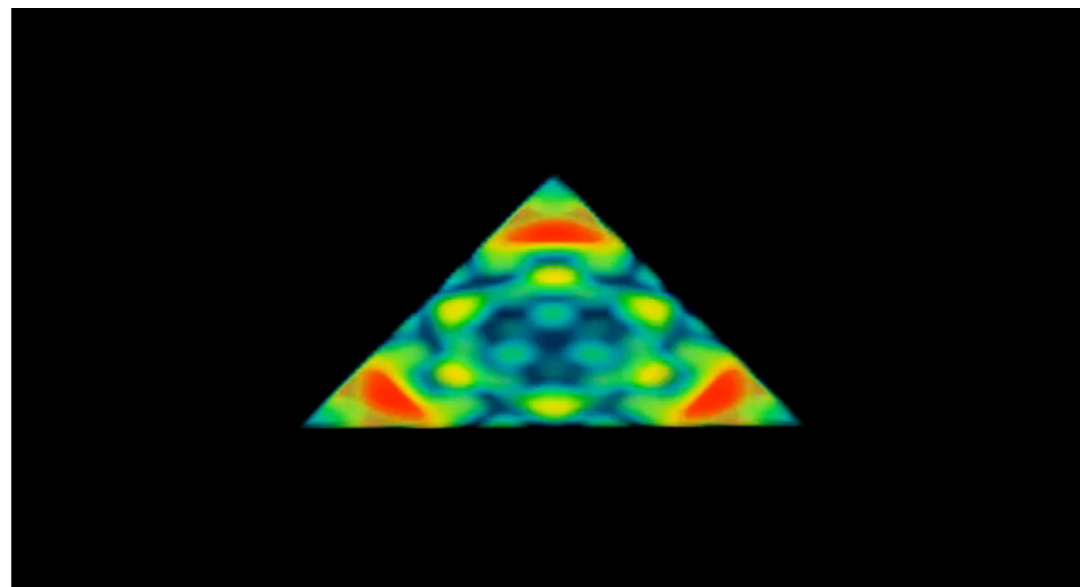
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Observational quantities

- Statistical quantities we want to evaluate

$$\langle \zeta(\mathbf{k}_1)\zeta(\mathbf{k}_2) \rangle = (2\pi)^3 \delta(\mathbf{k}_1 + \mathbf{k}_2) P(k)$$

$$\langle \zeta(\mathbf{k}_1)\zeta(\mathbf{k}_2)\zeta(\mathbf{k}_3) \rangle = (2\pi)^3 \delta(\mathbf{k}_1 + \mathbf{k}_2 + \mathbf{k}_3) B(k_1, k_2, k_3)$$

$$f_{\text{NL}} = \frac{5}{6} \frac{B(k_1, k_2, k_3)}{P(k_1)P(k_2) + P(k_1)P(k_3) + P(k_2)P(k_3)}$$

- Basic predictions

$$P(k) \sim Ak^{-3}$$

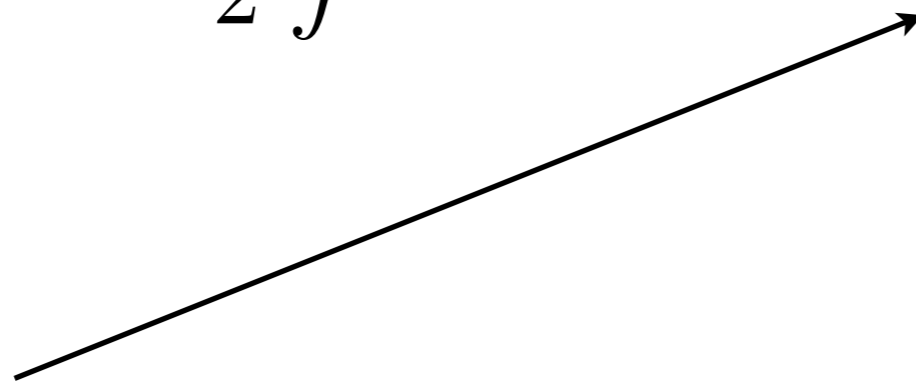
$$f_{\text{NL}} \sim \text{slow roll (for canonical single field)}$$

Calculating statistics

$$S = \frac{1}{2} \int d^4x \sqrt{-g} [M_p^2 R + \mathcal{L}_m]$$

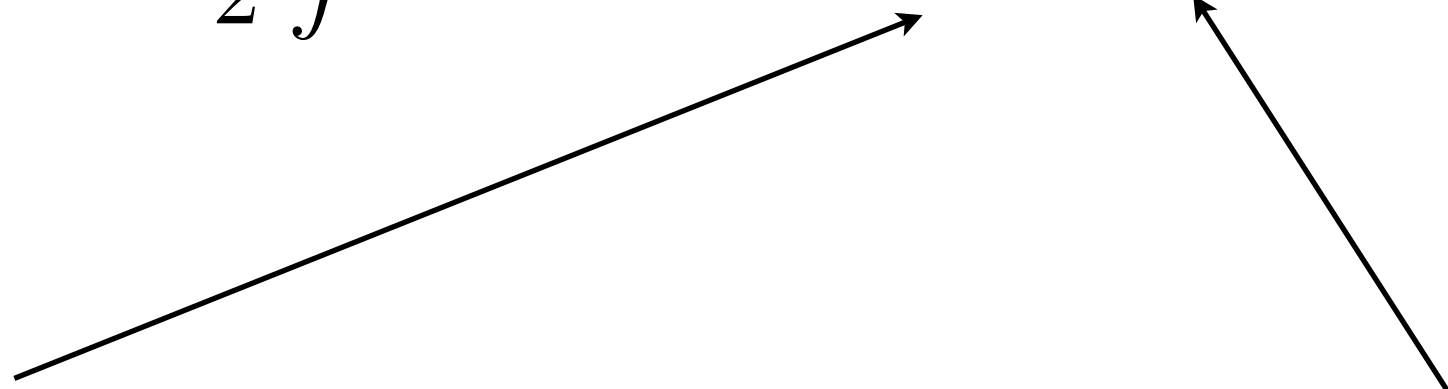
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$$S = \frac{1}{2} \int d^4x \sqrt{-g} [M_p^2 R + \mathcal{L}_m]$$



$$ds^2 = -(1 + 2\Phi)dt^2 + a^2(\delta_{ij} + h_{ij})dx^i dx^j$$

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$$\mathcal{L}_m = -G_{IJ} g^{\mu\nu} \partial_\mu \phi^I \partial_\nu \phi^J - V$$
$$\phi^I + \delta\phi^I$$

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action expanded order by order in fluctuations Q^I
and gravitational waves (tensor) h_{ij}

Calculating statistics

$$S = S_{(2)} + S_{(3)}$$

The diagram shows the equation $S = S_{(2)} + S_{(3)}$ at the top. Below it, two arrows point downwards from the terms $S_{(2)}$ and $S_{(3)}$ to the expressions $\mathcal{O}(2)$ in Q^I and $\mathcal{O}(3)$ in Q^I respectively.

$$\mathcal{O}(2) \text{ in } Q^I \quad \mathcal{O}(3) \text{ in } Q^I$$

Maldacena 2003; Seery and Lidsey 2006; Chen *et al.* 2007; Elliston *et al.* 2012; many others

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\downarrow \searrow

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Lagrangian or Hamiltonian equations of motion for Q^I

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$$\langle Q^I(k_1) Q^J(k_2) \rangle = (2\pi)^3 \delta(\mathbf{k}_1 + \mathbf{k}_2) \Sigma^{IJ}(k_1)$$

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$$Q^I \rightarrow \zeta$$

Calculating the statistics — transport method

- Our approach (schematically)

$$\frac{dQ^I}{dt} = u^I_J Q^J + \frac{1}{2} u^I_{JK} Q^J Q^K$$

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Background and k dependent quantities

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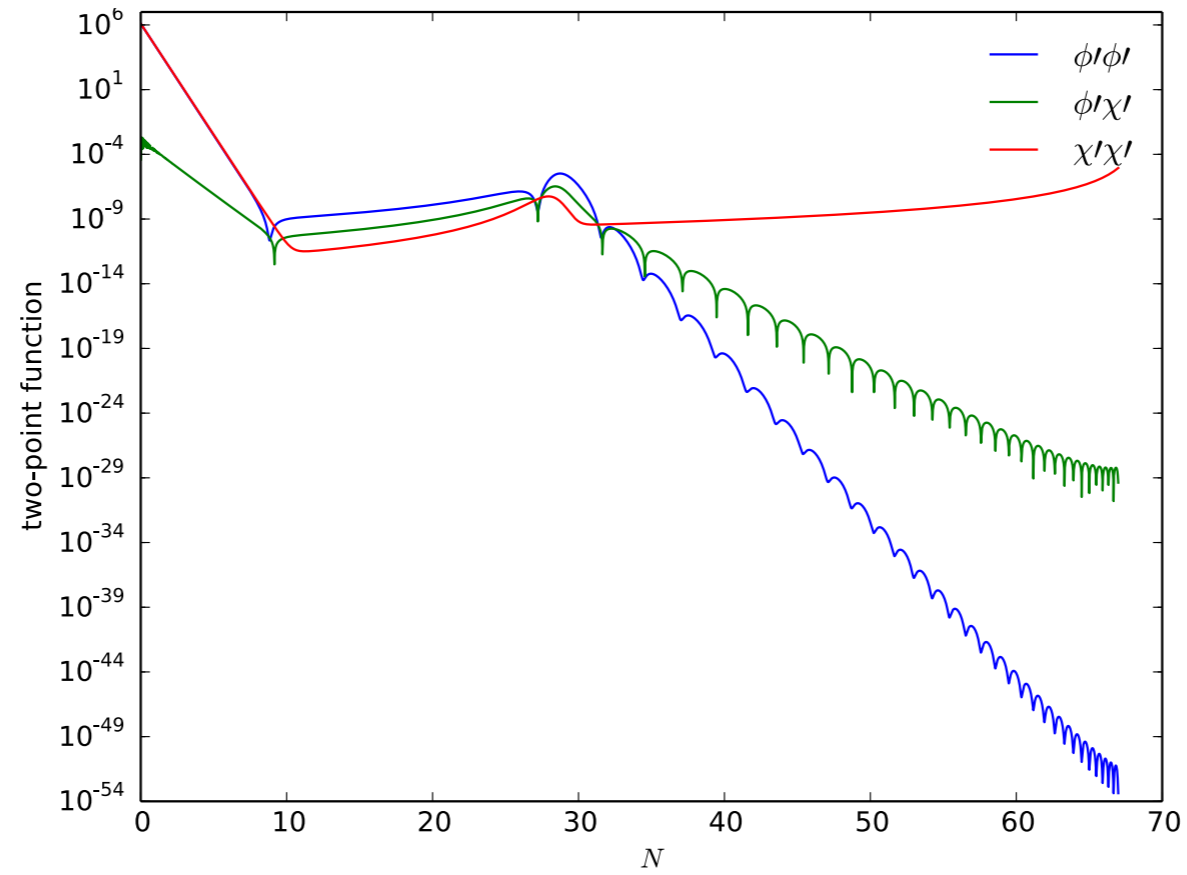


$$\frac{d}{dt} \Sigma^{IJ} = u^I_K \Sigma^{KJ} + u^J_K \Sigma^{IK}$$

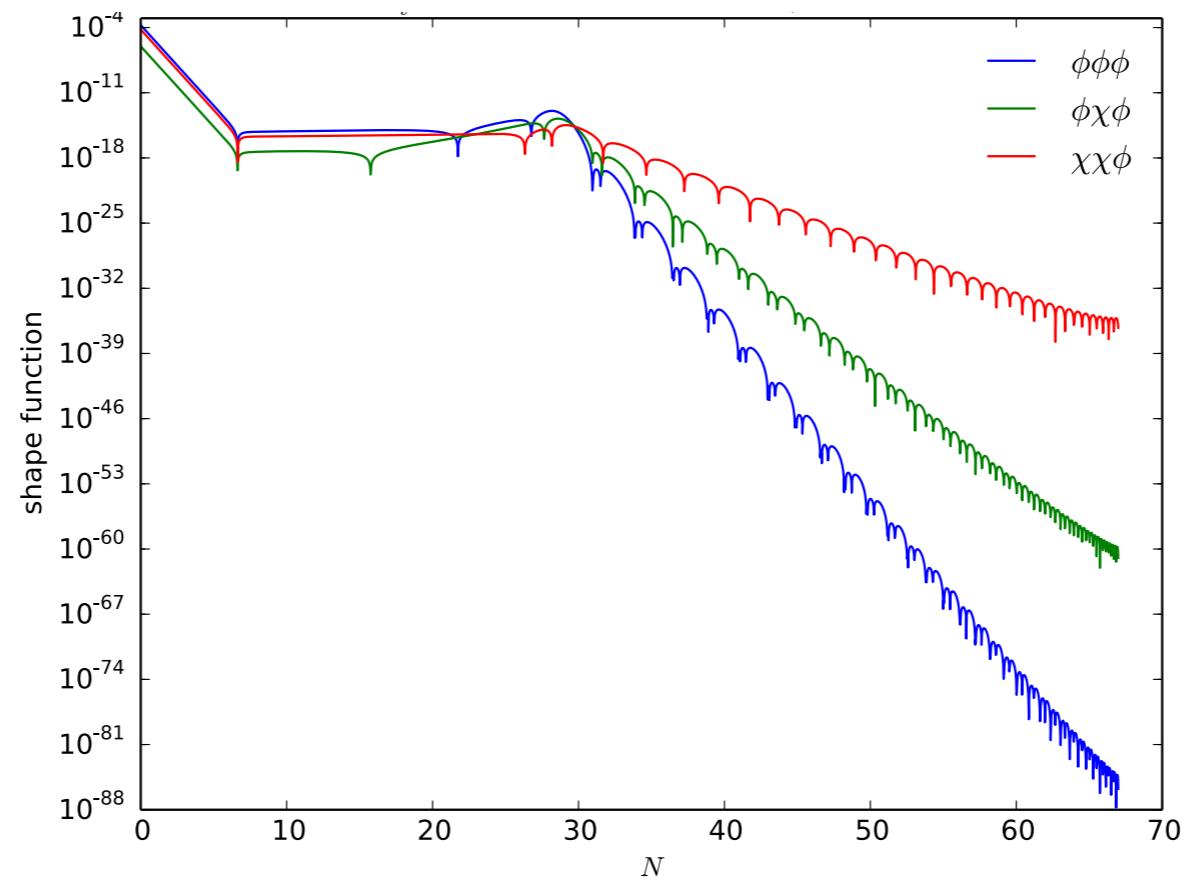
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Ideal for a numerical implementation — solve from Bunch Davis vacuum

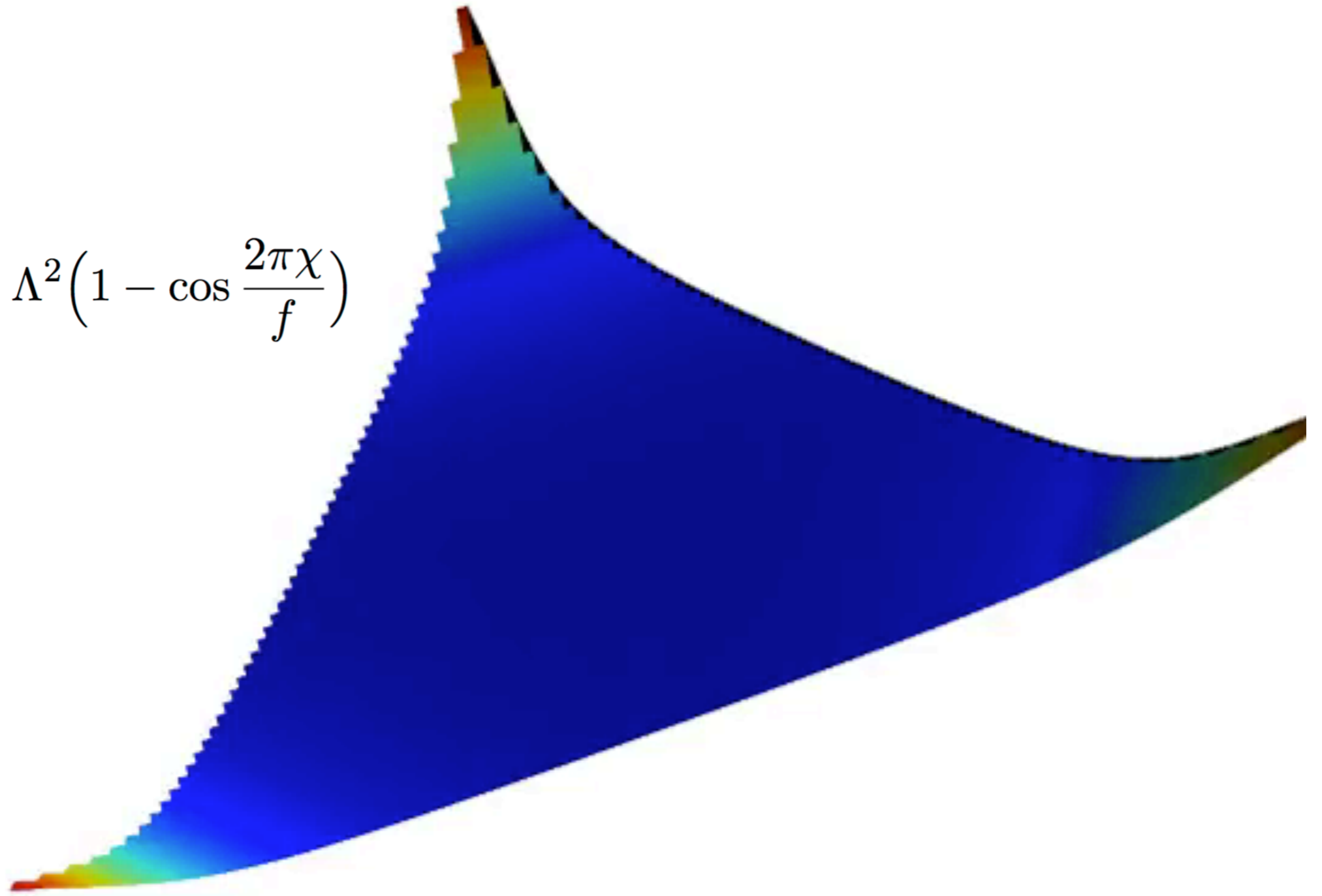
evolution of Σ



evolution of B

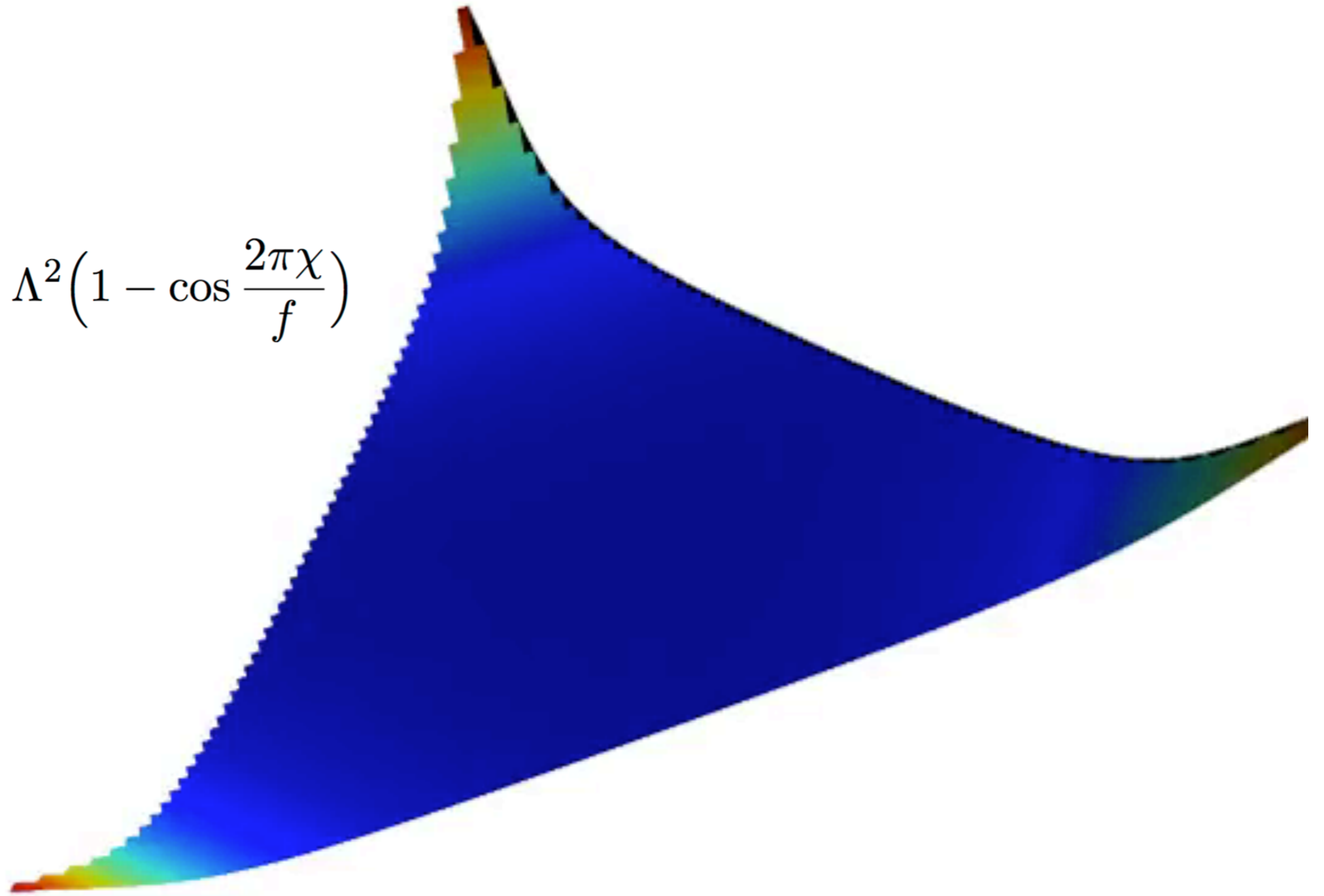


$$V = \frac{1}{4}g\phi^4 + \Lambda^2\left(1 - \cos\frac{2\pi\chi}{f}\right)$$



Slice through reduced bispectrum with $k_1 + k_2 + k_3$ fixed

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Slice through reduced bispectrum with $k_1 + k_2 + k_3$ fixed

Demonstration interlude

$$V = \frac{1}{2}m_{\phi}^2\phi^2 + \frac{1}{2}m_{\chi}^2\chi^2$$

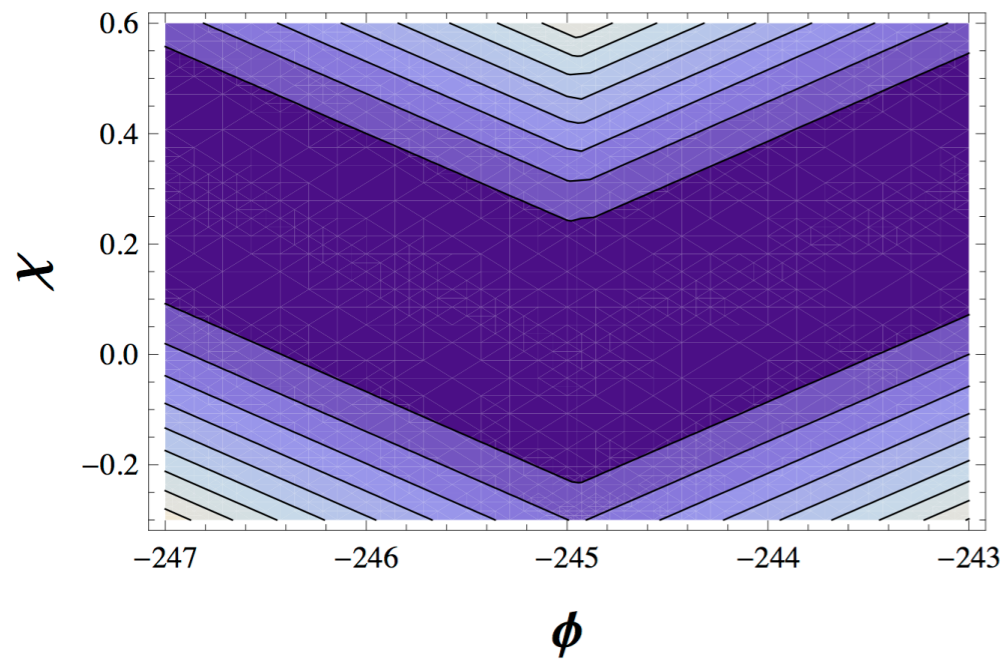
Models

- **Model driven** - string theory, supergravity, MSSM, Standard Model.
At a minimum we should be able to test all models
 - Either concrete models, or random potentials e.g. Dias, Frazer and Marsh (2017), Bjorkmo and Marsh (2017)
- **Phenomenological** - how do multi-field dynamics differ from single field dynamics? - the great hope is that we could detect new fields!
- **New effects** - extra light/heavy fields, curved field space metric -> curved trajectories, isocurvature modes -> Non-Gaussianity Byrnes et al. 2008; Hall and Choi Chen & Wang 2009; Tolley and M. Wyman 2010; Achúcarro et al. 2011 PBH production e.g. Germani, Prokopec (2017,1018), Tomberg, Räsänen (2018) , Byrnes, Cole, Patil (2018)
- **Probabilistic** for many fields a probabilistic interpretation may be needed for many fields e.g. Frazer 2014

Heavy field with turn (c.f. non-geodesic motion)

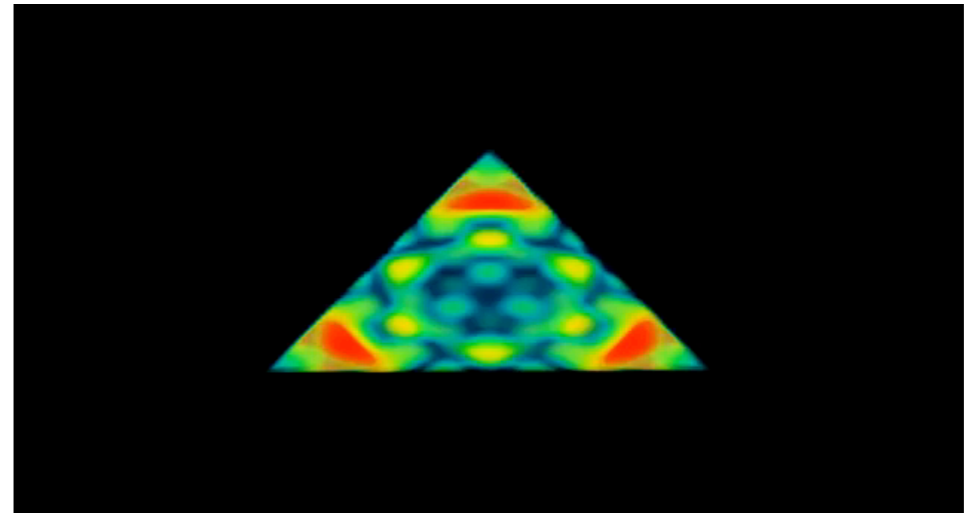
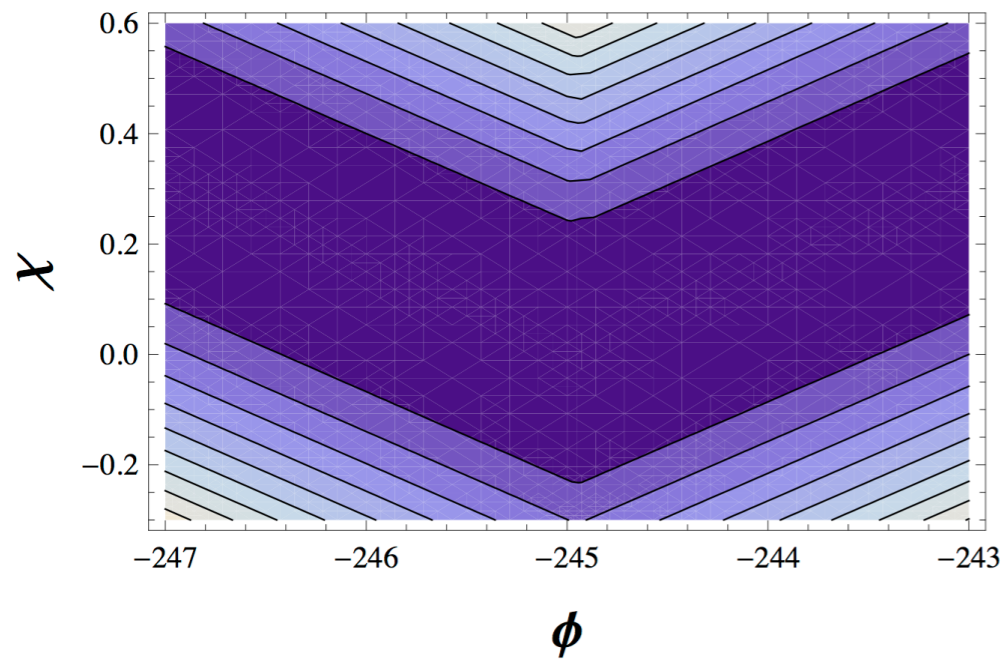
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Goa, Langlois and Mizuno (2014)



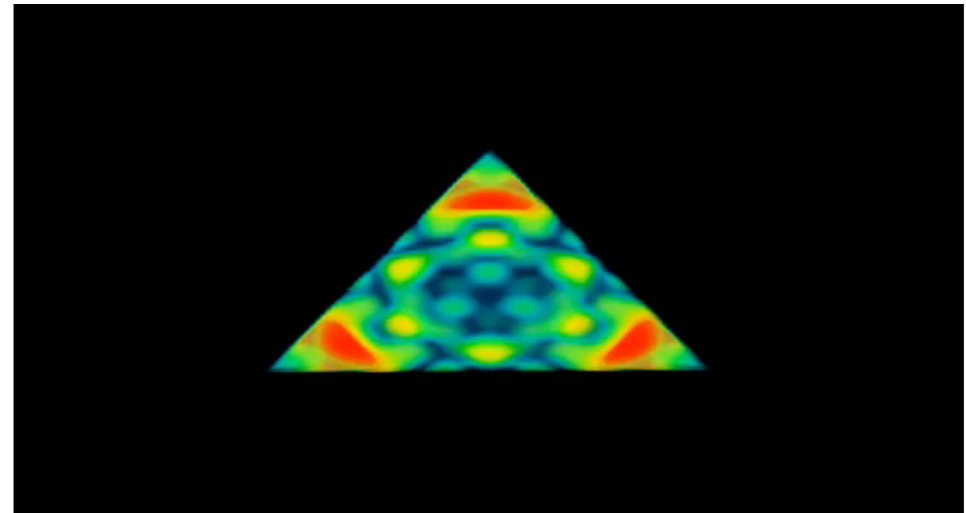
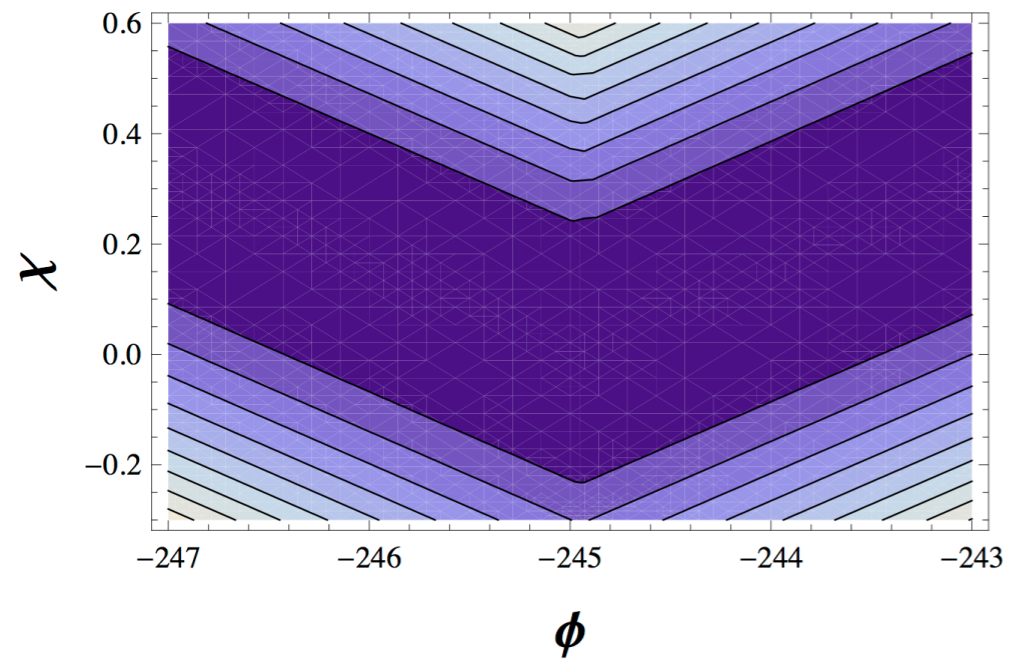
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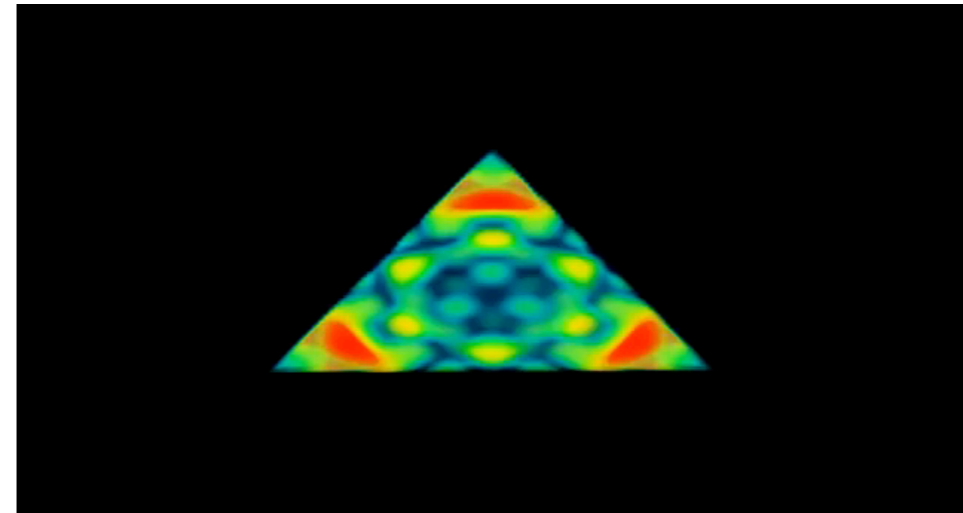
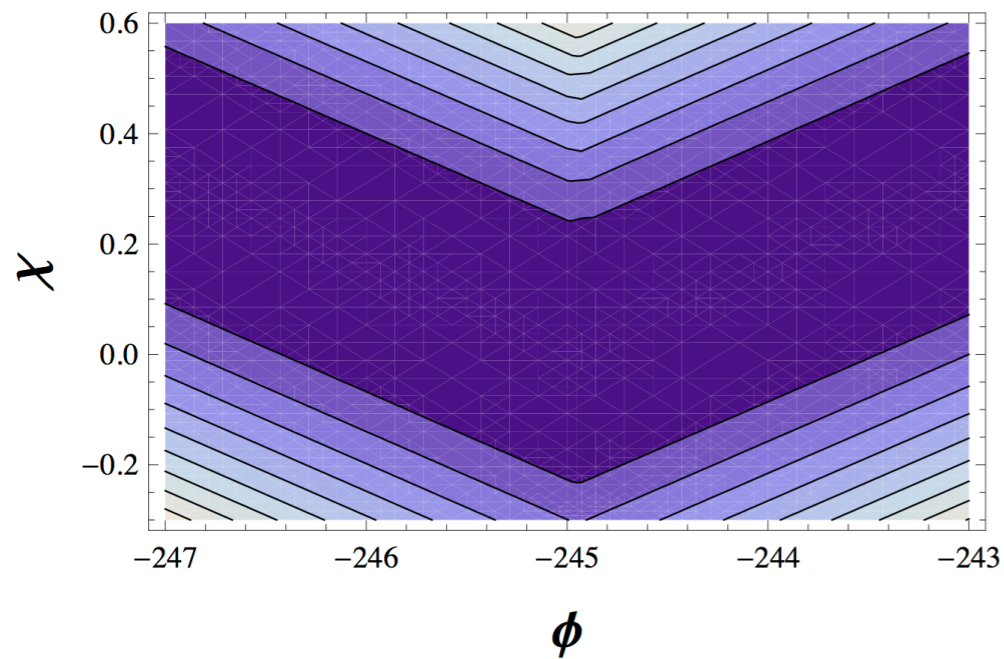
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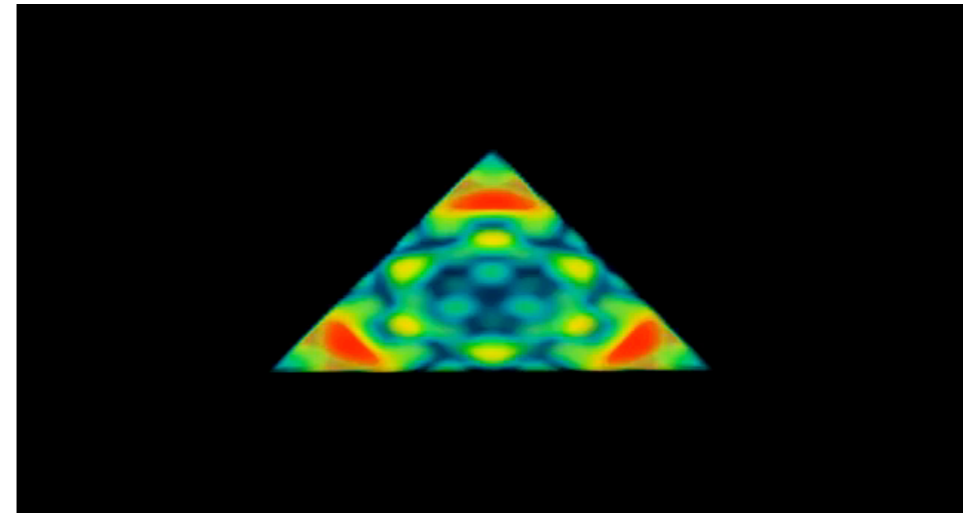
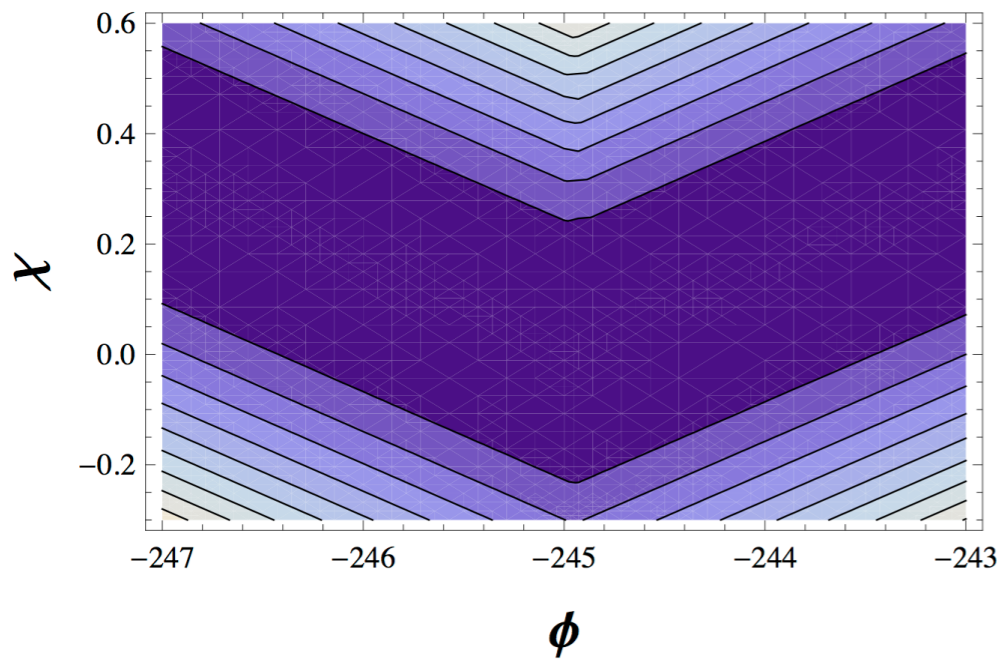
Achucarro, Hardeman, Palma, Patil (2010)

$$\Gamma(\phi_1) = \frac{\Gamma_0}{\cosh^2 \left(2 \left(\frac{\phi_1 - \phi_{1(0)}}{\Delta\phi_1} \right) \right)}$$

$$G_{IJ} = \begin{pmatrix} 1 & \Gamma(\phi_1) & 0 \\ \Gamma(\phi_1) & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

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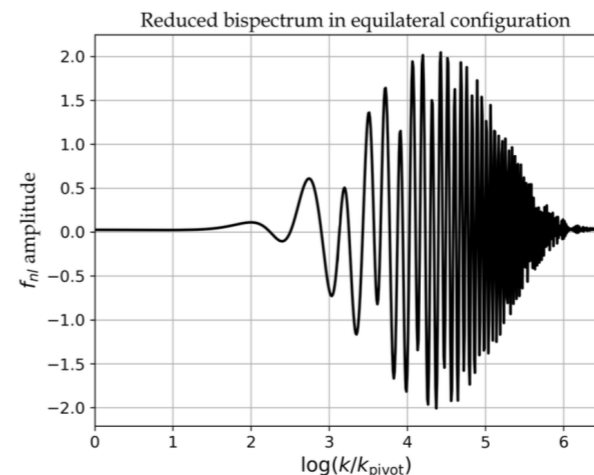
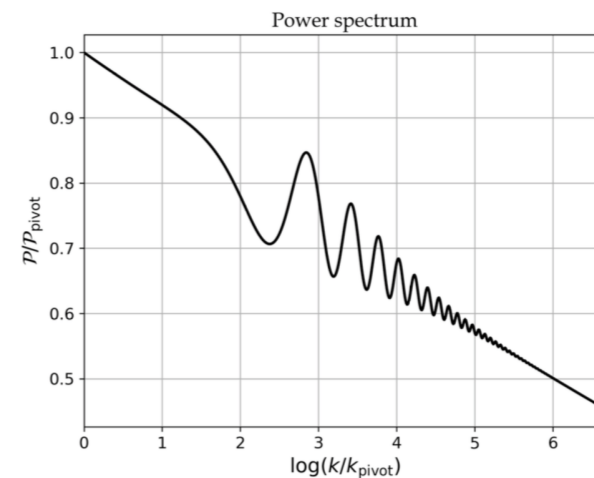
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Non-Minimal coupling to gravity i.e. for multifield alpha attractors

Ronayne, Carrilho, Mulryne and Tenkanen (2018)

$$S_J = \int d^4x \sqrt{-g} \left(\frac{1}{2} \delta_{IJ} g^{\mu\nu} \partial_\mu \phi^I \partial_\nu \phi^J - \frac{M_{\text{P}}^2}{2} (1 + f(\phi^I)) g^{\mu\nu} R_{\mu\nu}(\Gamma) - V(\phi^I) \right)$$

$$f(\phi^I) = \sum_I \xi_I^{(n)} \left(\frac{\phi^I}{M_{\text{P}}} \right)^n$$

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$$S_E = \int d^4x \sqrt{-g} \left(\frac{1}{2} G_{IJ}(\phi^I) \partial_\mu \phi^I \partial^\mu \phi^J - \frac{1}{2} M_{\text{P}}^2 R - V(\phi^I) \Omega^{-2}(\phi^I) \right)$$

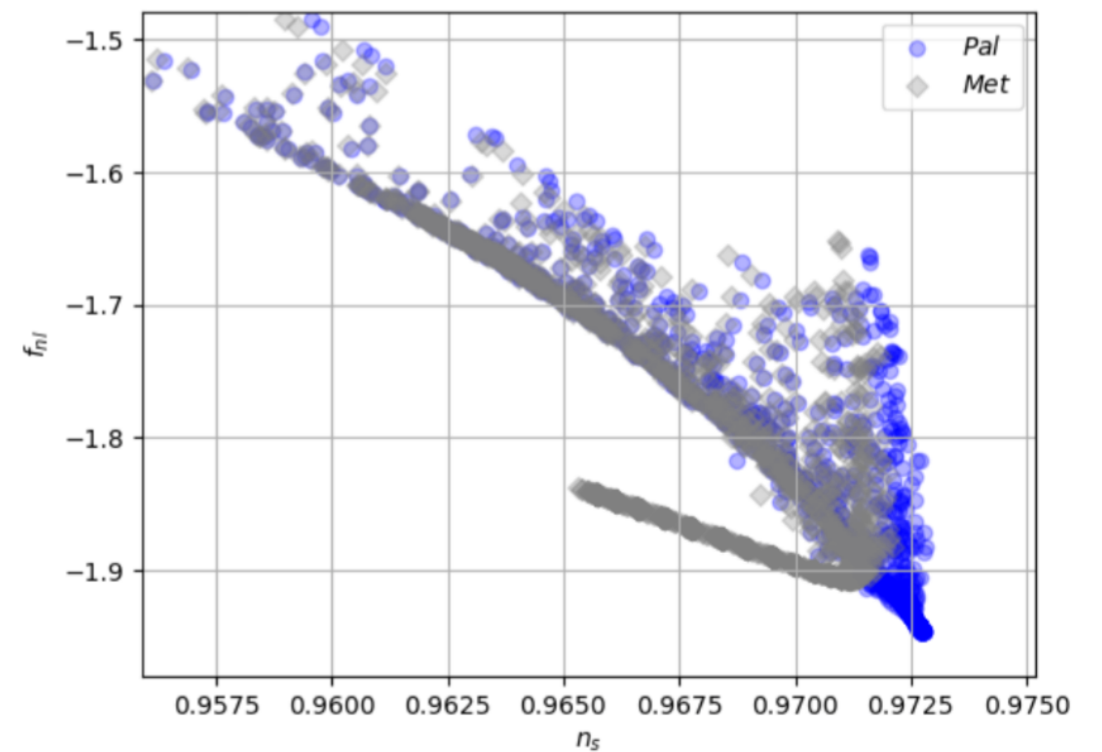
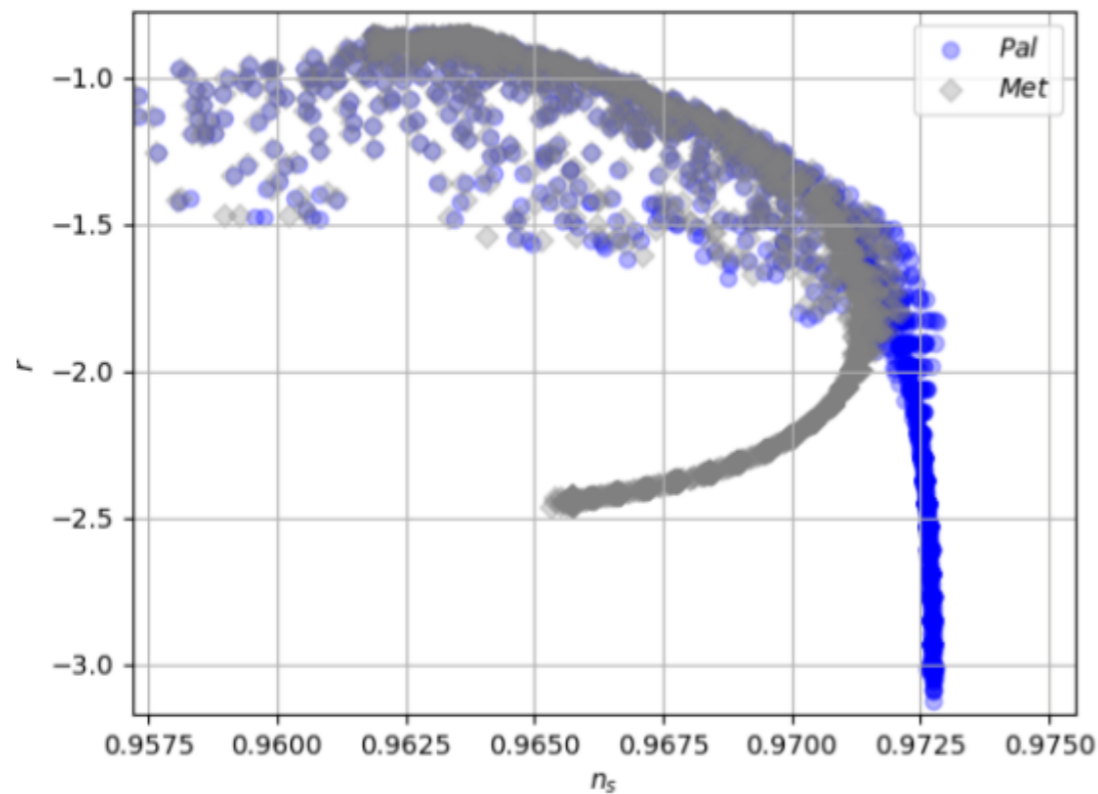
$$G_{IJ} = \Omega^{-1} \delta_{IJ} + \frac{3}{2} \nu M_{\text{P}}^2 \Omega^{-2} \frac{\partial \Omega}{\partial \phi^I} \frac{\partial \Omega}{\partial \phi^J}$$

0 for metric, 1 for Palatini

Non-Minimal coupling to gravity i.e. for multifield alpha attractors

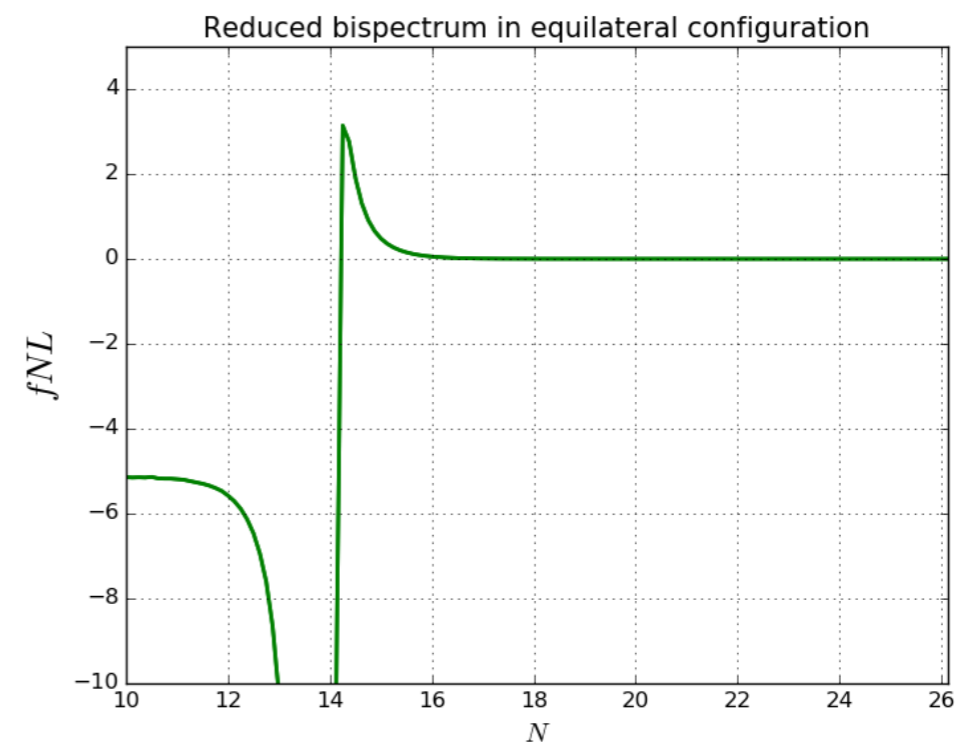
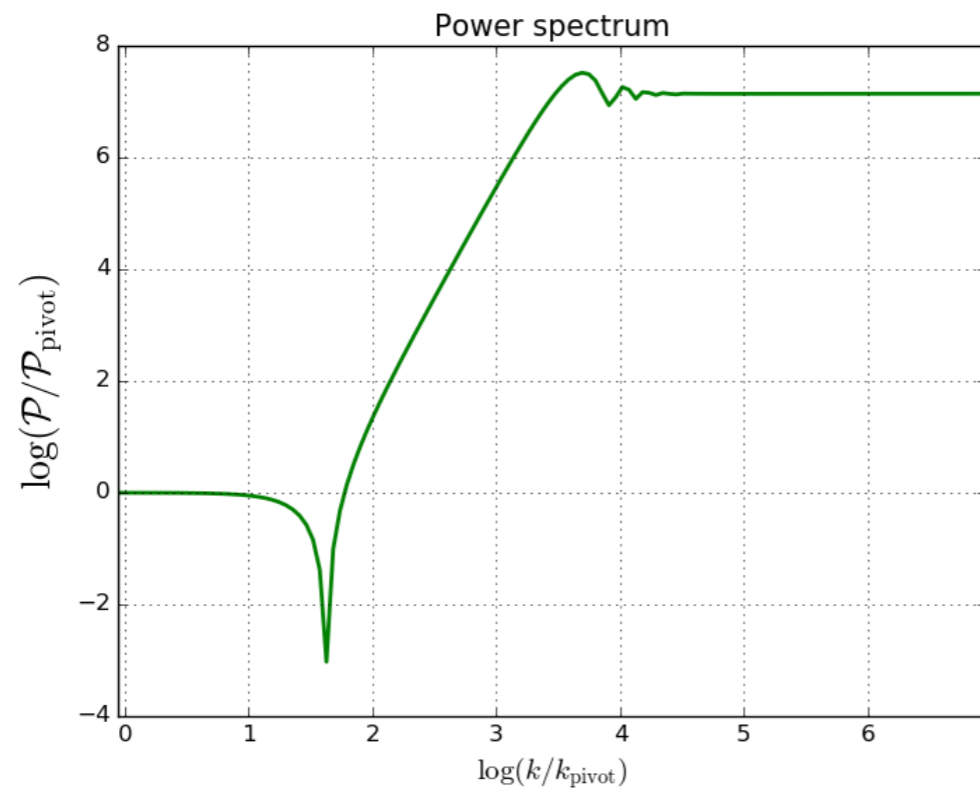
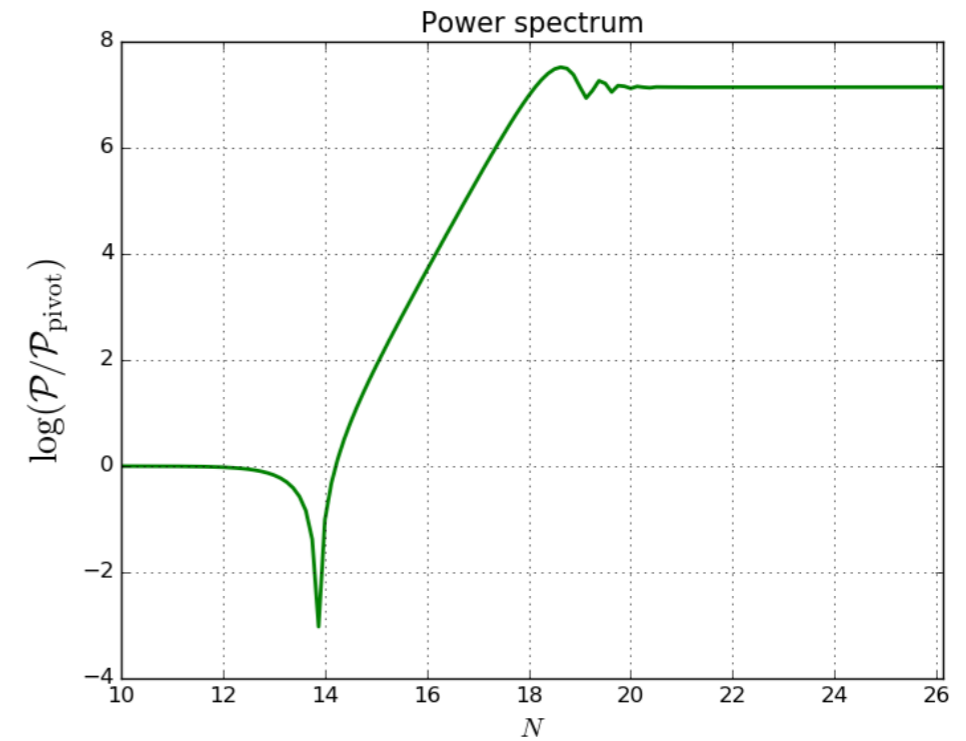
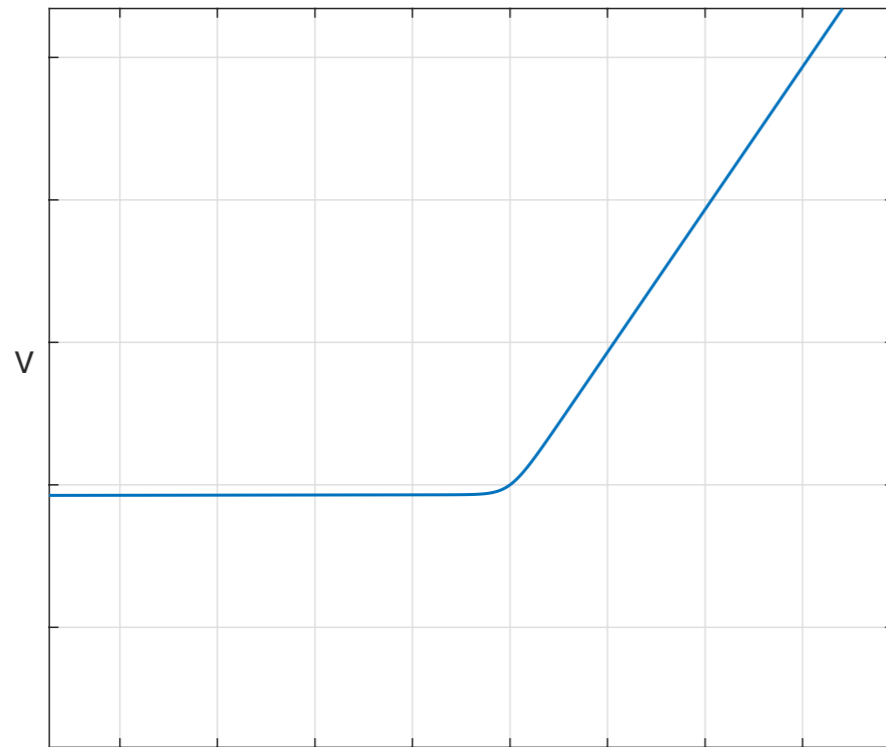
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Primordial black holes

e.g. Germani, Prokopec (2017,1018), Tomberg Räsänen (2018) , Byrnes, Cole, Patil (2018)



Reheating

PyTransport with perturbative reheating (in progress)

- Often isocurvature modes left at end of inflation and so zeta evolves
- Phenomenological way forward is to introduce decay to other radiation and other fluids, gives (with associated perturbed equations to second order)

$$D_t \dot{\phi}^I + 3H \dot{\phi}^I - \Gamma_a^{IJ} \dot{\phi}_J + G^{IJ} V_{,J} = 0$$

$$\dot{\rho}_a + 3H \gamma_a \rho_a + \Gamma_a^{IJ} \dot{\phi}_I \dot{\phi}_J = 0$$

PyTransport with perturbative reheating (in progress)

- e.g. N-axion (c.f. Kim, Liddle, Seery (2009), uncoupled case leads to observable non-Gaussianity)

$$\mathcal{L} = \frac{1}{2} G_{IJ} \partial\phi^I \partial\phi^J + \sum_K \Lambda_K^4 (1 - \cos(\phi^K))$$

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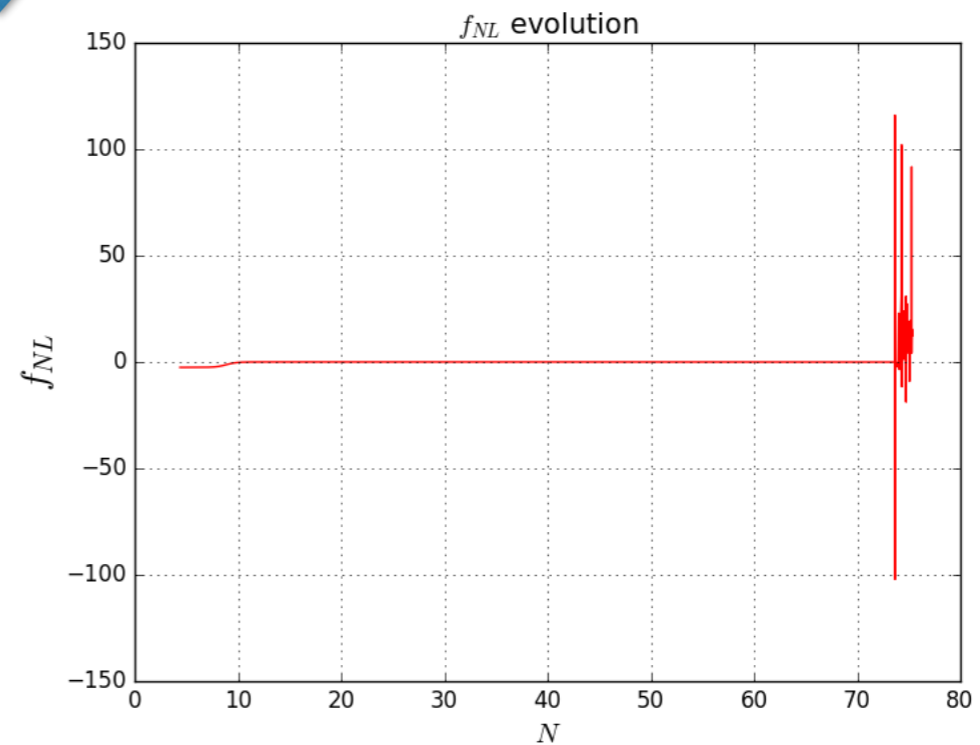
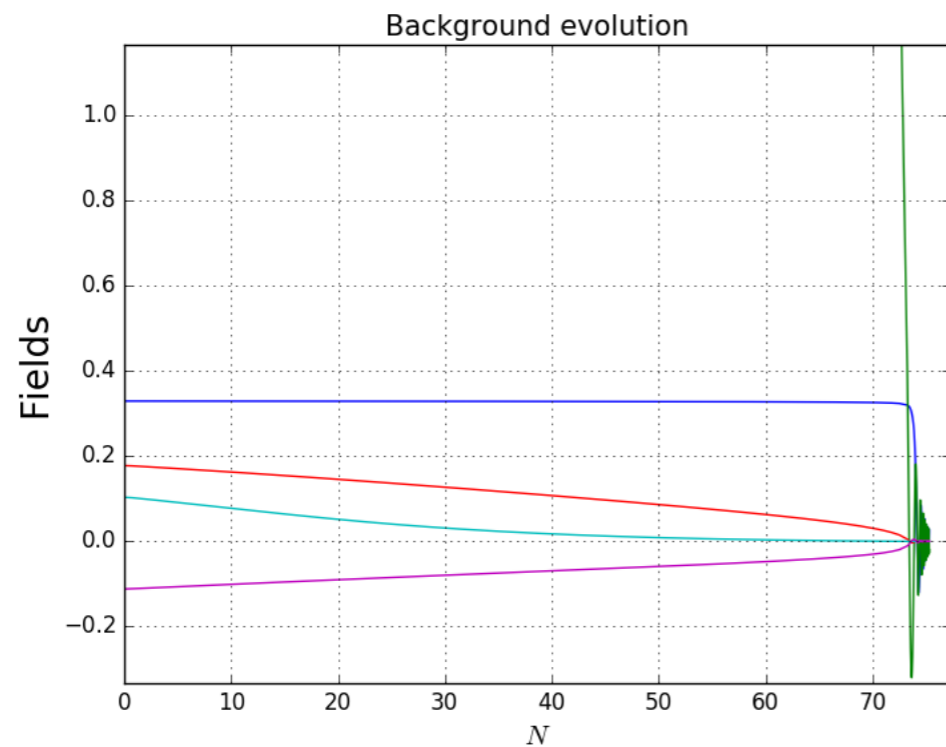
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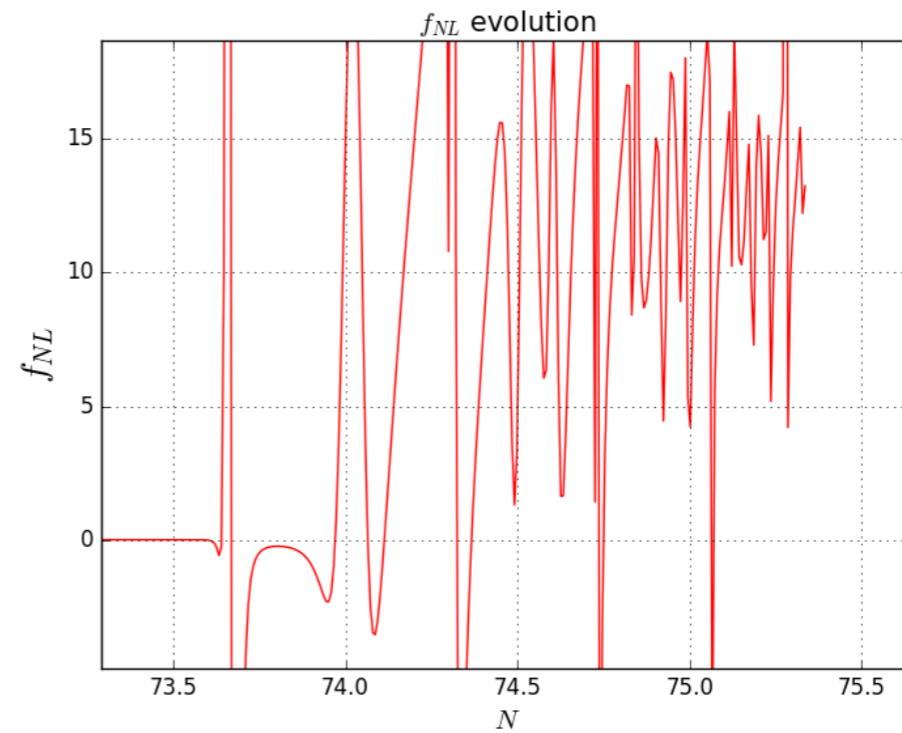
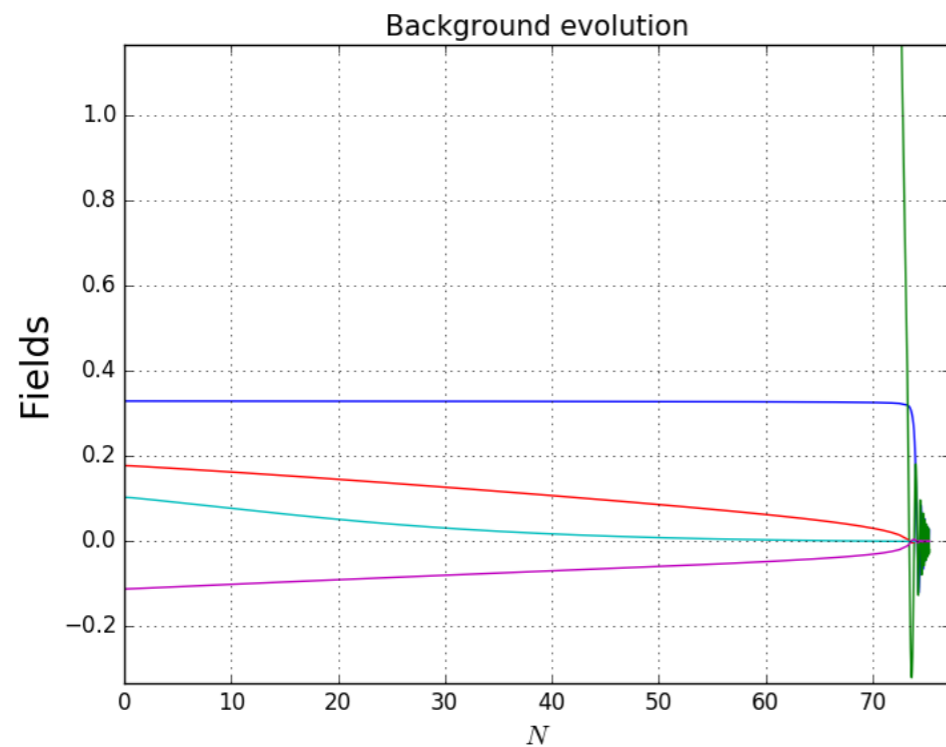
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PyTransport with perturbative reheating (in progress)

- e.g. N-axion (c.f. Kim, Liddle, Seery (2009), uncoupled case leads to observable non-Gaussianity)

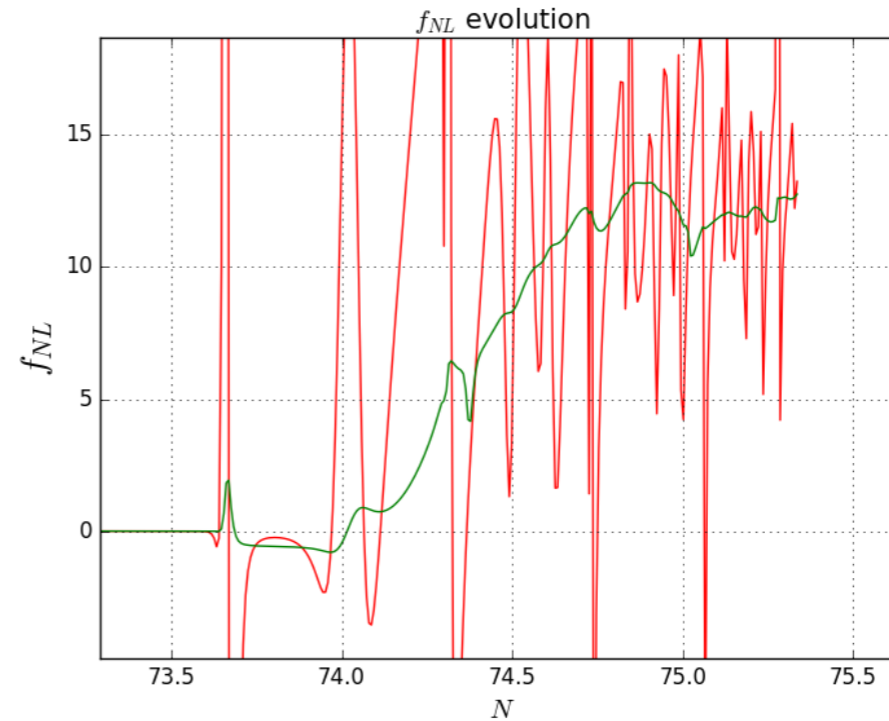
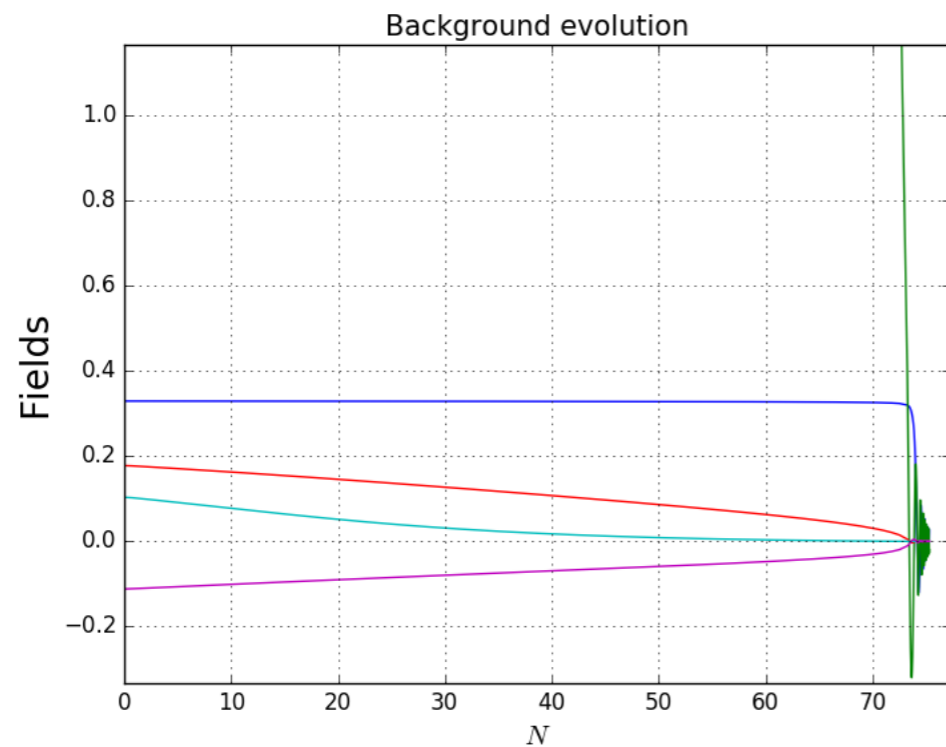
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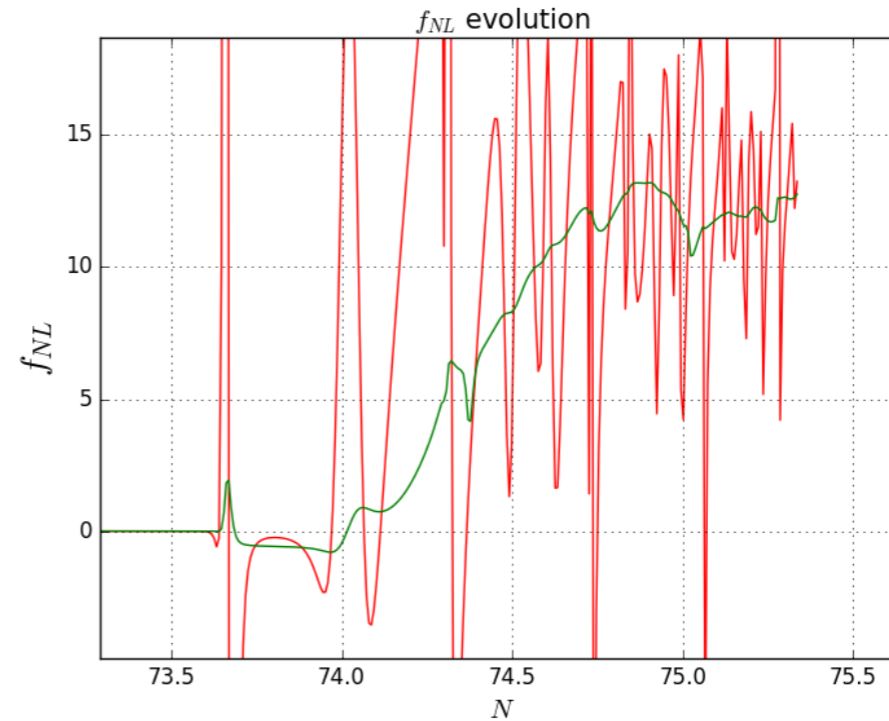
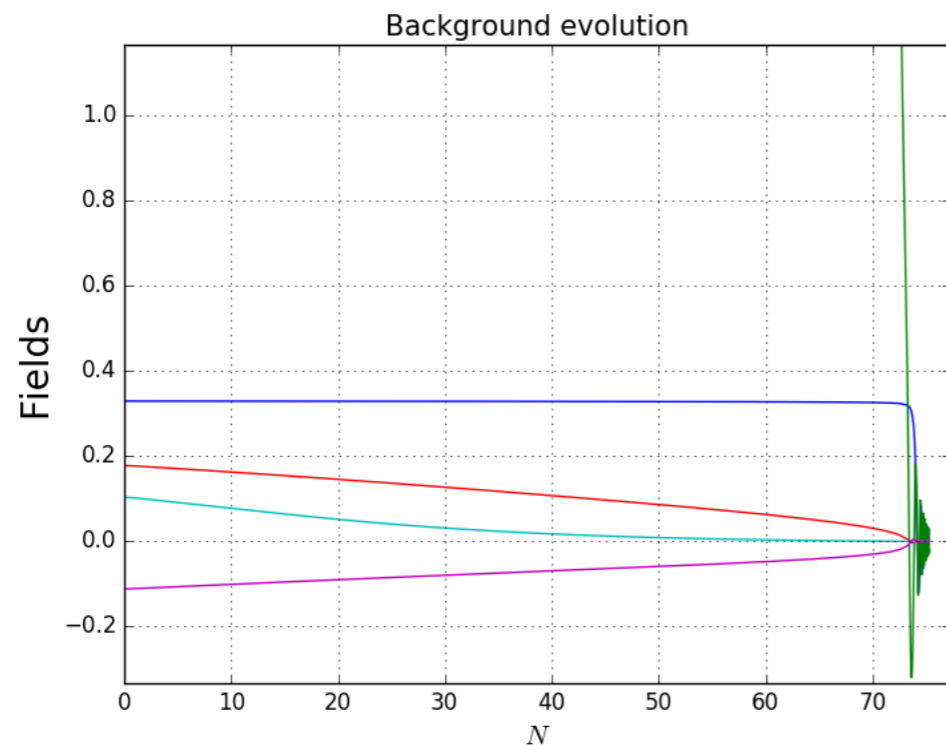
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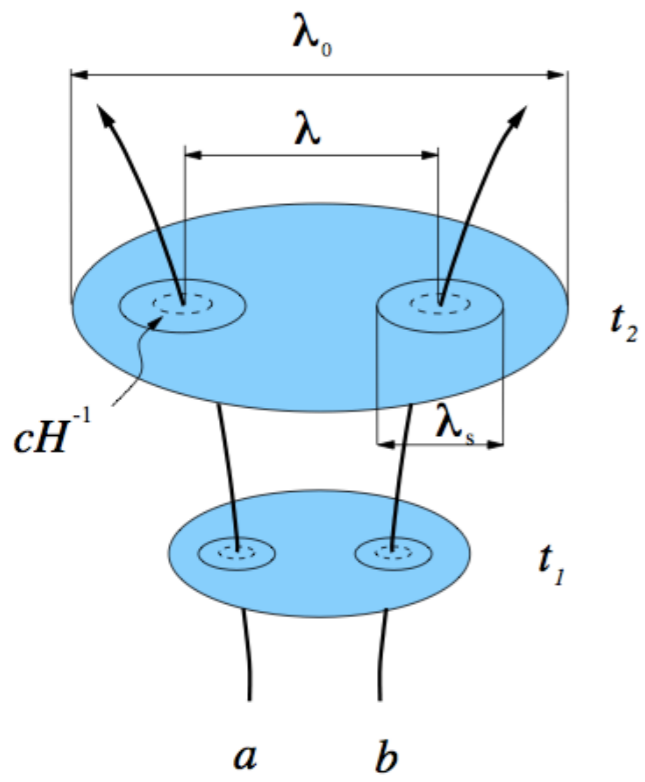


- More generally, Gaussian random landscape around a minimum (c.f. Bjorkmo and Marsh (2017))

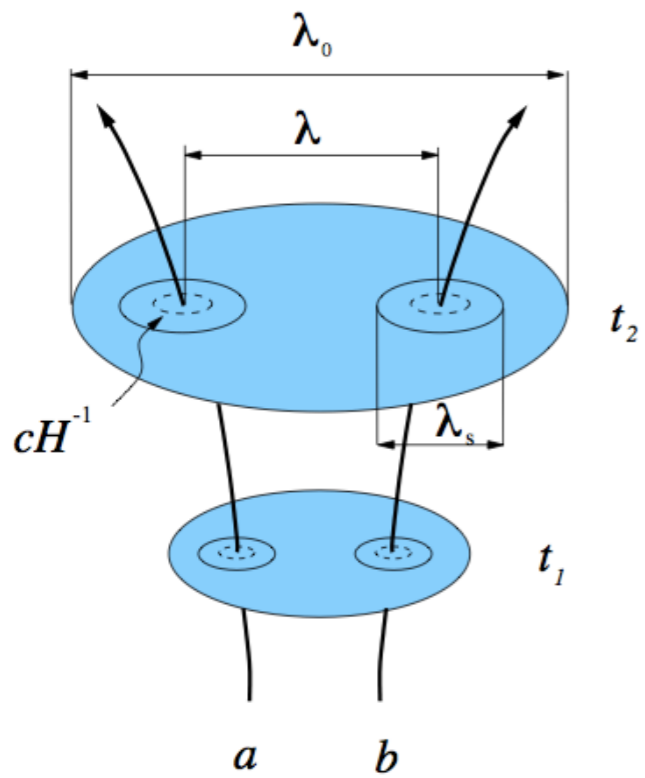
Perturbations through non-perturbative reheating

- What happens if isocurvature present and reheating is non-perturbative (i.e. some form of preheating)
- Dynamics must be tracked using lattice simulations
- Perturbations can be tracked using δN
- However usual expansion can't be used
- Archetypal example is massless preheating

$$V = \frac{1}{4}\lambda\phi^4 + \frac{1}{2}g^2\phi^2\chi^2$$

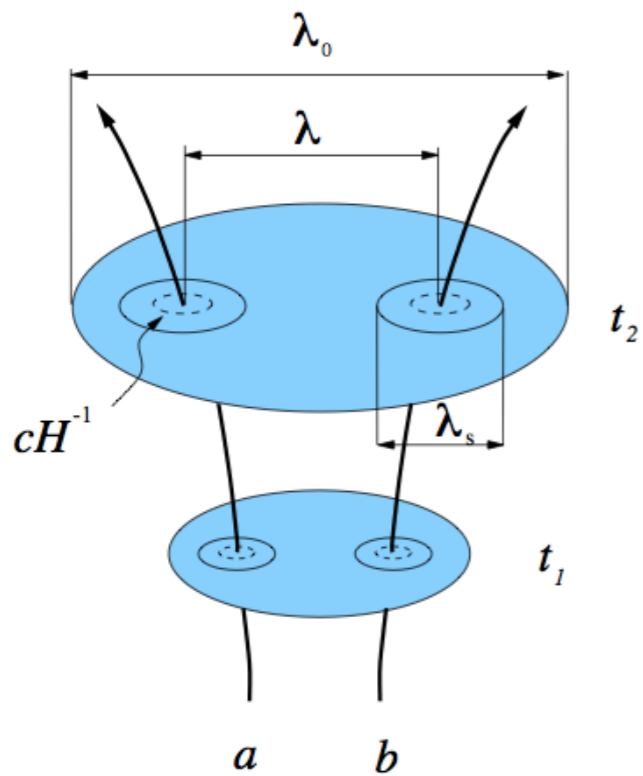


Wands et al., 2000



$$\zeta(\mathbf{x}) = \delta N(\mathbf{x}) = N(\vec{\chi}(\mathbf{x})) - \bar{N}$$

Wands et al., 2000



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$$\delta N(\mathbf{x}) = N_{,I} \delta \chi^I(\mathbf{x}) + \frac{1}{2} N_{,IJ} \left(\delta \chi^I(\mathbf{x}) \delta \chi^J(\mathbf{x}) - \overline{\delta \chi^I \delta \chi^J} \right)$$

Wands et al., 2000

- But....

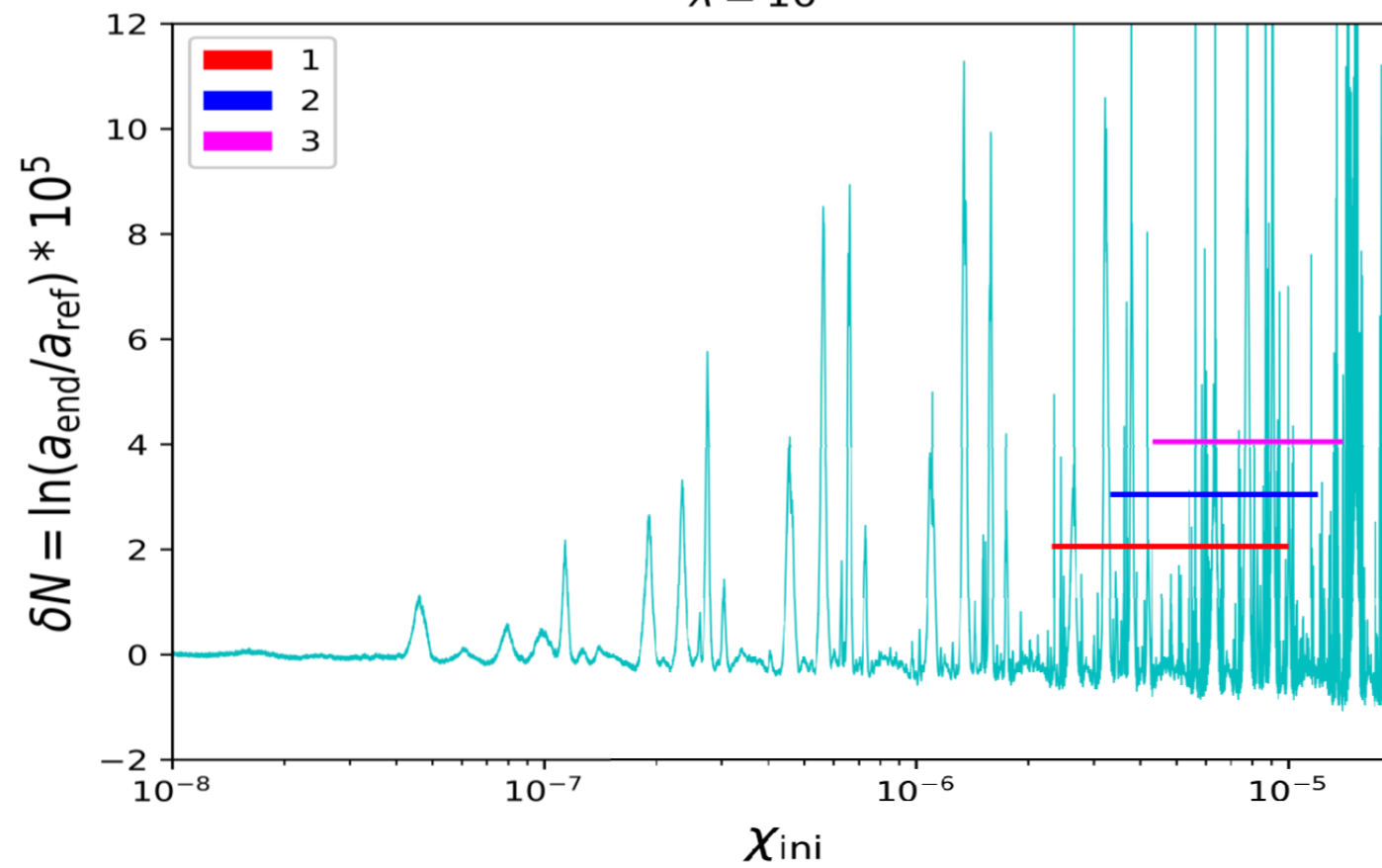
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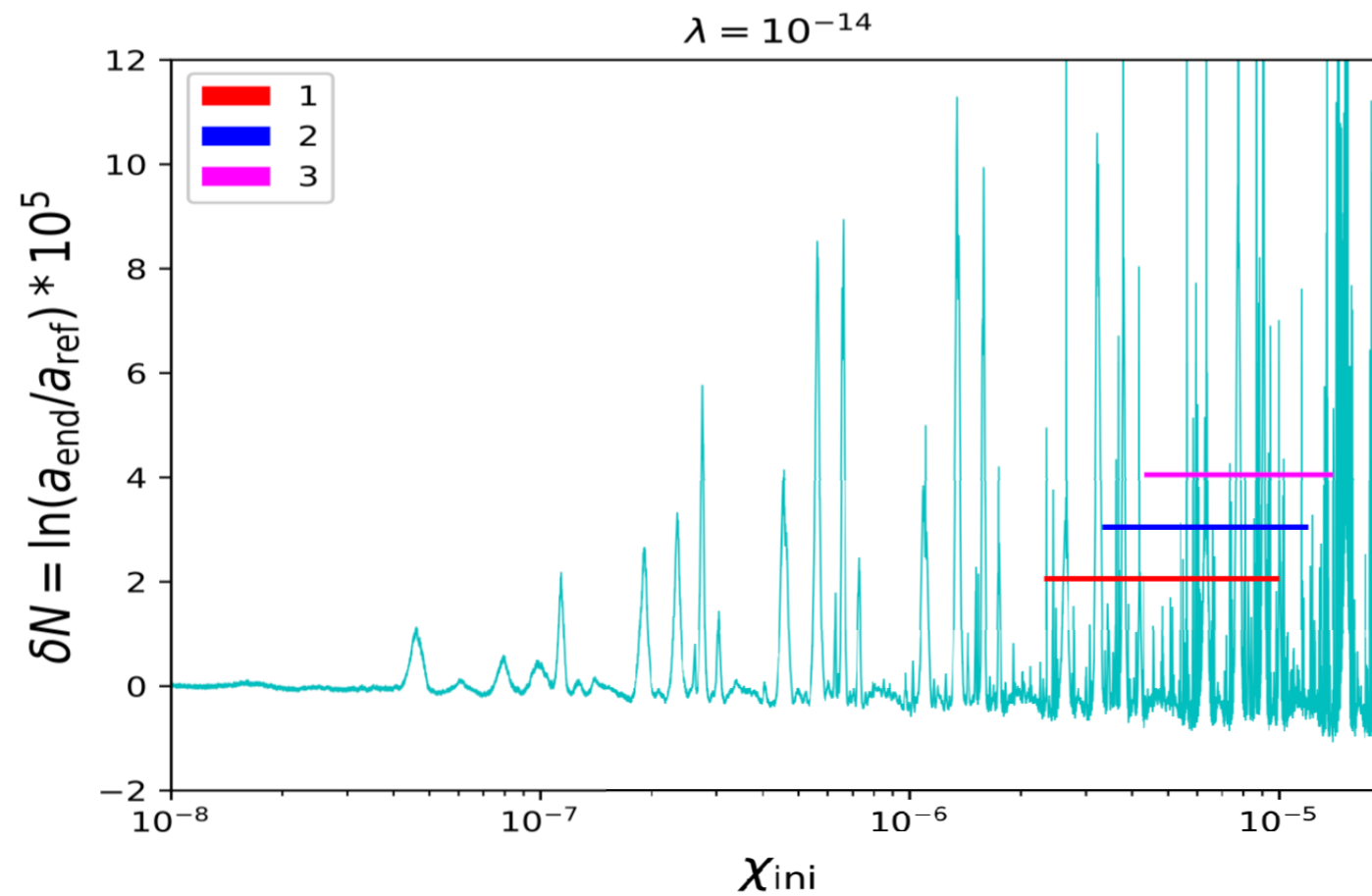


$$\lambda = 10^{-14}$$



- But....

$$V = \frac{1}{4}\lambda\phi^4 + \frac{1}{2}g^2\phi^2\chi^2$$



- A lot of work on this e.g. Chambers, Rajantie (2008), Bond, Frolov, Huang, Kofman (2009); Chambers, Nurmi, Rajantie (2010); Suyama and Yokoyama (2013); Bethke, Figueroa, Rajantie (2013)

- If one wants to know correlates, use full expression:

$$\begin{aligned}\langle \zeta_1 \dots \zeta_m \rangle &= \langle (N_1 - \bar{N}) \dots (N_m - \bar{N}) \rangle \\ &= \int d\vec{\chi}_1 \dots \int d\vec{\chi}_m (N_1 - \bar{N}) \dots (N_m - \bar{N}) \\ &\quad \times \mathcal{P}(\vec{\chi}_1, \dots, \vec{\chi}_m)\end{aligned}$$

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 \end{aligned}$$

- But often don't know distribution, just the moments (from PyTransport for example)
- Try a different expansions see Suyama and S. Yokoyama (2013); Bethke, Figueroa, Rajantie (2013)
- Assume field space perturbations are close to Gaussian, and

$$\langle \delta\phi^I(\mathbf{x}_1) \delta\phi^J(\mathbf{x}_2) \rangle < \langle \delta\phi^I(\mathbf{x}) \delta\phi^J(\mathbf{x}) \rangle$$

- Leads to:

$$P_\zeta(k) \approx \tilde{N}_I \tilde{N}_J \Sigma^{IJ}(k)$$

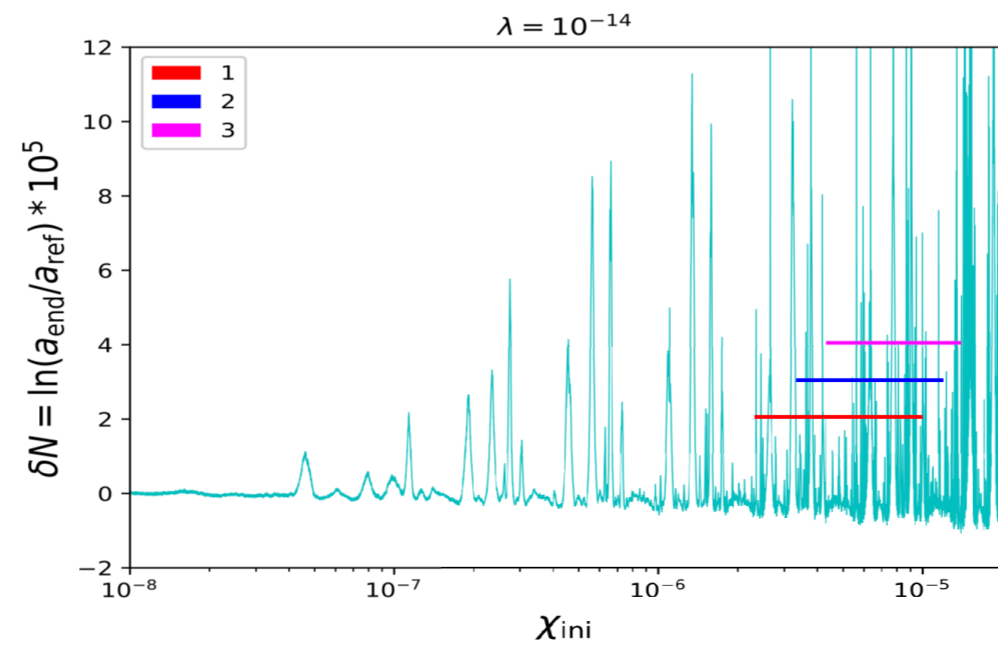
$$B_\zeta(k_1, k_2, k_3) \approx \tilde{N}_I \tilde{N}_J \tilde{N}_K \alpha^{IJK}(k_1, k_2, k_3) \\ + (\tilde{N}_I \tilde{N}_J \tilde{N}_{KL} \Sigma^{IK}(k_1) \Sigma^{JL}(k_2) \\ + \text{cyclic}).$$

- With:

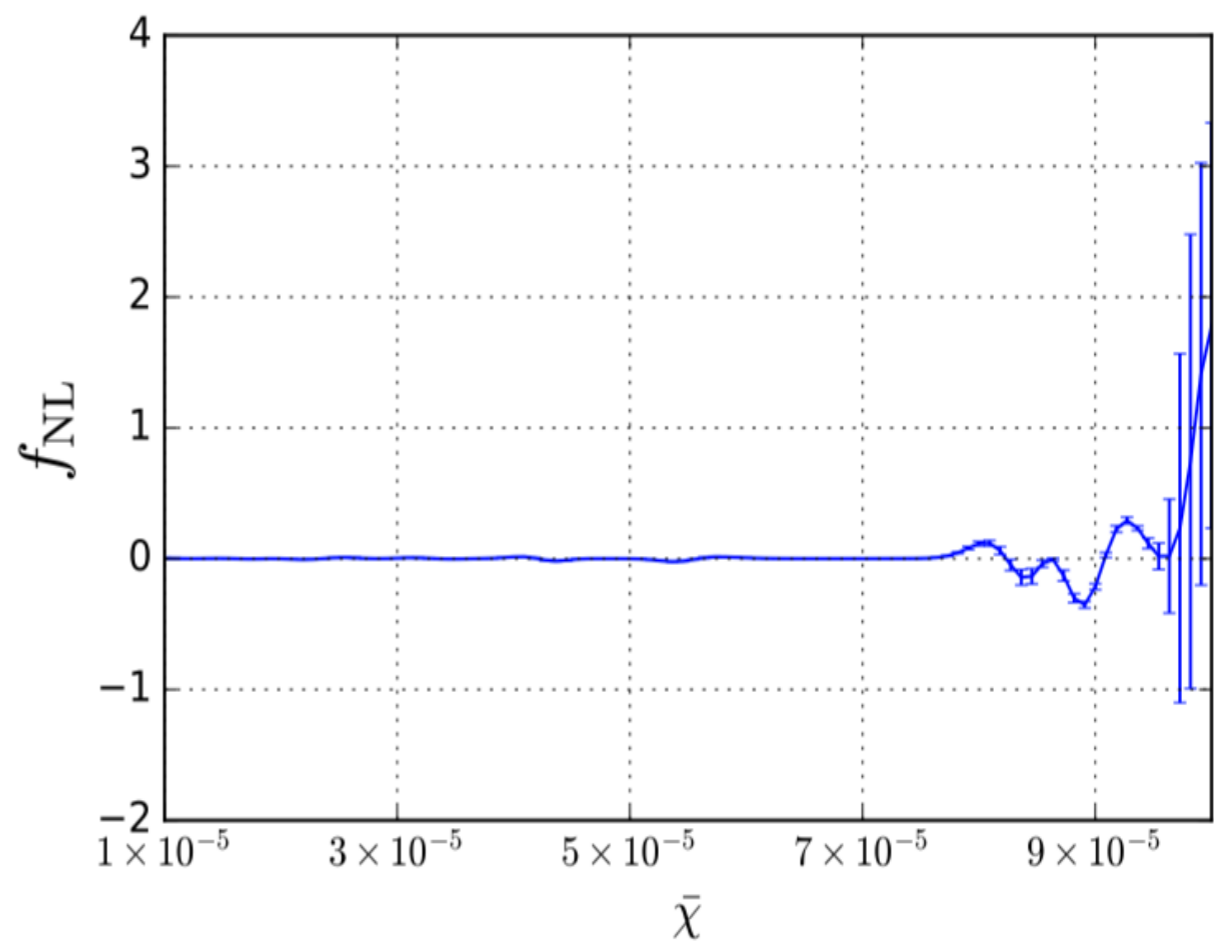
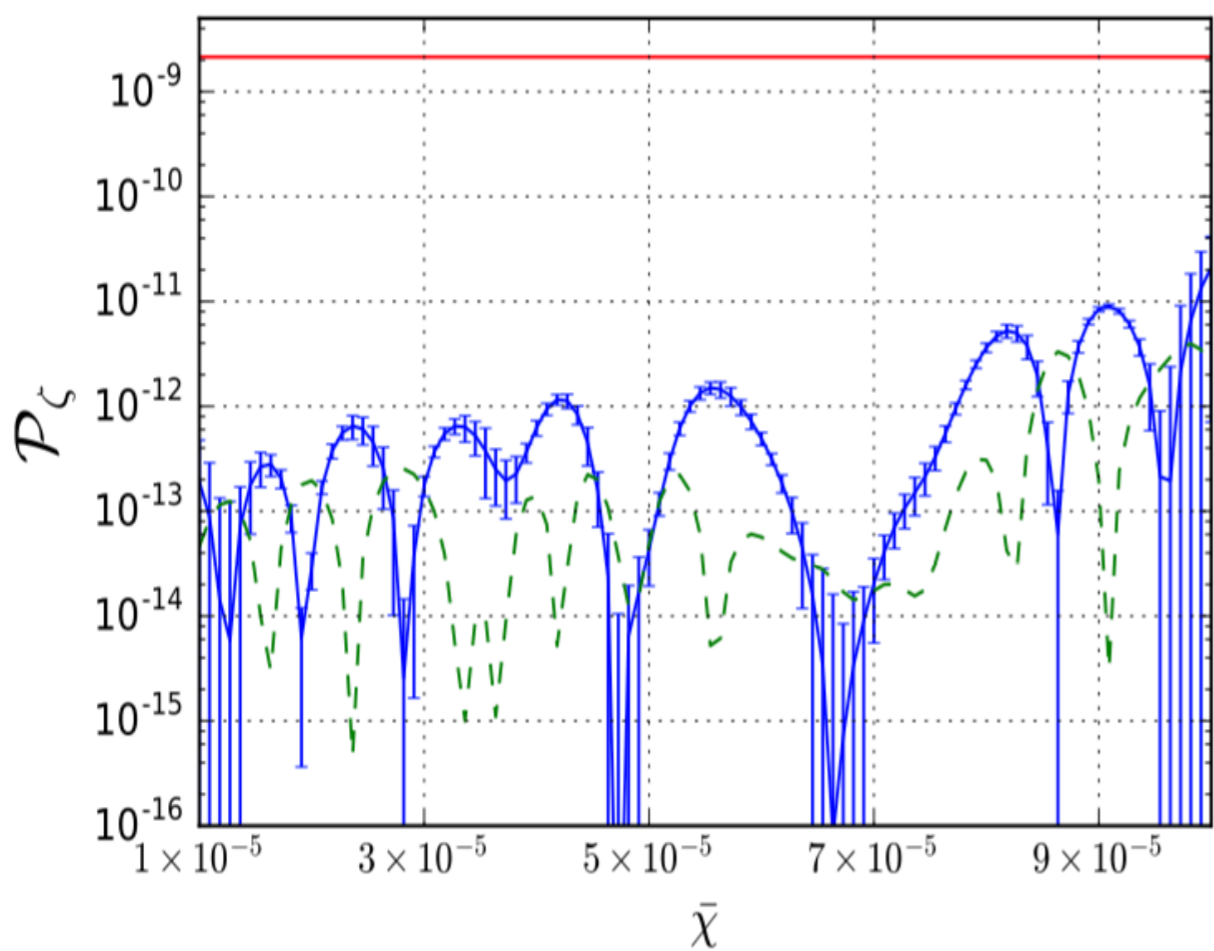
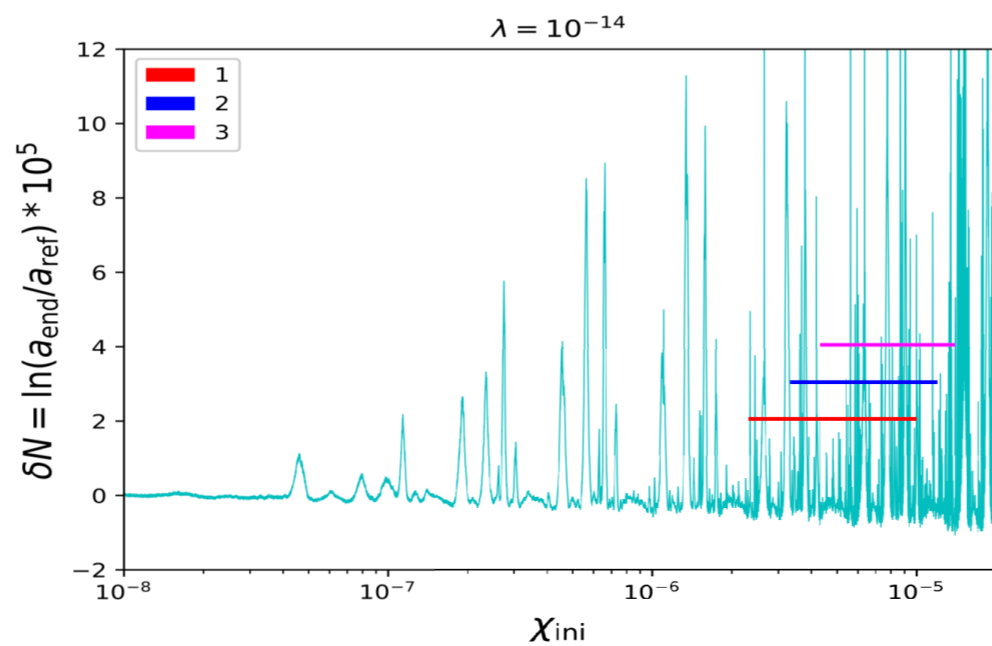
$$\tilde{N}_I = \Sigma_{IJ}^{-1} \int d\vec{\chi}_1 \mathcal{P}_G(\vec{\chi}_1) N_1 \delta\chi_1^J$$

$$\tilde{N}_{IJ} = \Sigma_{IK}^{-1} \Sigma_{JL}^{-1} \int d\vec{\chi}_1 \mathcal{P}_G(\vec{\chi}_1) (N_1 - \bar{N}) \delta\chi_1^K \delta\chi_1^L$$

- Allows....



- Allows....





Calculating the statistics — usual method

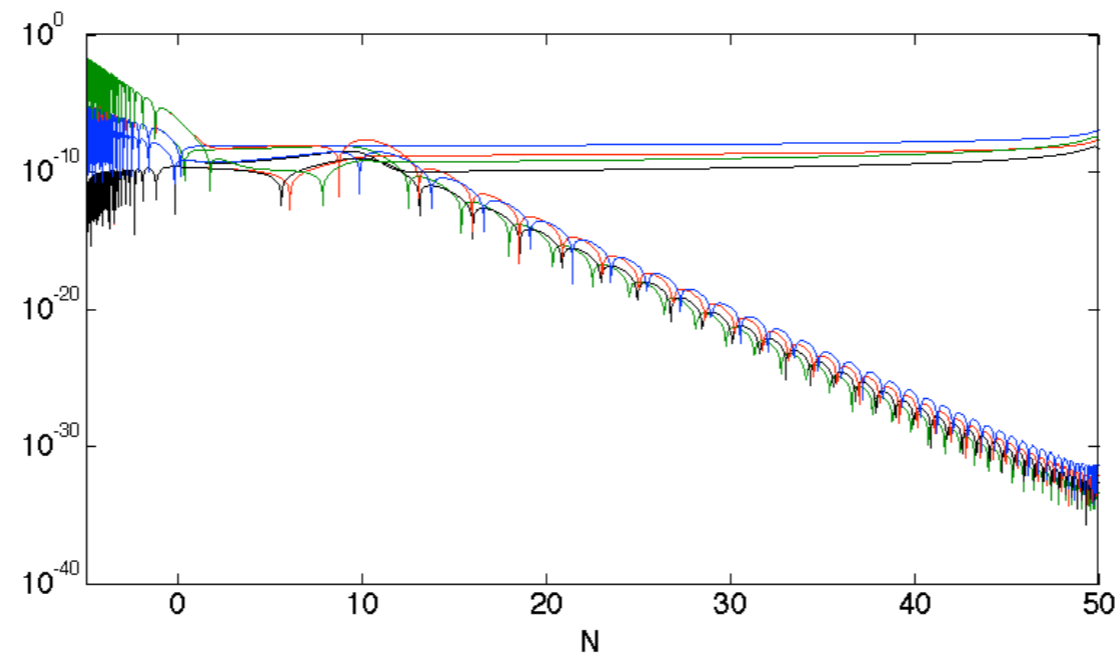
- Usual method Salopek and bond (1985):

$$Q^I(k) = \Psi^I_L(t, k) a^L(k) + \Psi^{*I}_L(t, k) a^{\dagger L}(-k)$$

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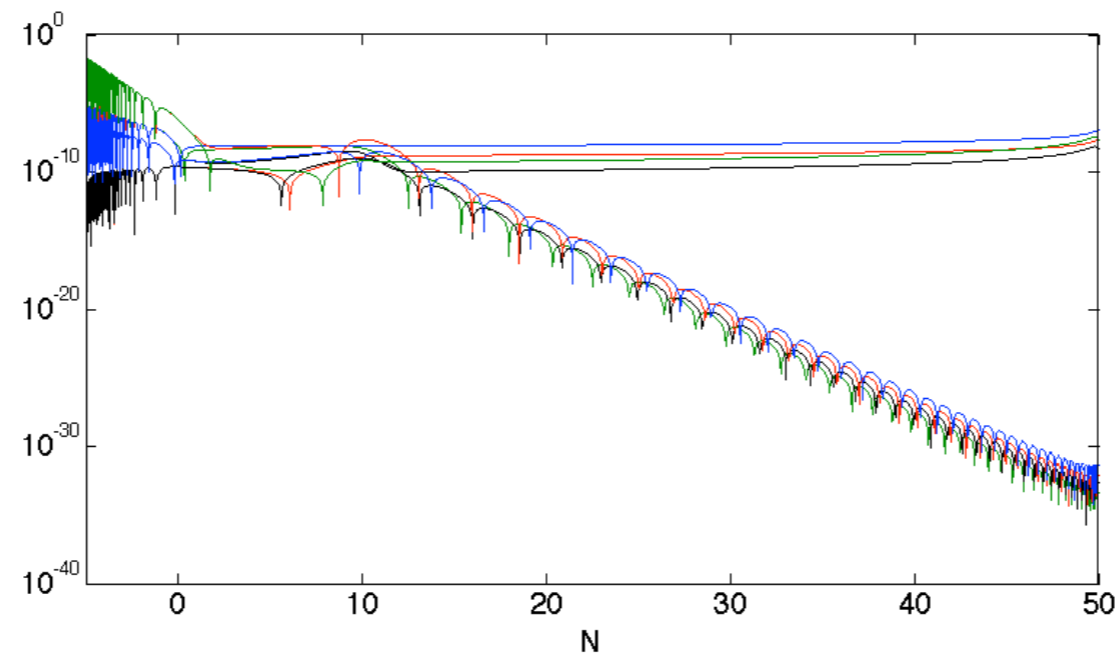
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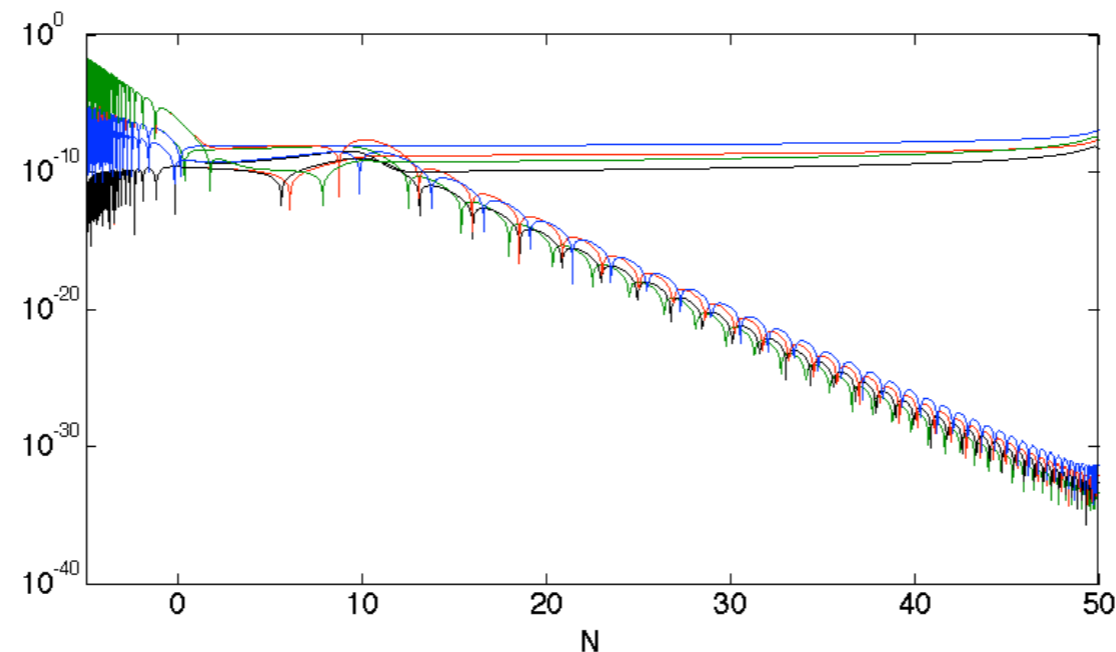


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$$\Sigma^{IJ} = \Psi^{IL} \Psi^{*J}_L$$

$$\langle Q^I Q^J Q^L \rangle = -i \int_{-\infty}^t dt' \langle [Q^I Q^J Q^L, H_{\text{int}}(t')] \rangle$$

Maldacena (2003)