

Non-Gaussianity

Beyond the CMB power spectrum

*James Fergusson
& Paul Shellard*

*Centre for Theoretical
Cosmology, DAMTP
University of Cambridge*

LECTURE 1

*Review: CMB and Inflation
Inflationary fluctuations
& Randomness*

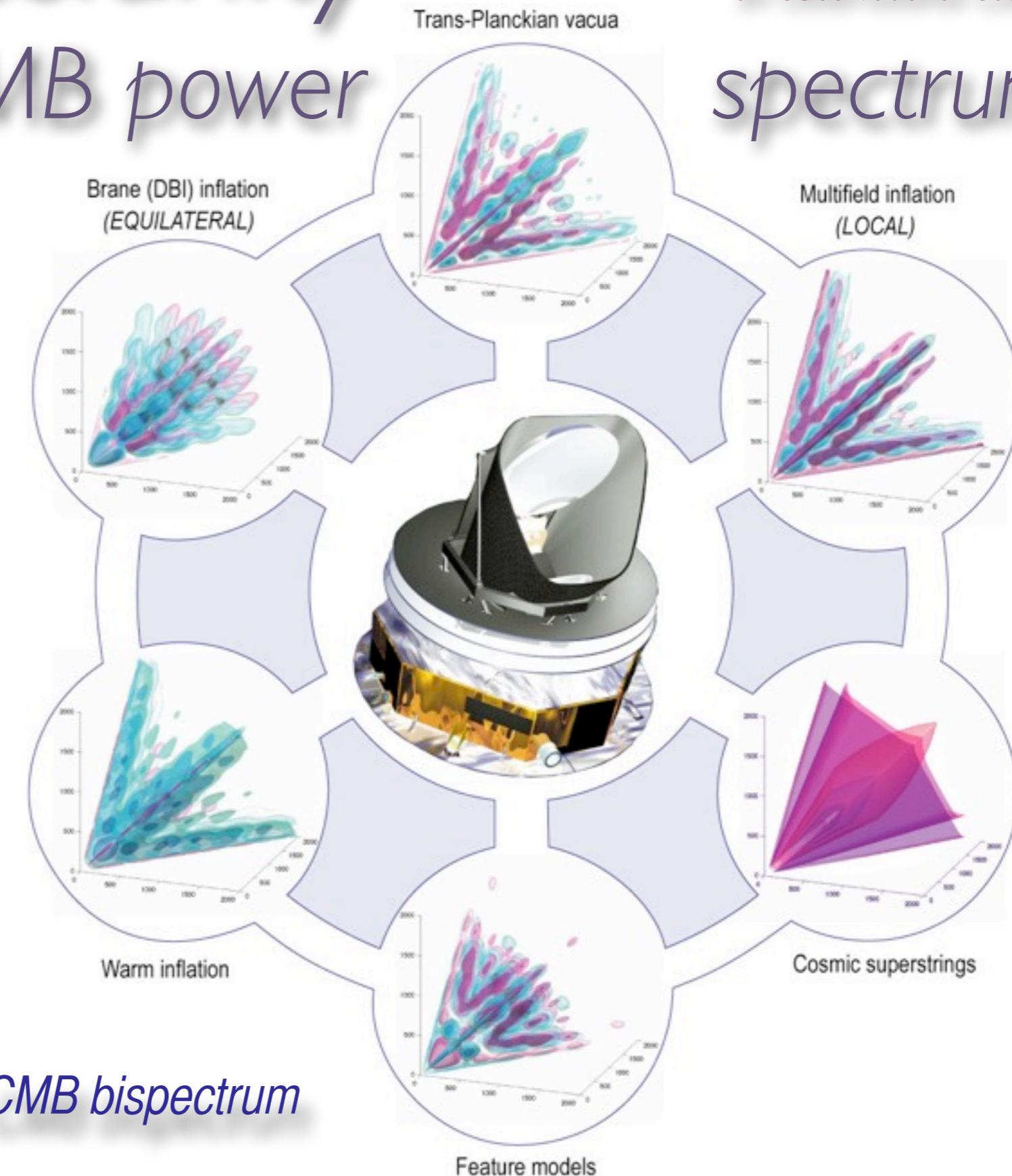
Primordial non-Gaussianity

The CMB bispectrum

CMB bispectrum estimation

General modal approach to CMB bispectrum

The
Azores
School
on Observational Cosmology



Non-Gaussianity

Beyond the CMB power spectrum

*James Fergusson
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*Centre for Theoretical
Cosmology, DAMTP
University of Cambridge
work with Michele Liguori
Donough Regan
& Marcel Schmittfull
also Hiro Funakoshi*

[arXiv:1006.1642](https://arxiv.org/abs/1006.1642)

[arXiv:1012.6039](https://arxiv.org/abs/1012.6039)

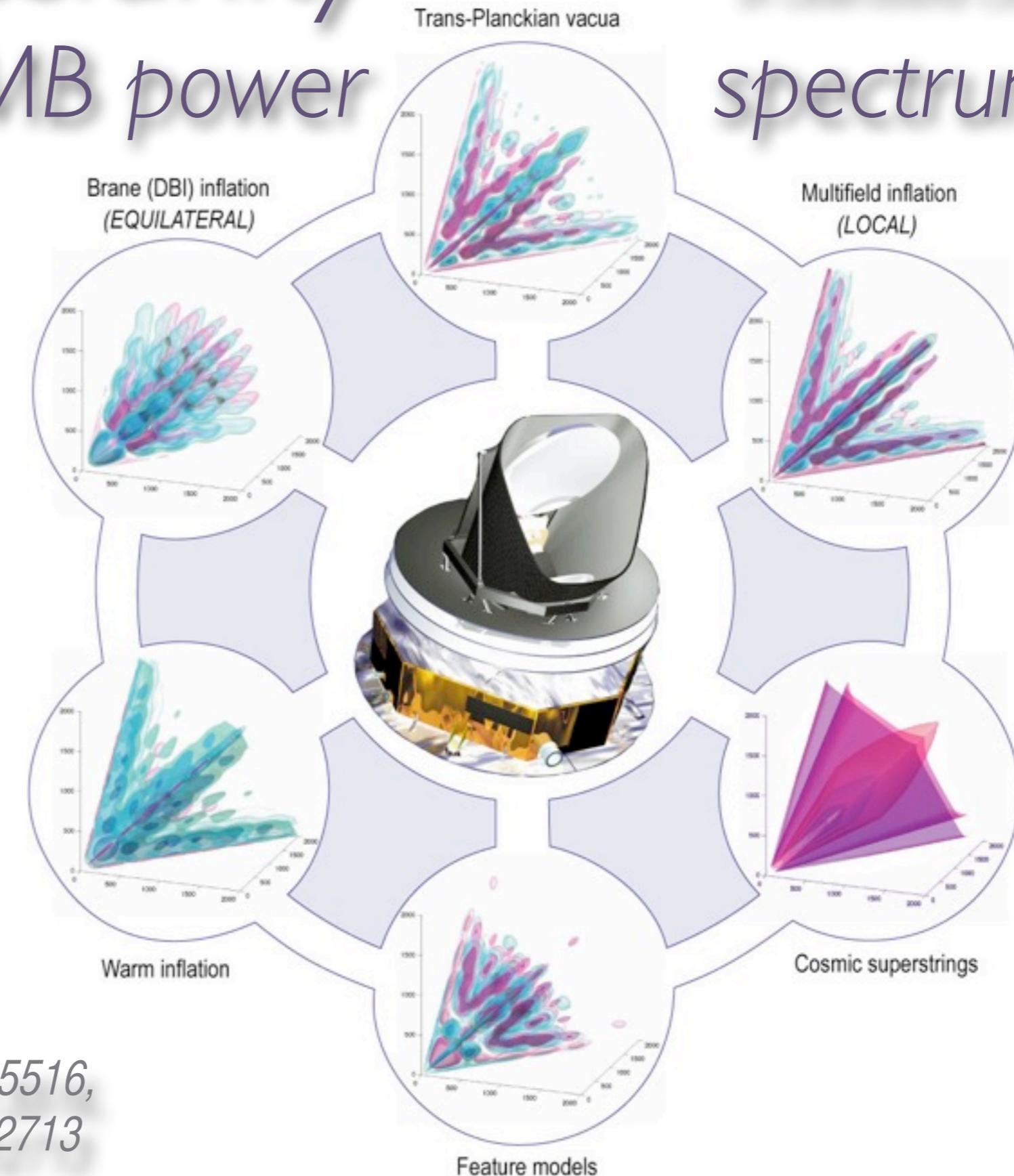
[arXiv:1108.3813](https://arxiv.org/abs/1108.3813)

[arXiv: 1105.2791](https://arxiv.org/abs/1105.2791)

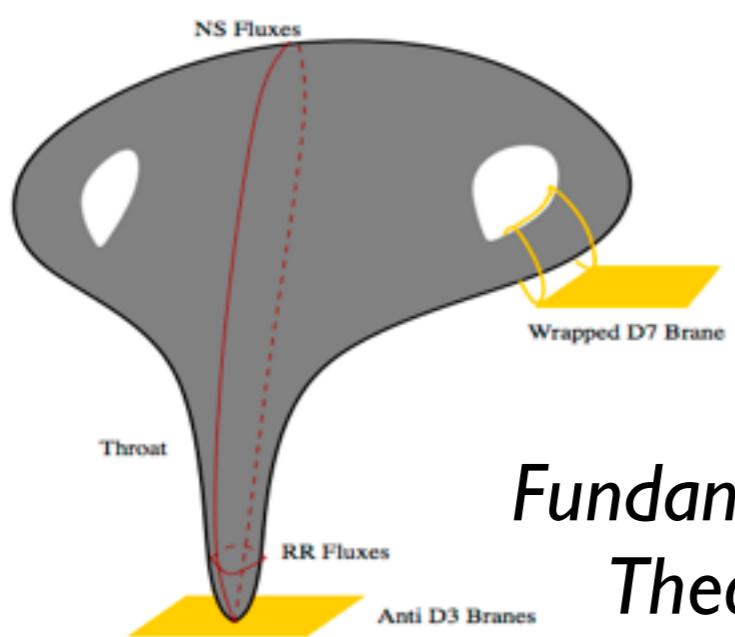
[arXiv:1008.1730](https://arxiv.org/abs/1008.1730)

([arXiv:1004.2915](https://arxiv.org/abs/1004.2915)), [arXiv:0912.5516](https://arxiv.org/abs/0912.5516),
[arXiv:0812.3413](https://arxiv.org/abs/0812.3413), [astro-ph/0612713](https://arxiv.org/abs/astro-ph/0612713)

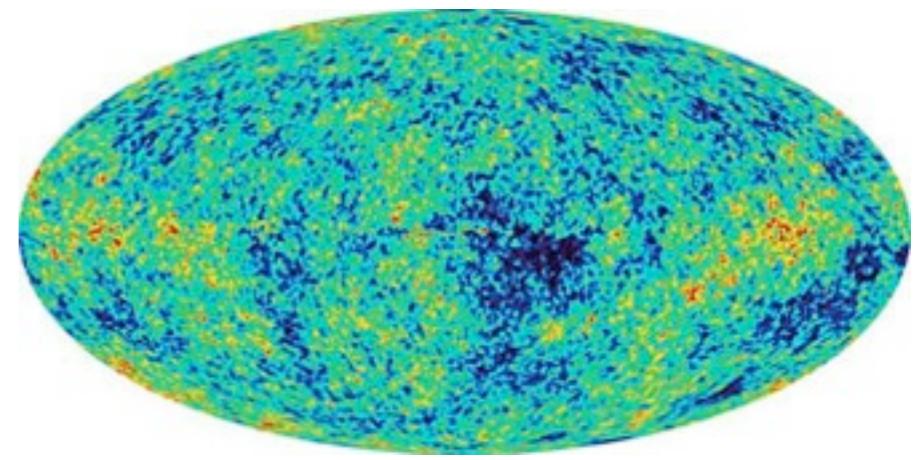
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Motivation



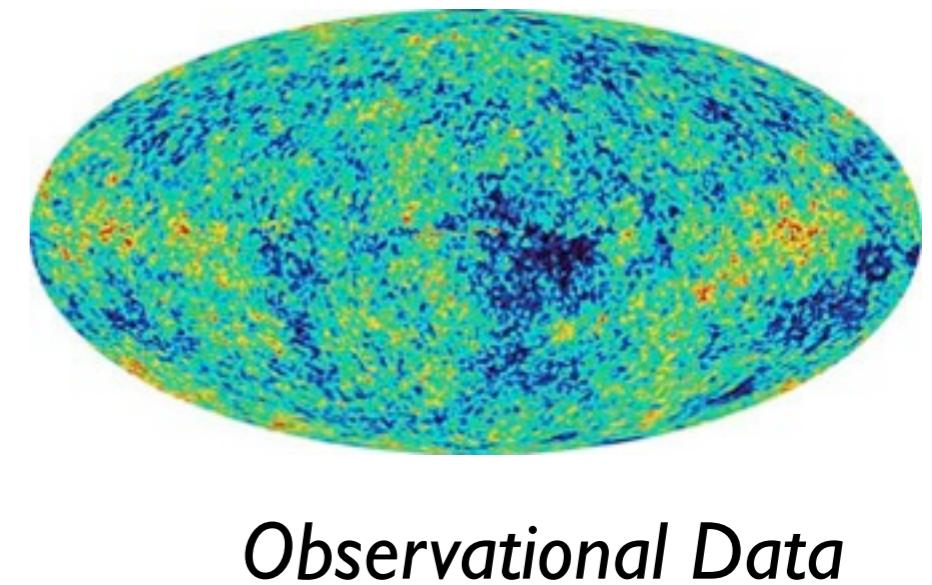
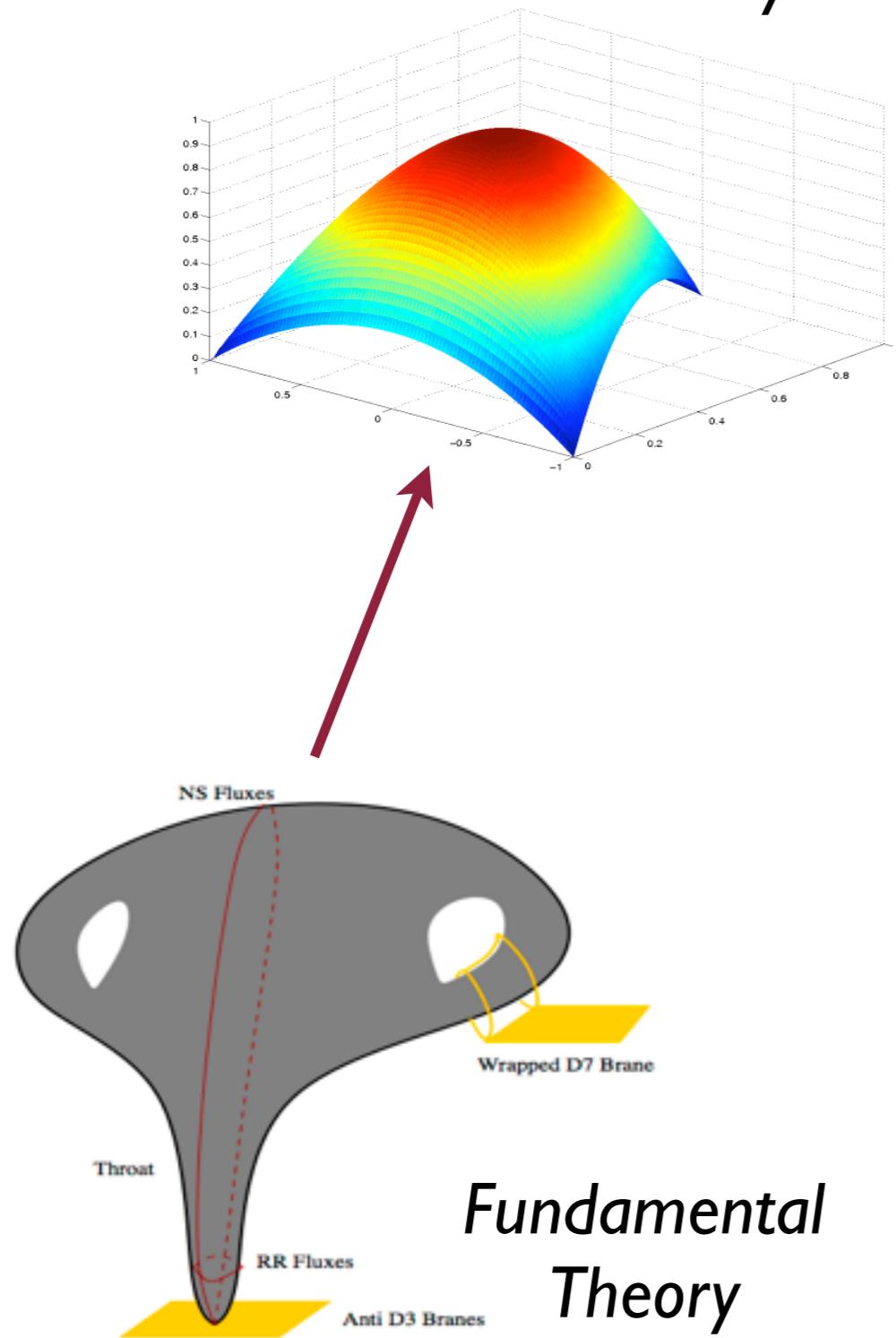
*Fundamental
Theory*



Observational Data

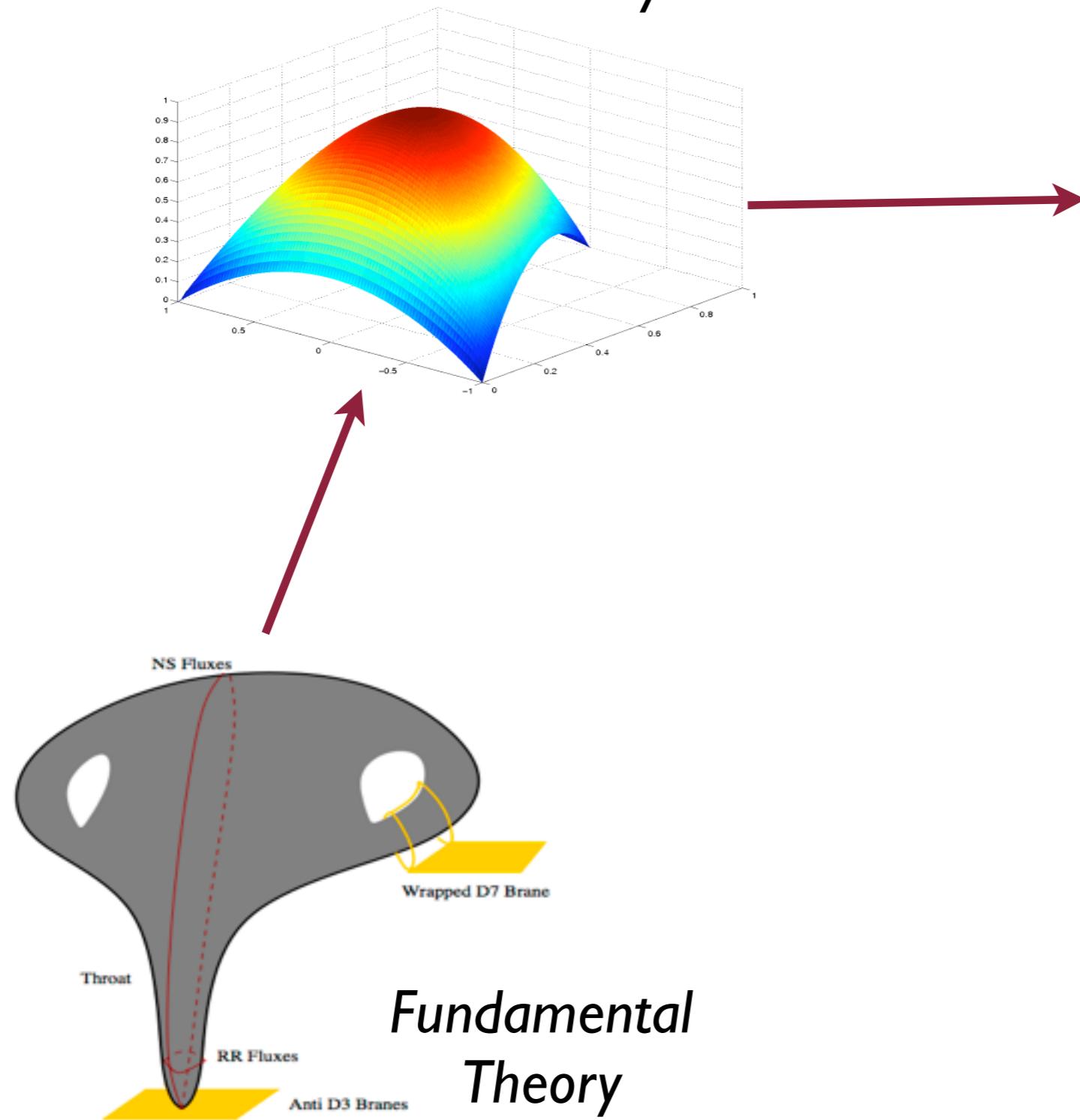
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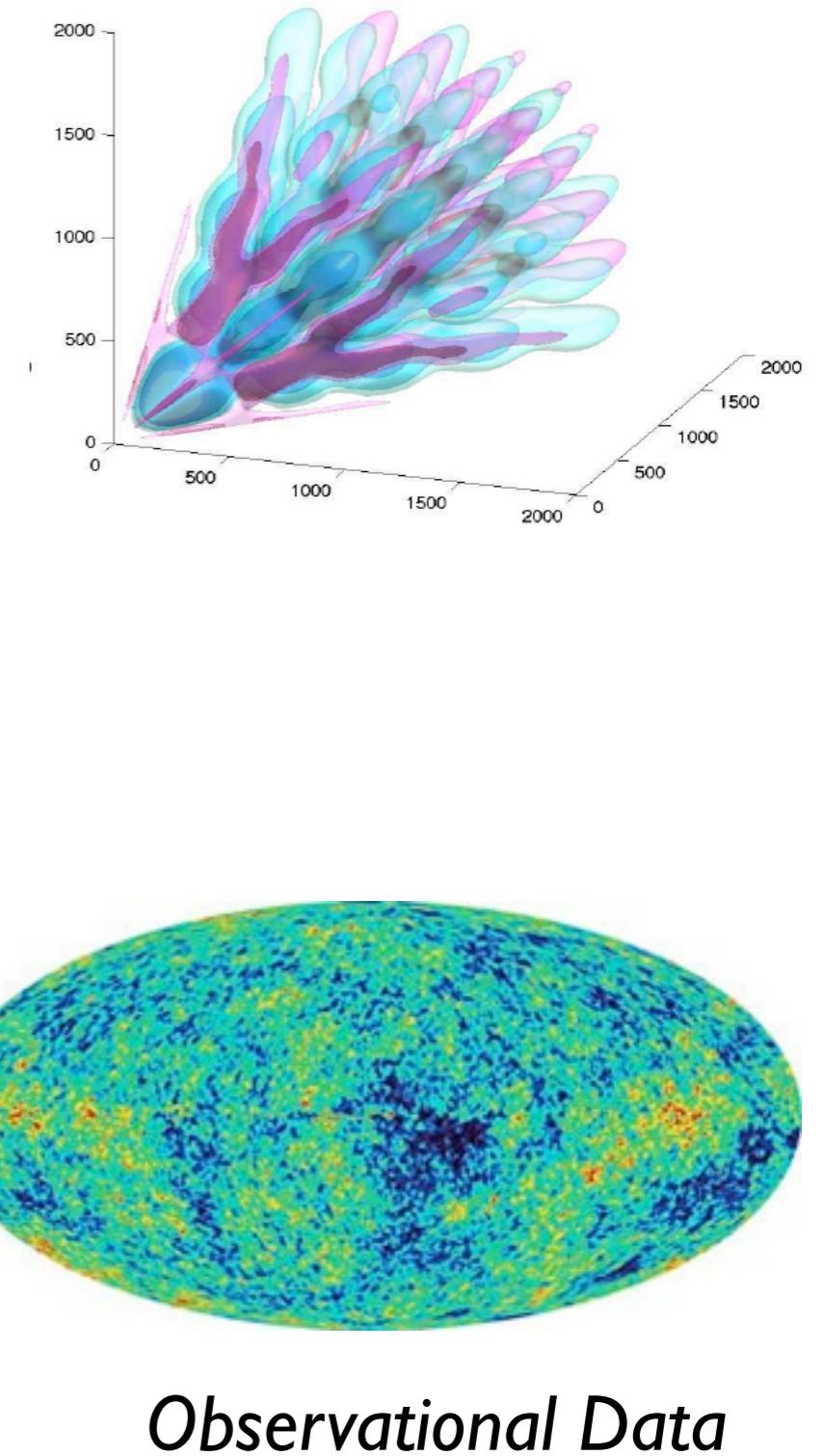


Motivation

Primordial non-Gaussianity



CMB (or LSS) fingerprint

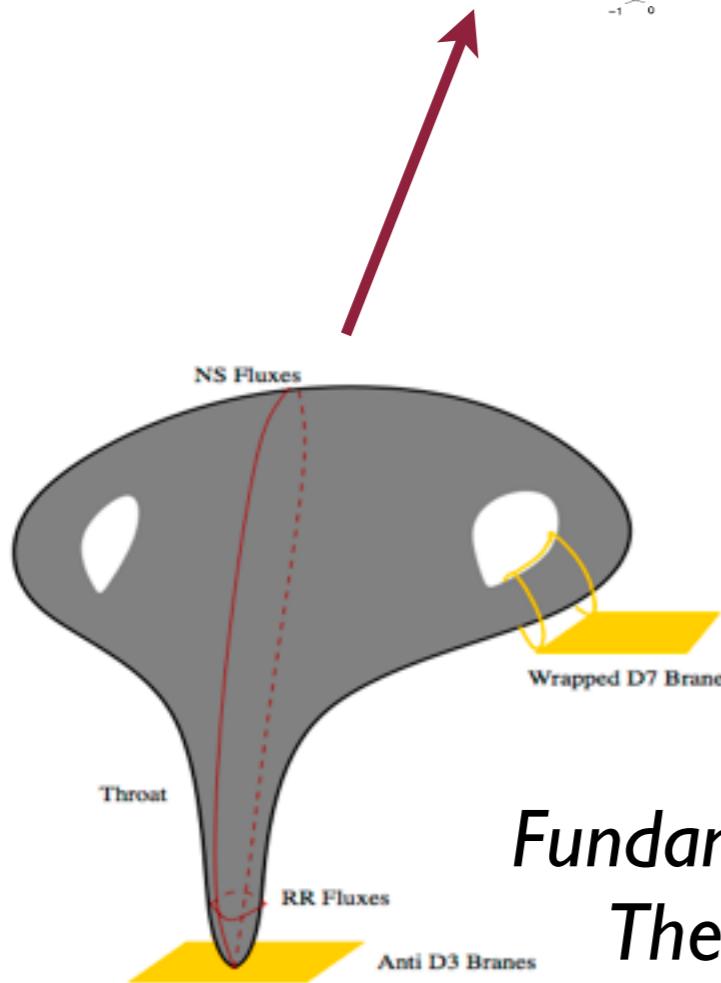
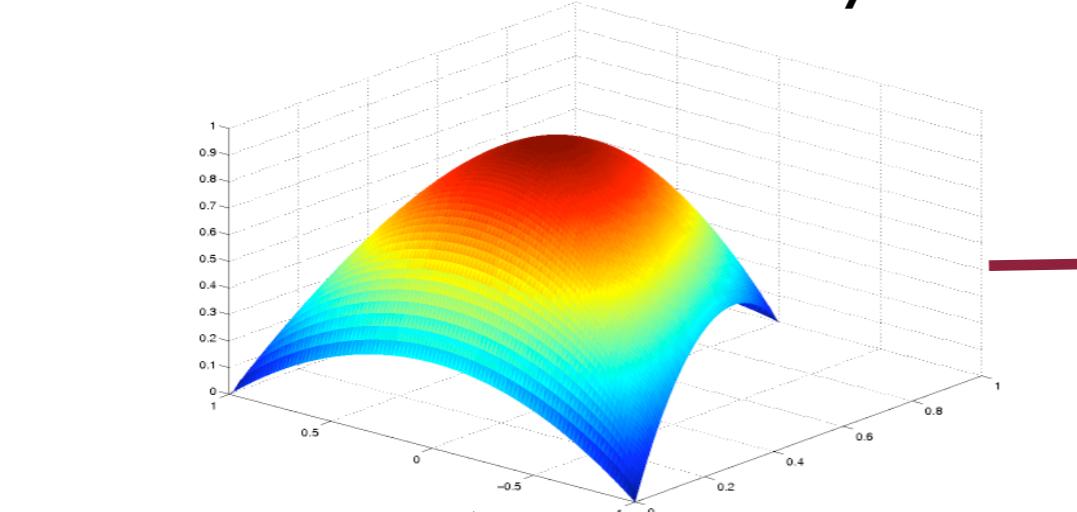


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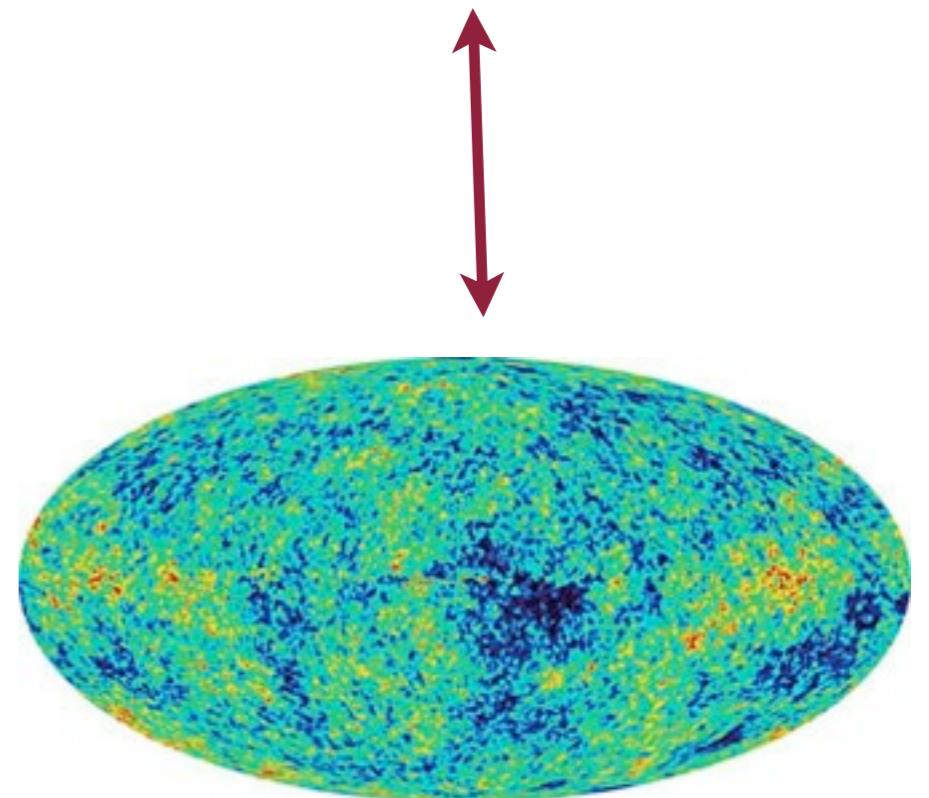
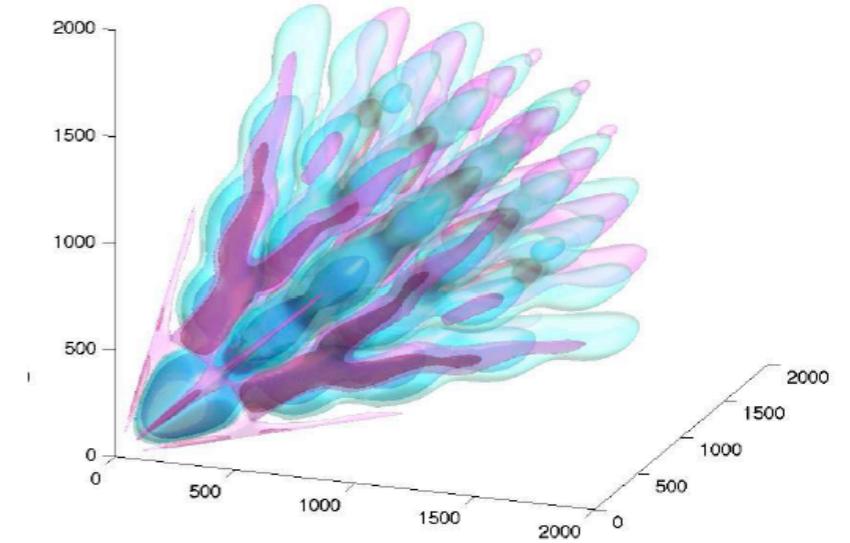
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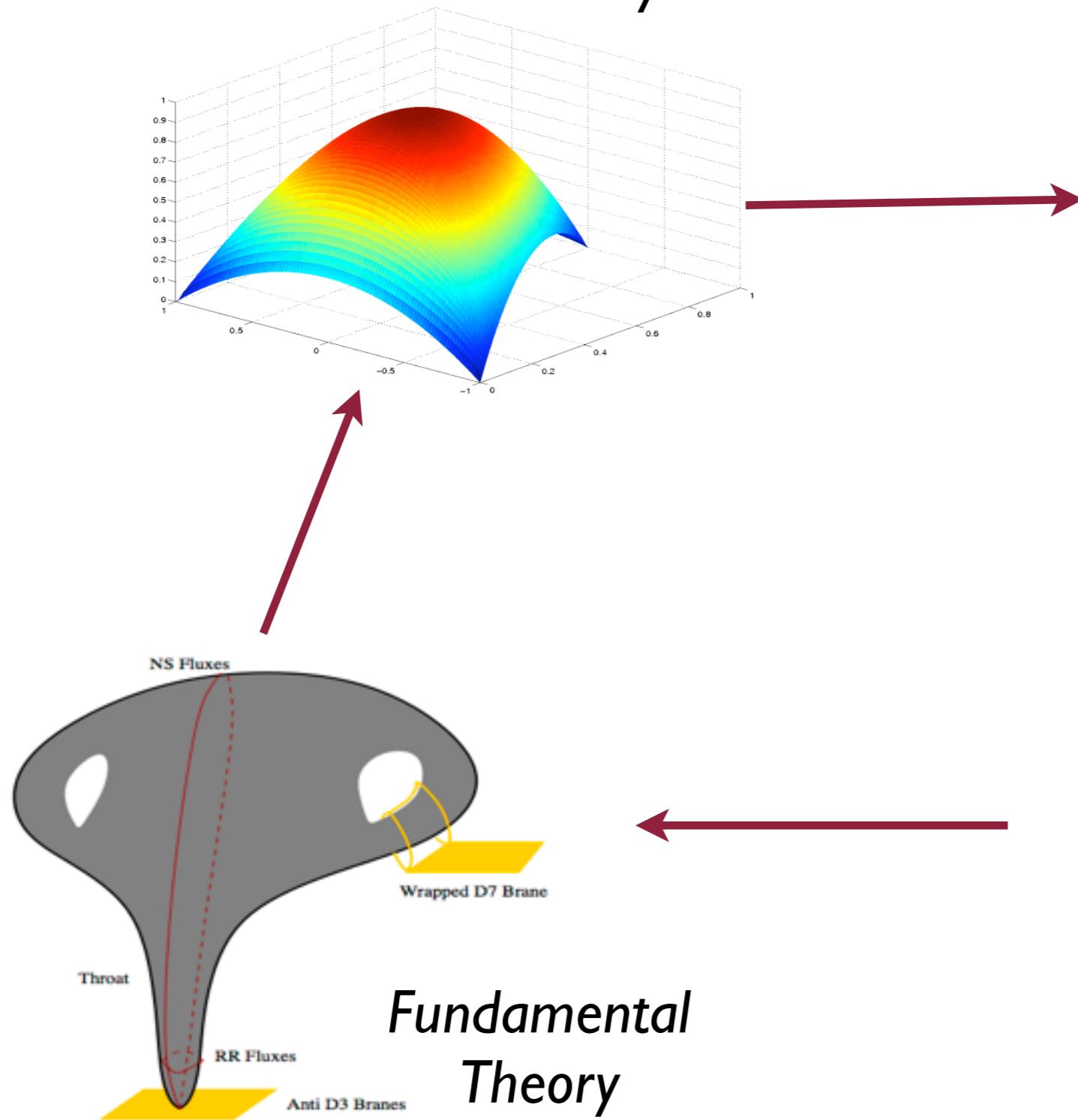
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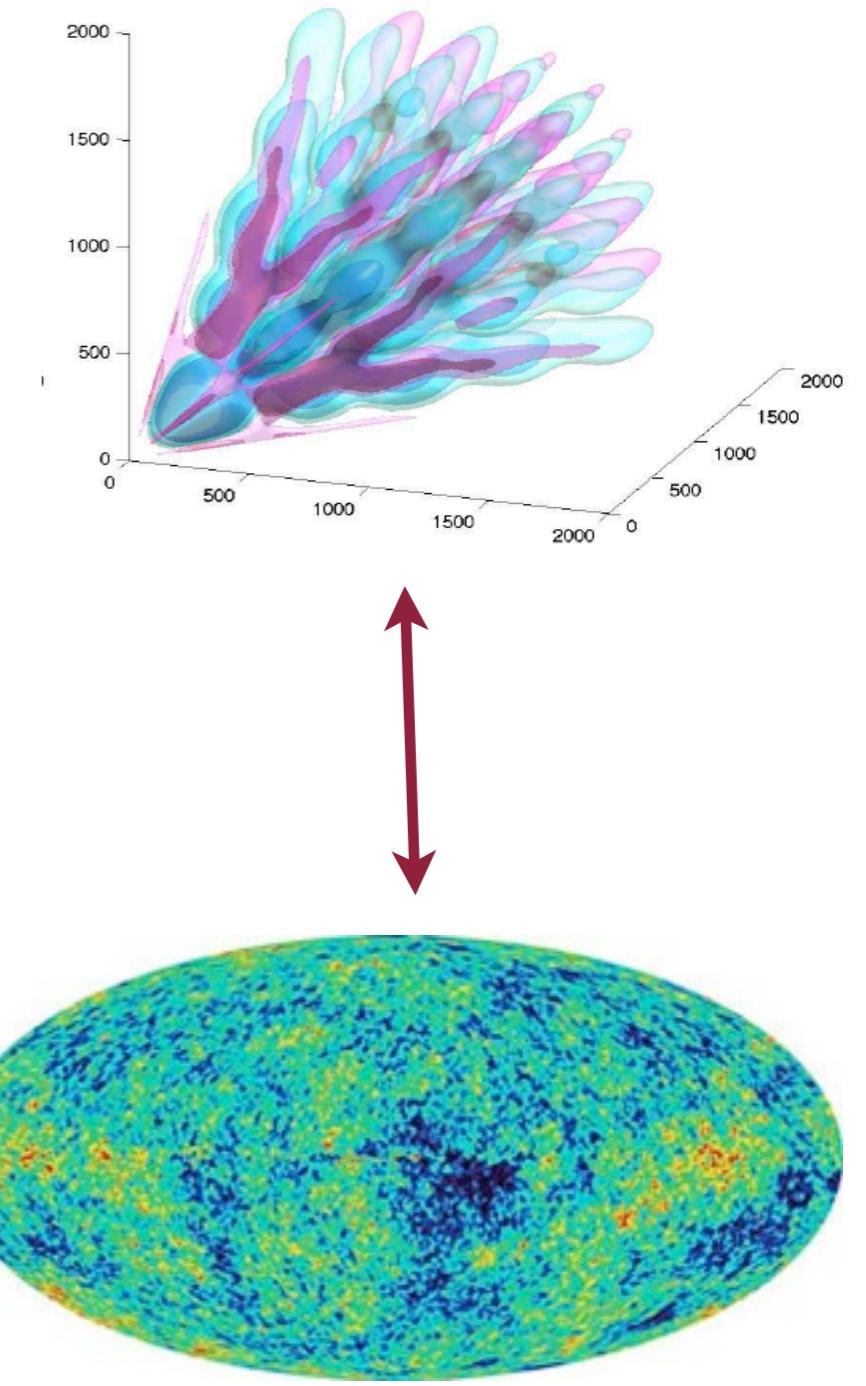
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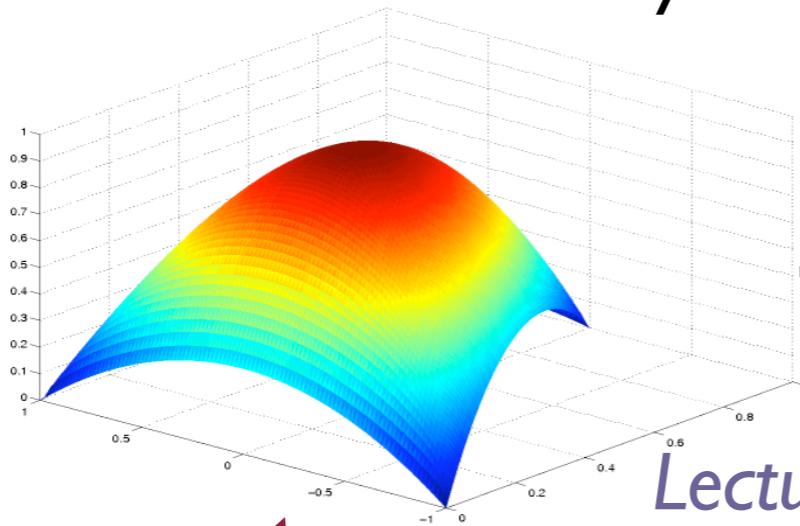
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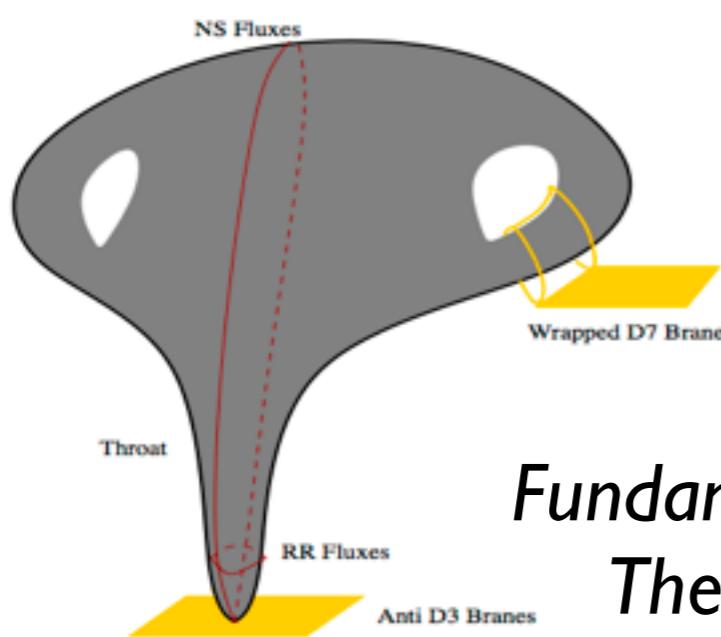
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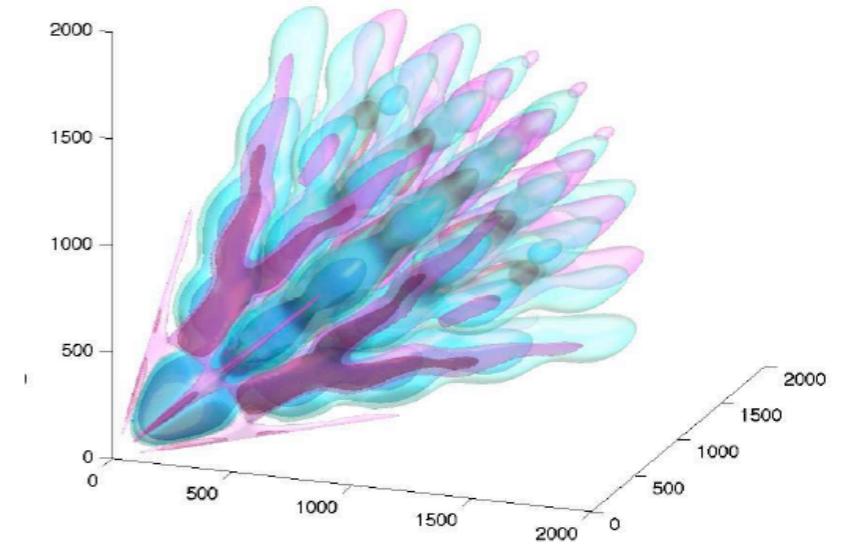


Lecture 1

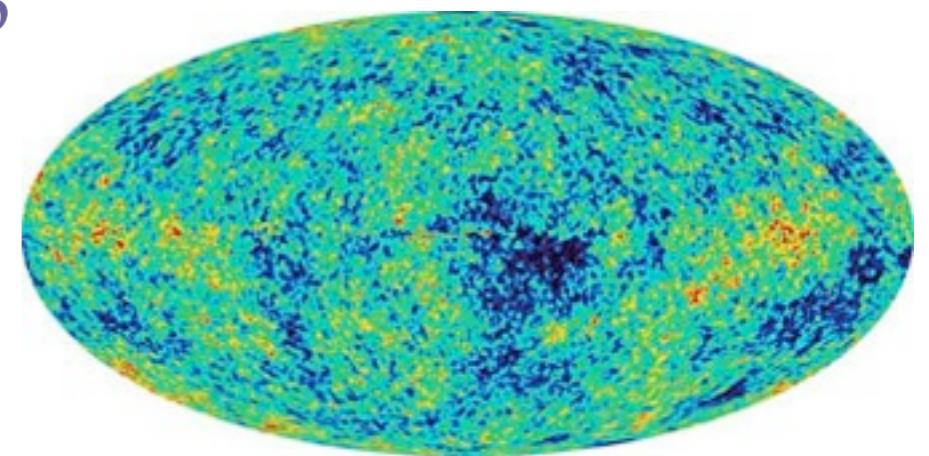


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Lecture 2

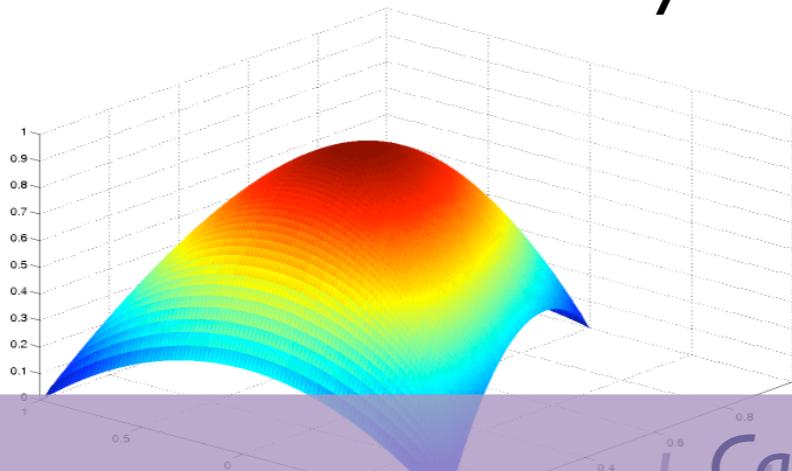


Lecture 3

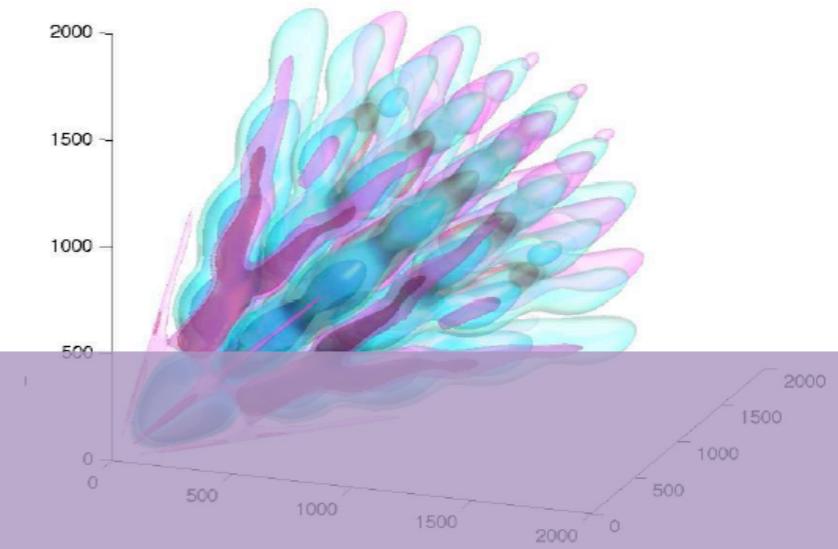
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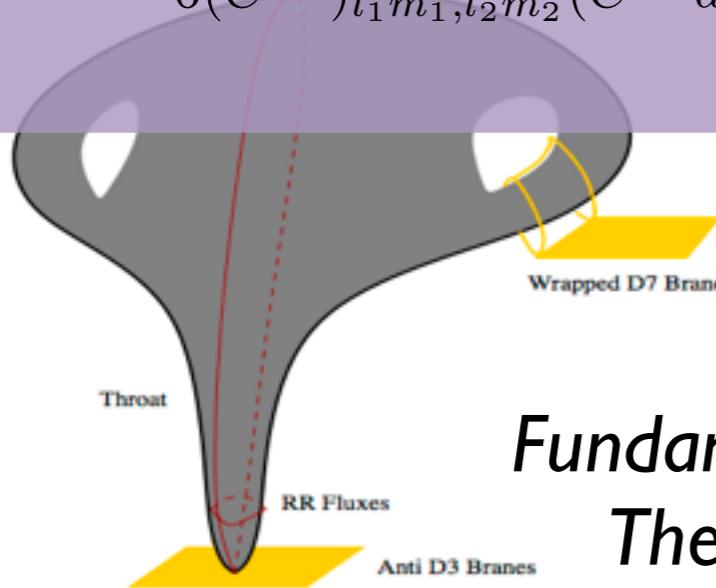
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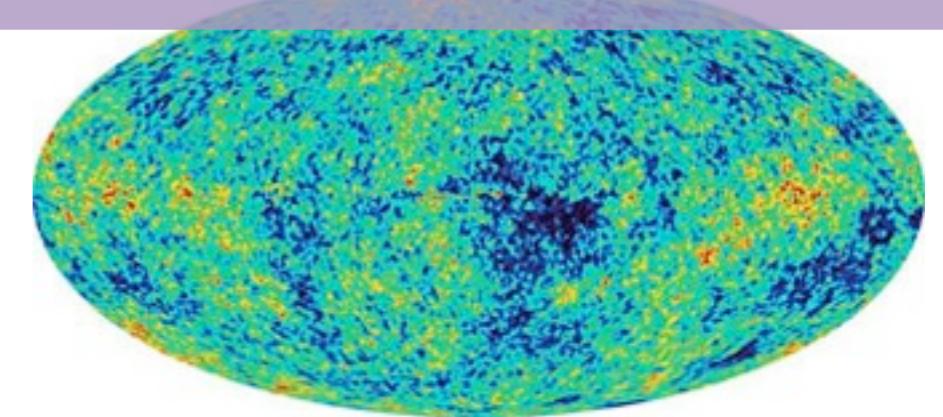
Lecture 1

$$\mathcal{E}^{\text{general}} = \frac{f_{\text{sky}}}{\tilde{N}} \sum_{l_i m_i} \langle a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3} a_{l_4 m_4} \rangle_c \left[(C^{-1} a^{\text{obs}})_{l_1 m_1} (C^{-1} a^{\text{obs}})_{l_2 m_2} (C^{-1} a^{\text{obs}})_{l_3 m_3} (C^{-1} a^{\text{obs}})_{l_4 m_4} \right. \\ \left. - 6(C^{-1})_{l_1 m_1, l_2 m_2} (C^{-1} a^{\text{obs}})_{l_3 m_3} (C^{-1} a^{\text{obs}})_{l_4 m_4} + 3(C^{-1})_{l_1 m_1, l_2 m_2} (C^{-1})_{l_3 m_3, l_4 m_4} \right],$$

Lecture 3

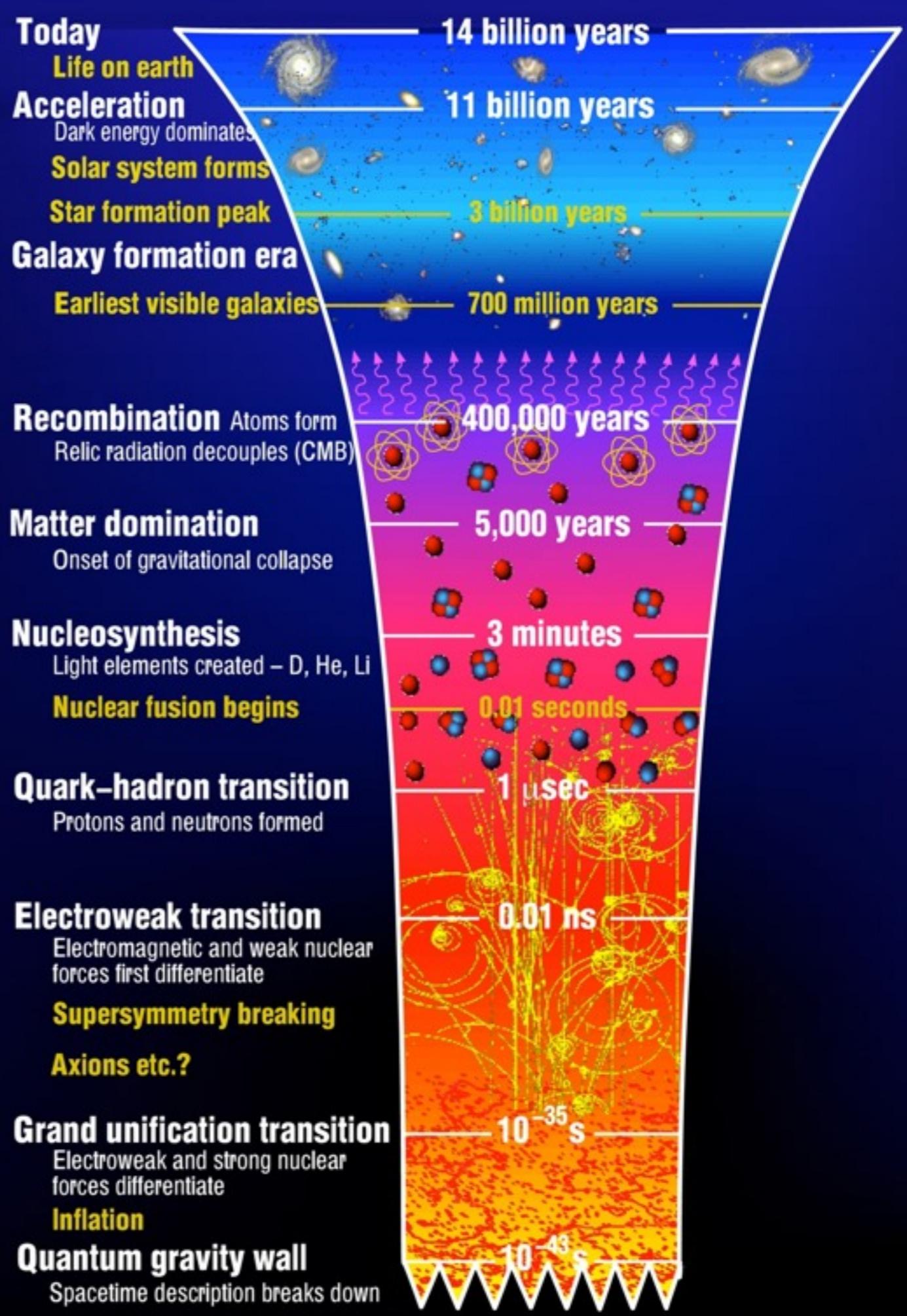


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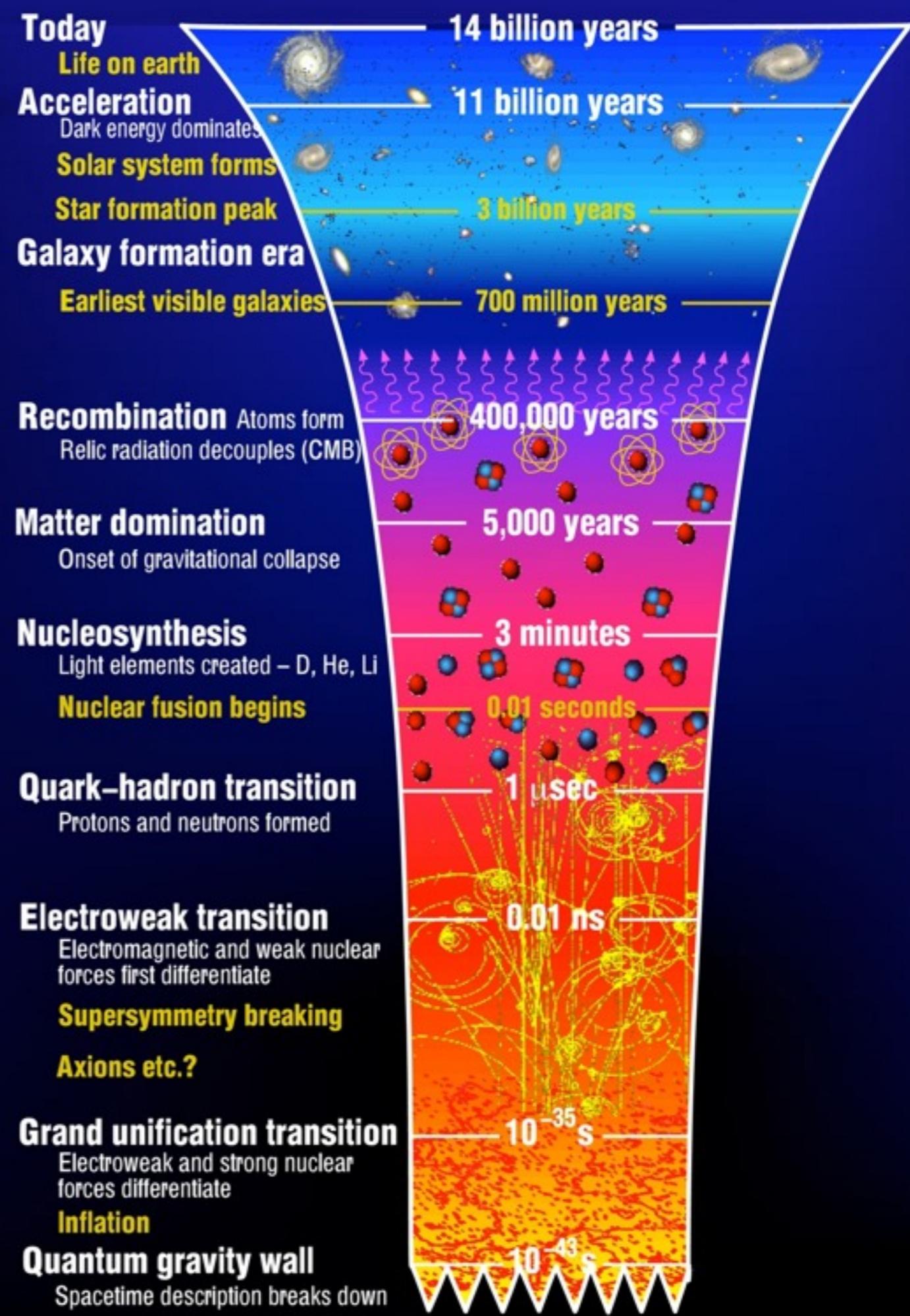
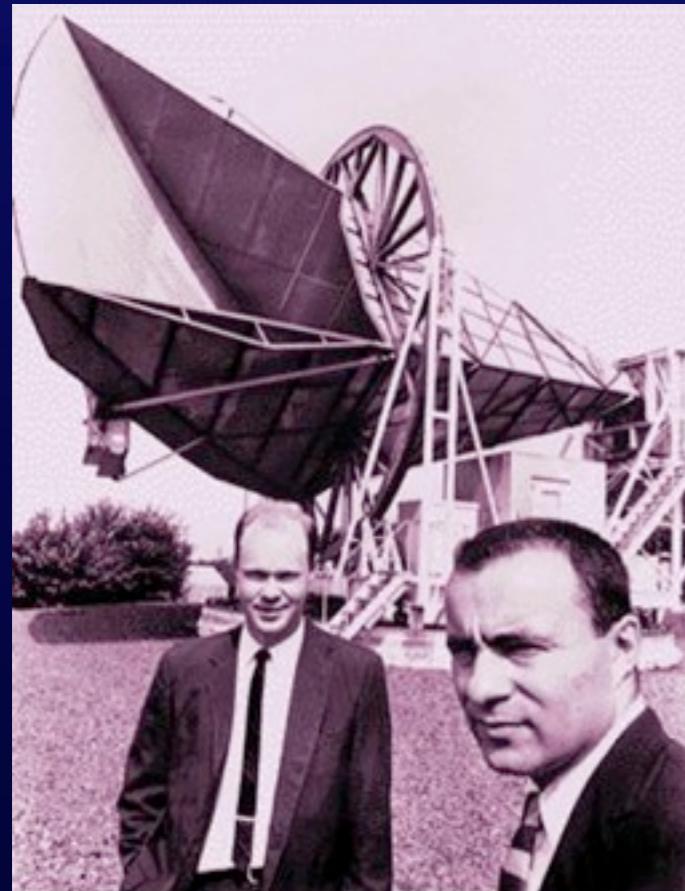
Observational Data

The standard cosmology ...



The standard cosmology ...

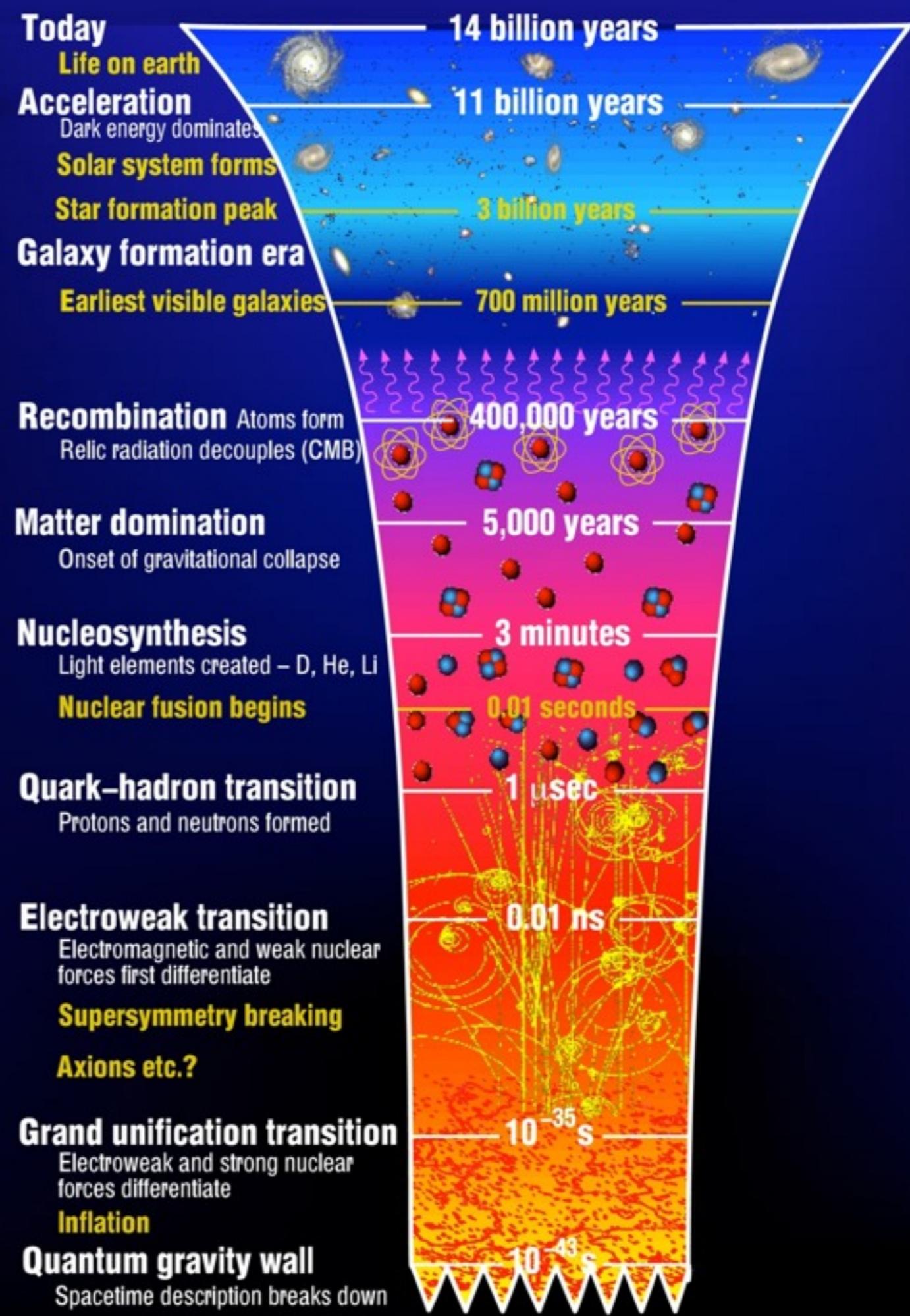
CMB discovery - 1965



The standard cosmology ...

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Precision cosmology - WMAP 2003



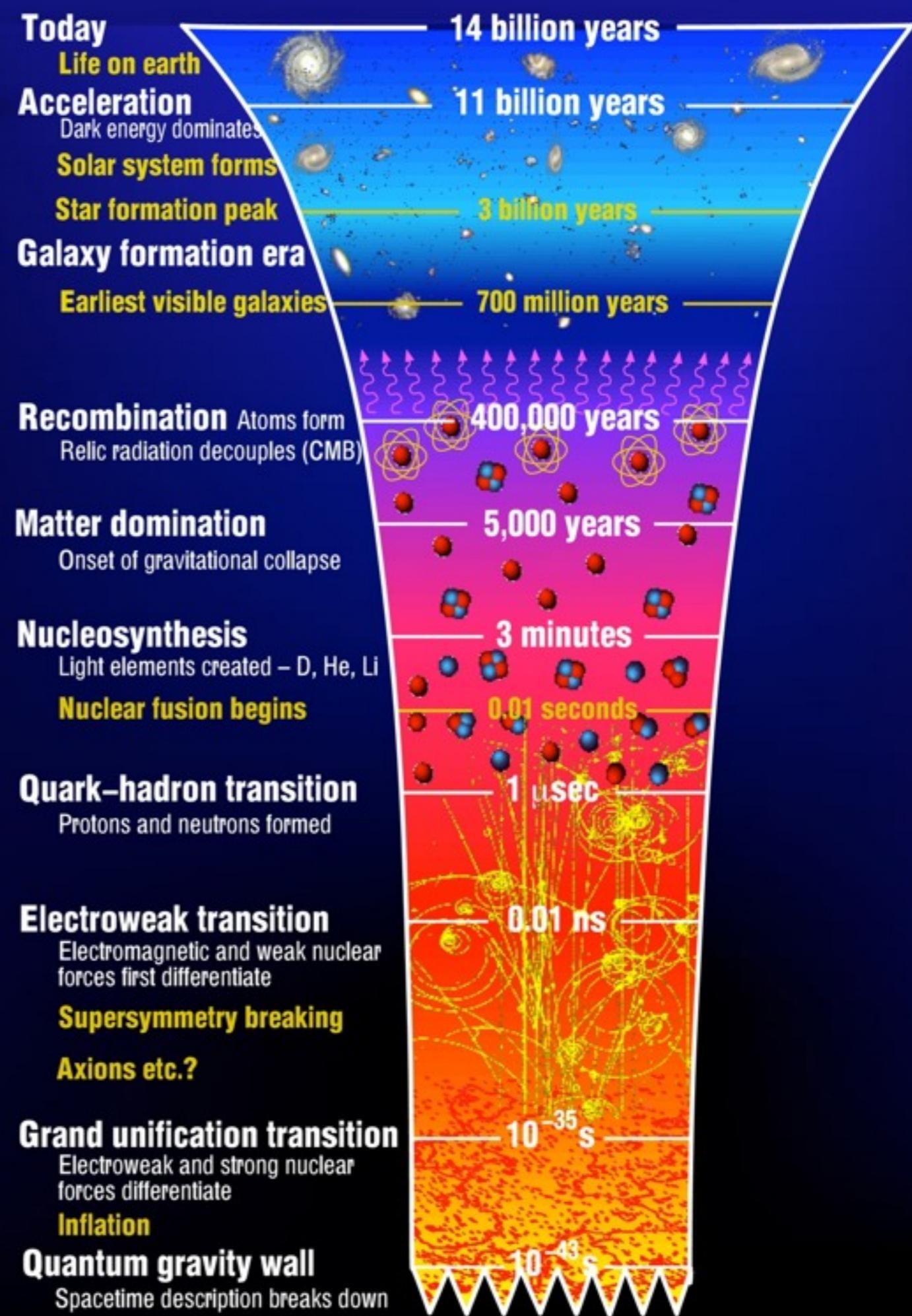
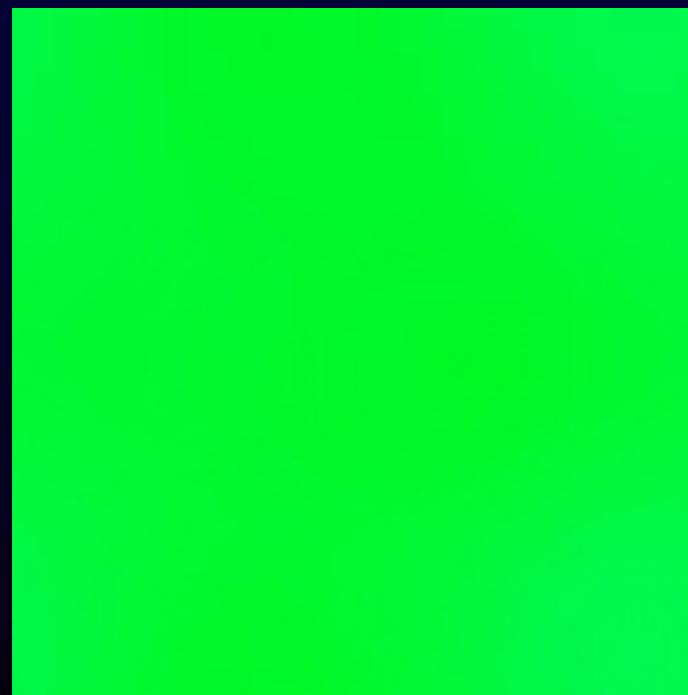
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With inflationary fluctuations



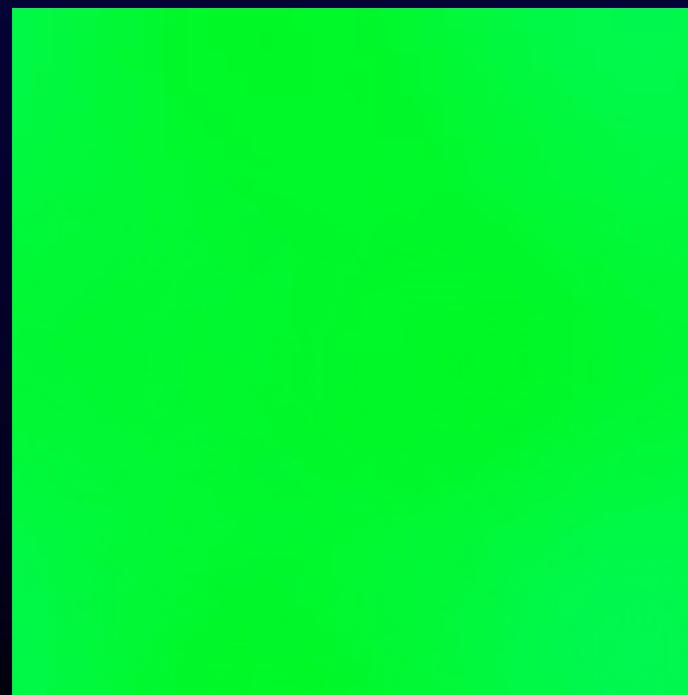
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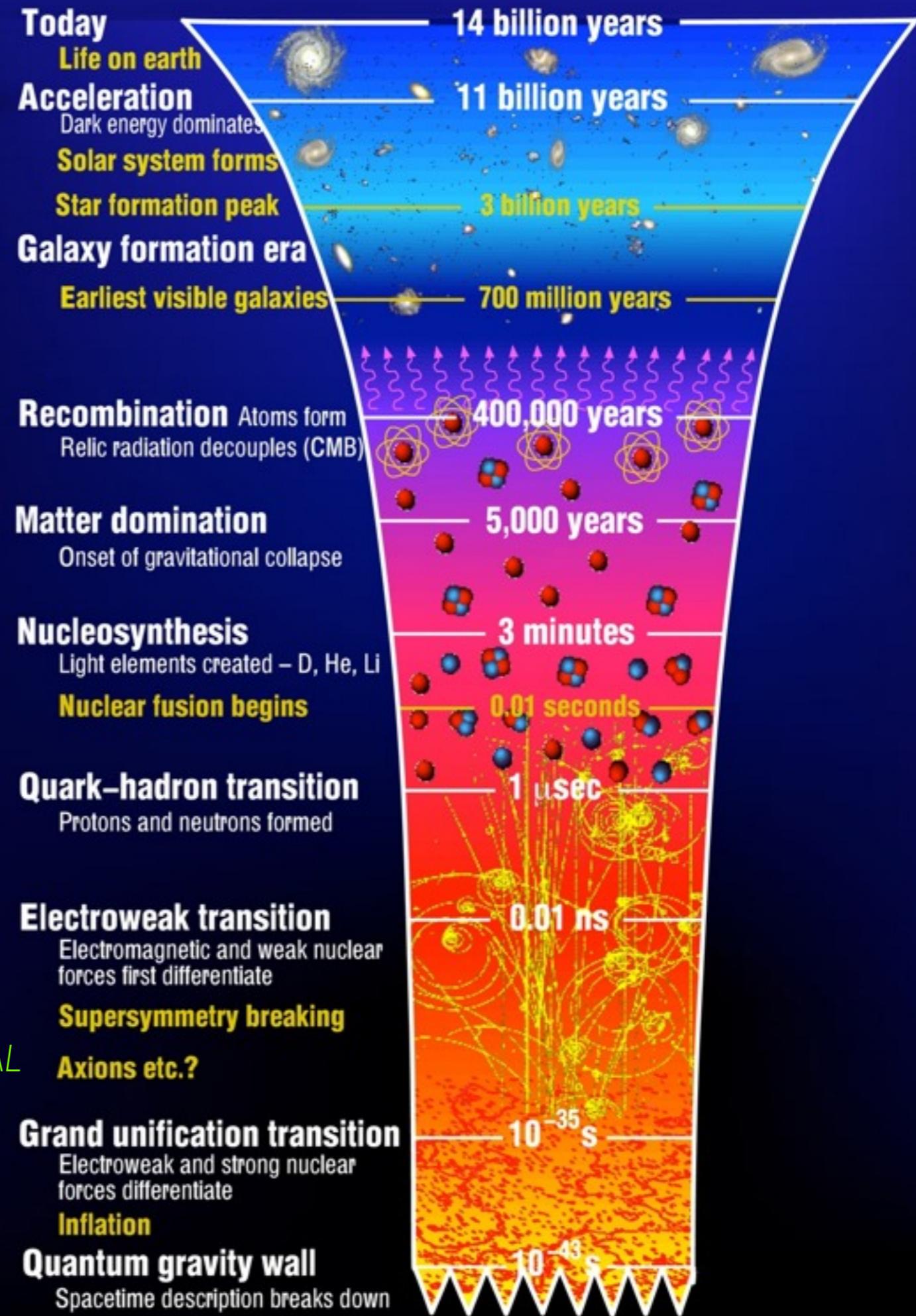
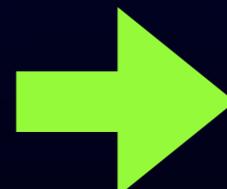
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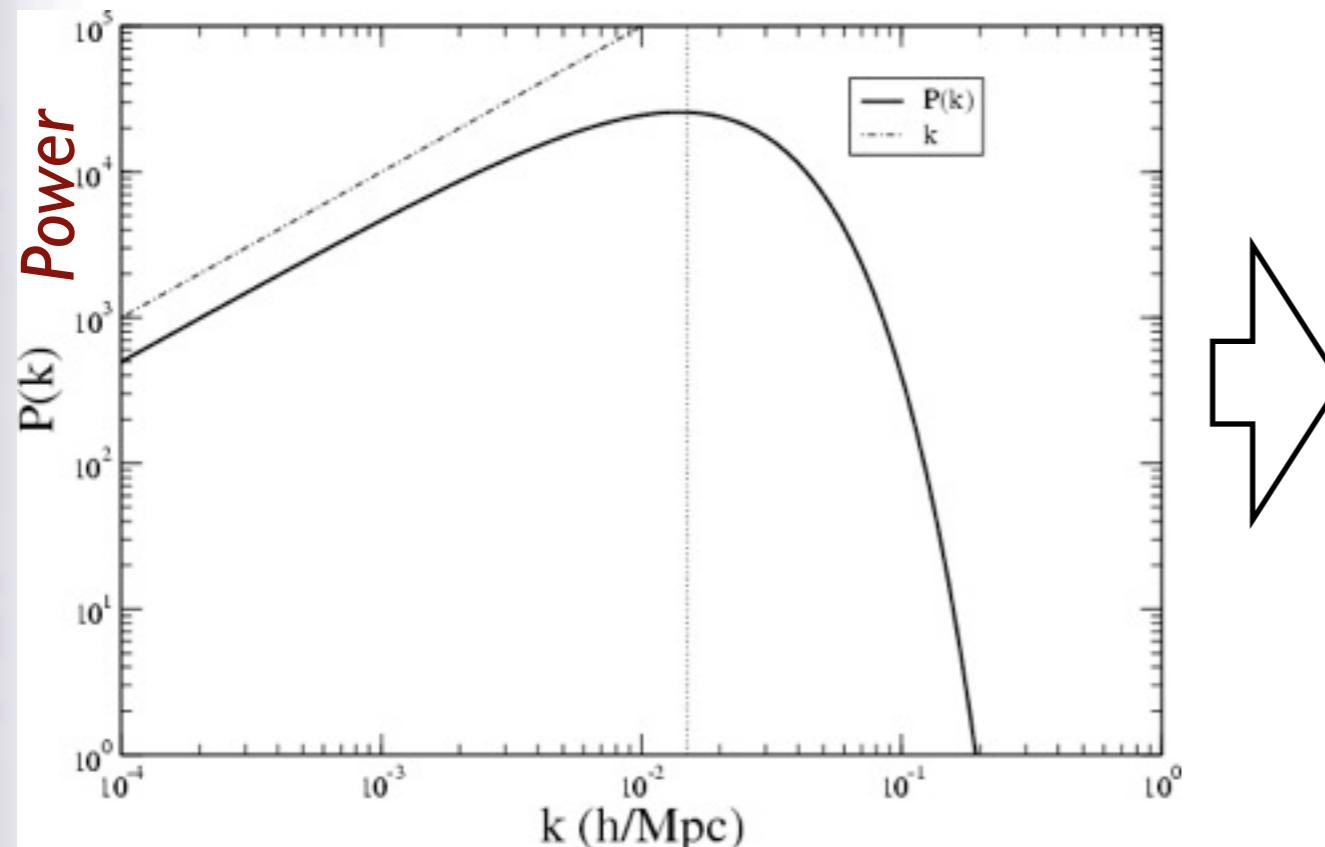
FUNDAMENTAL
THEORY



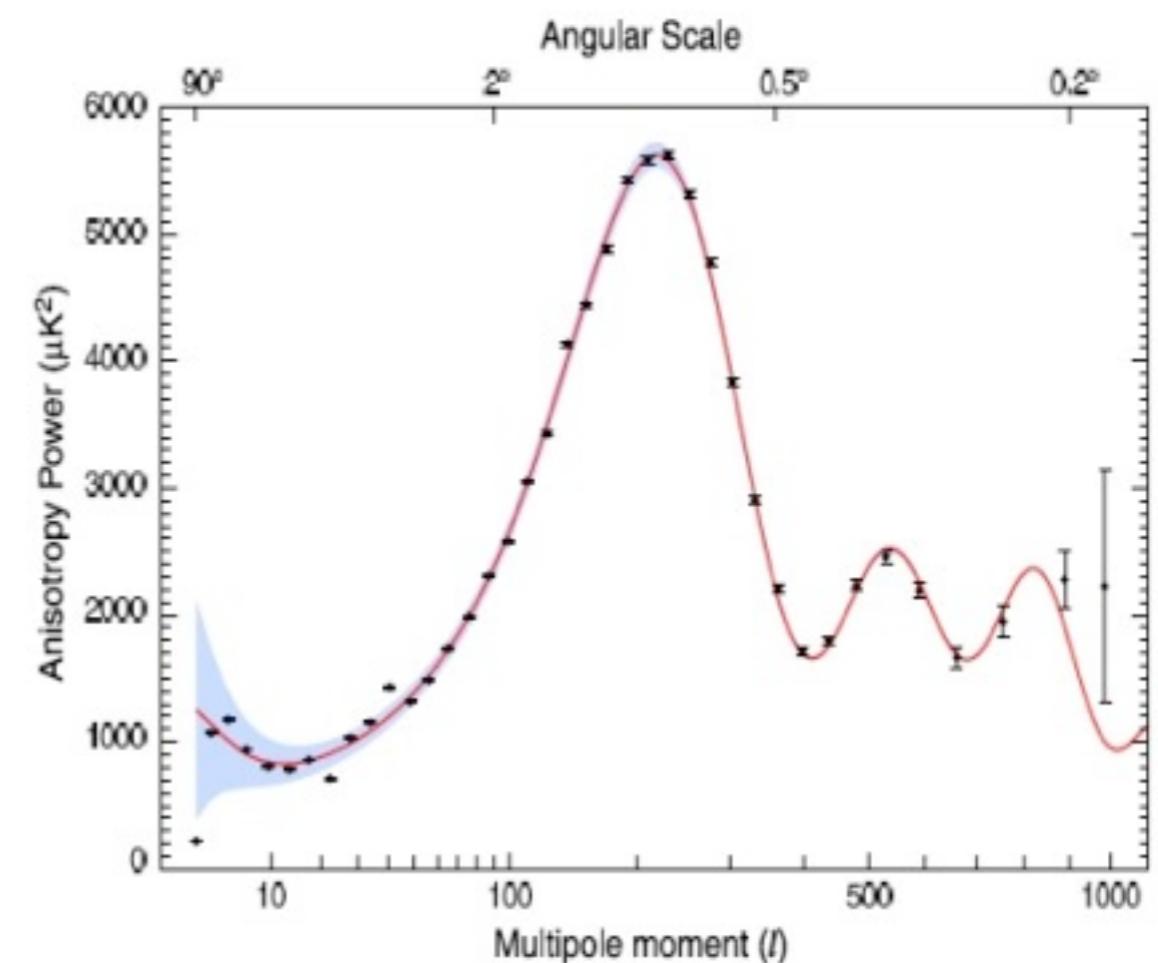
WMAP data analysis

Why is the CMB so good for cosmology? Answer: **Linearity!**

Assume simple primordial model - inflation (small scale-free seeds)



Simple initial power spectrum



CMB angular power spectrum

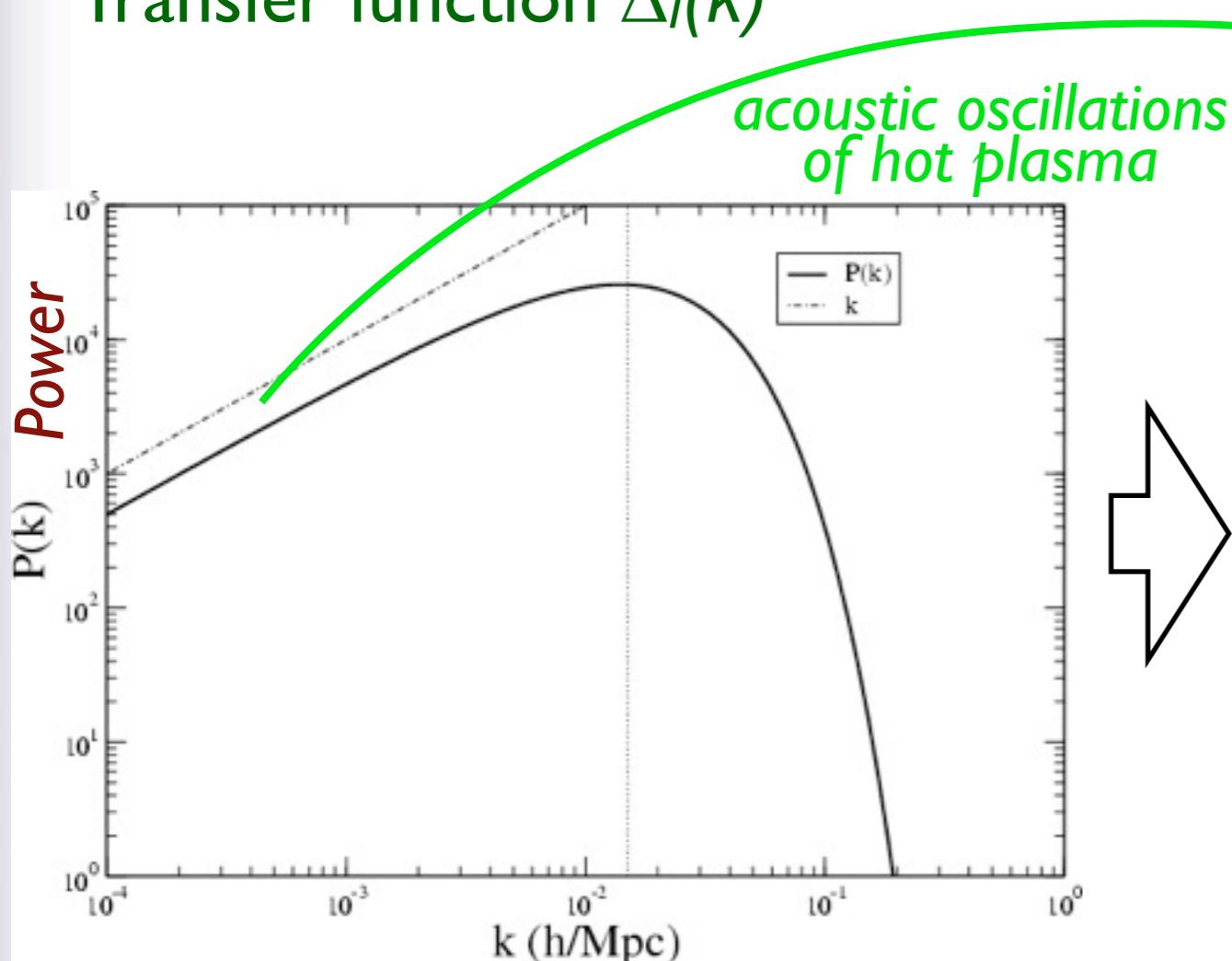
Transfer function $\Delta_l(k)$ depends on $H_0, \Omega, \Omega_\Delta, \Omega_M, \Omega_B$, etc.

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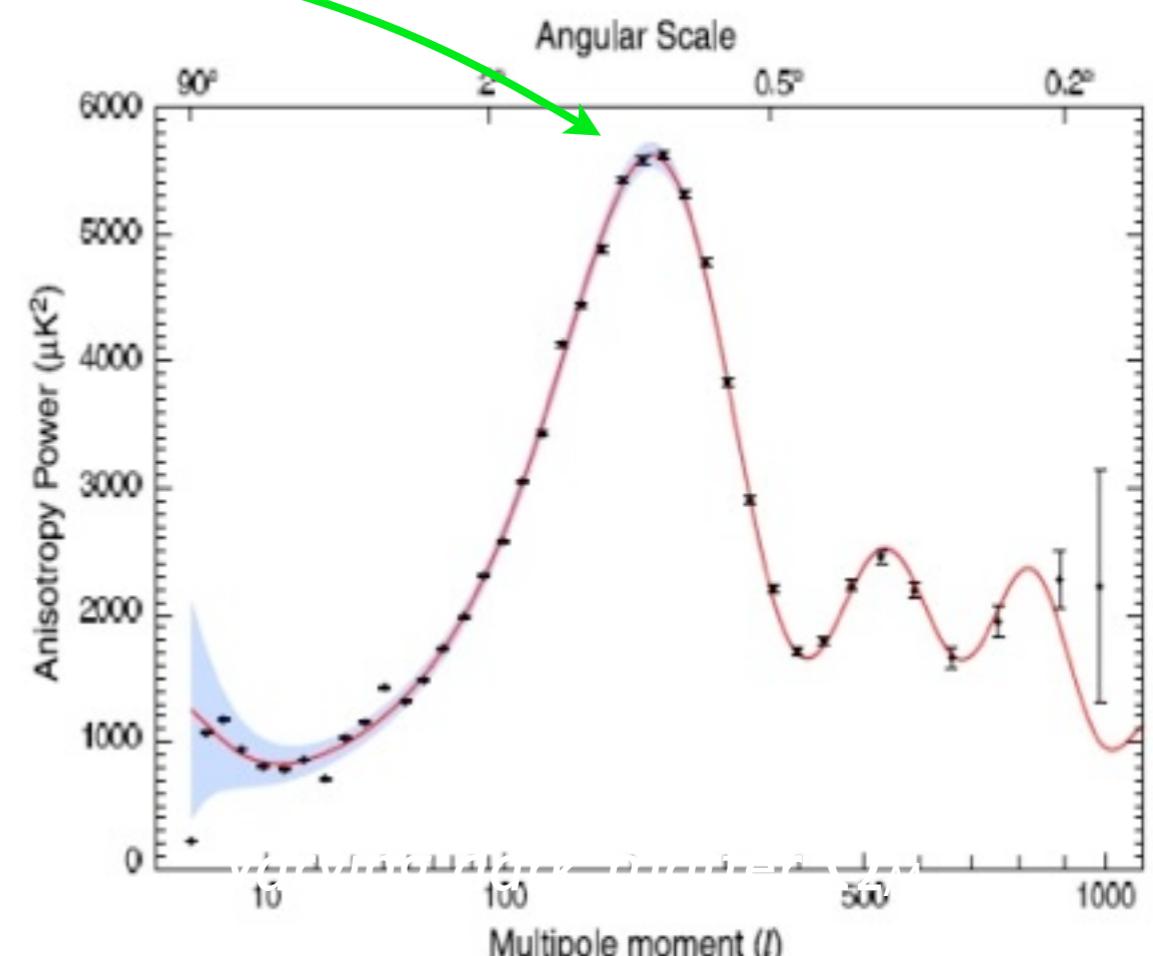
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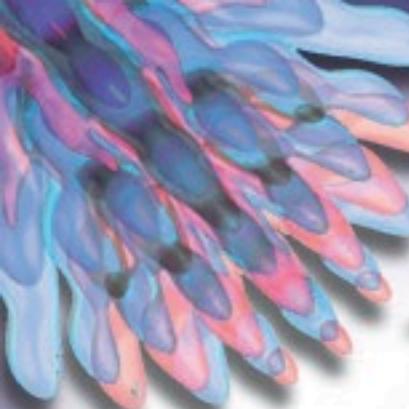


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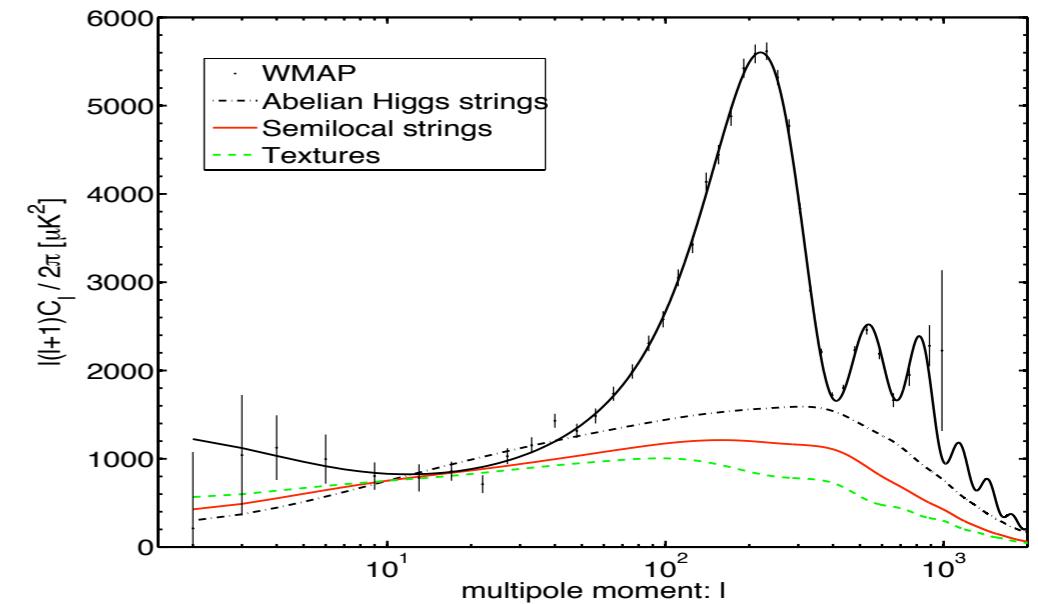
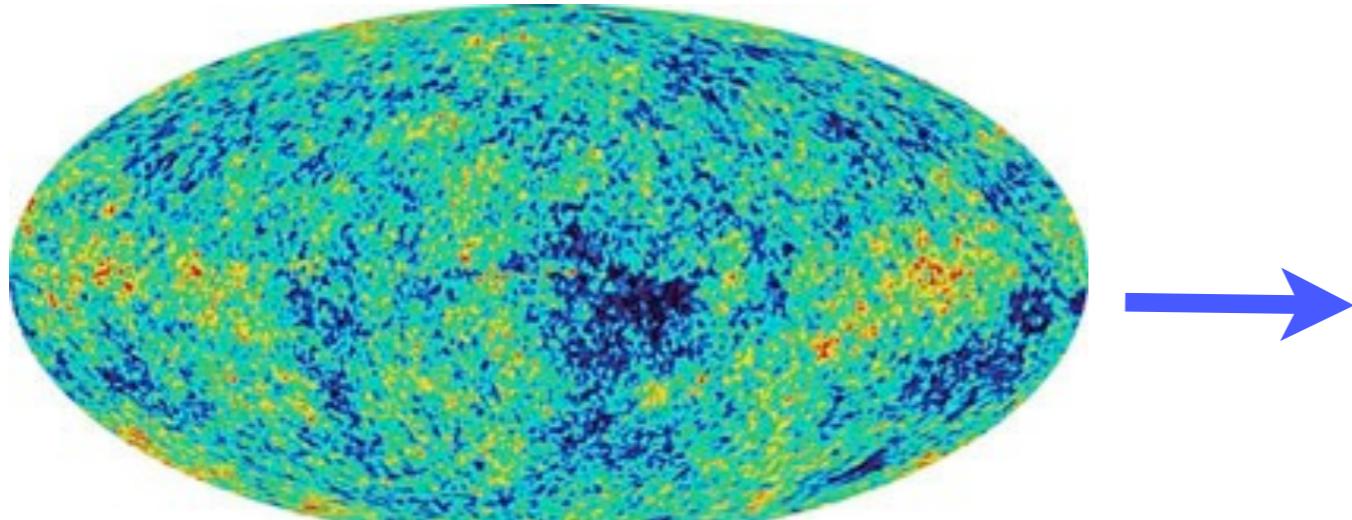
A Gaussian Universe?

The triumph of the inflationary concordance model ...

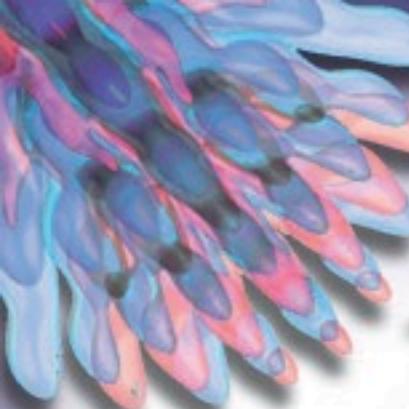
A self-consistent concordance model with (WMAP 7yr results):

4.56(± 0.16)% baryons, 22.7(± 1.4)% dark matter, 72.8(± 1.5)% dark energy with spatial flatness (to 1%) and an age of 13.75(± 0.11) billion years.

All based on the two-point correlator (power spectrum)



i.e. massive compression of data (e.g. WMAP 10^6 pixels to 1000 l's)



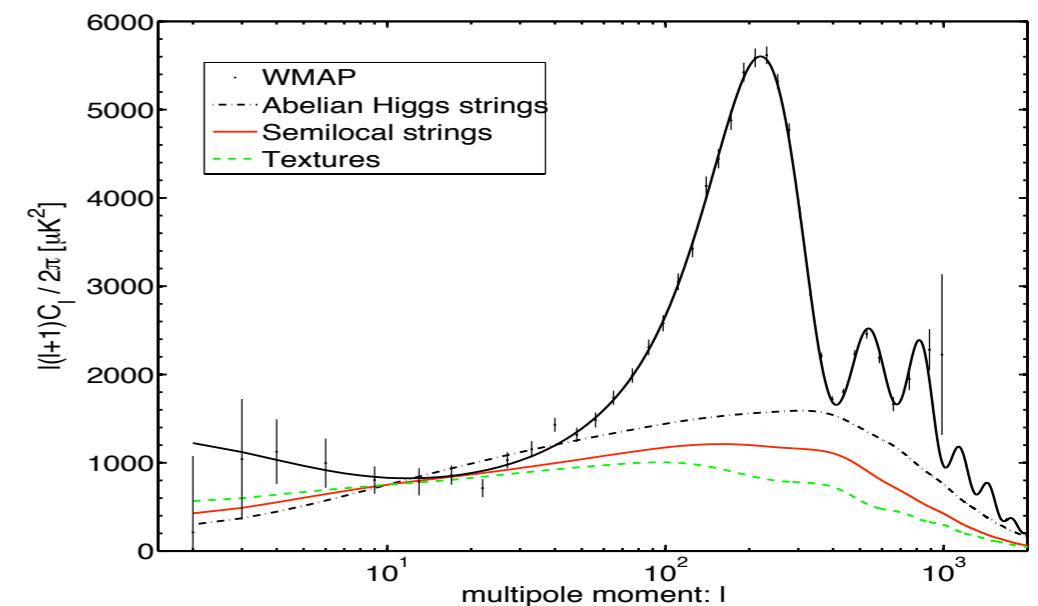
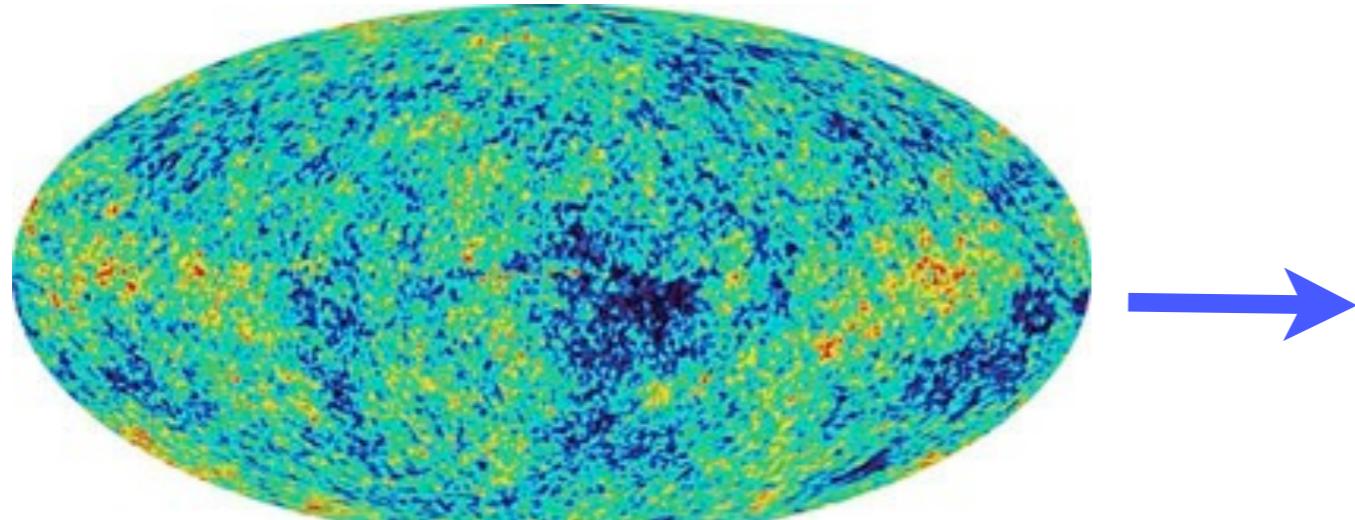
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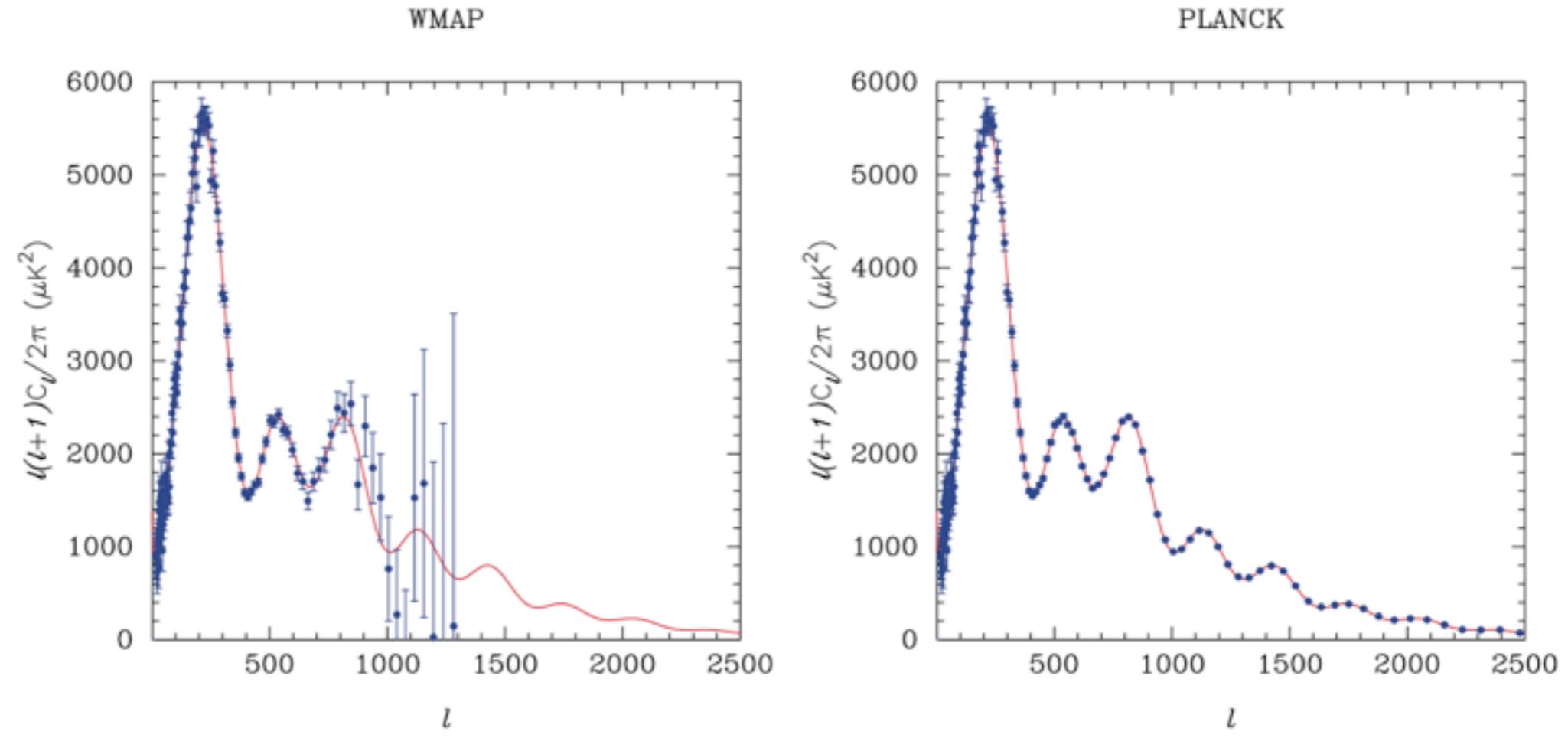


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But is this all there is to be learnt about the early universe?

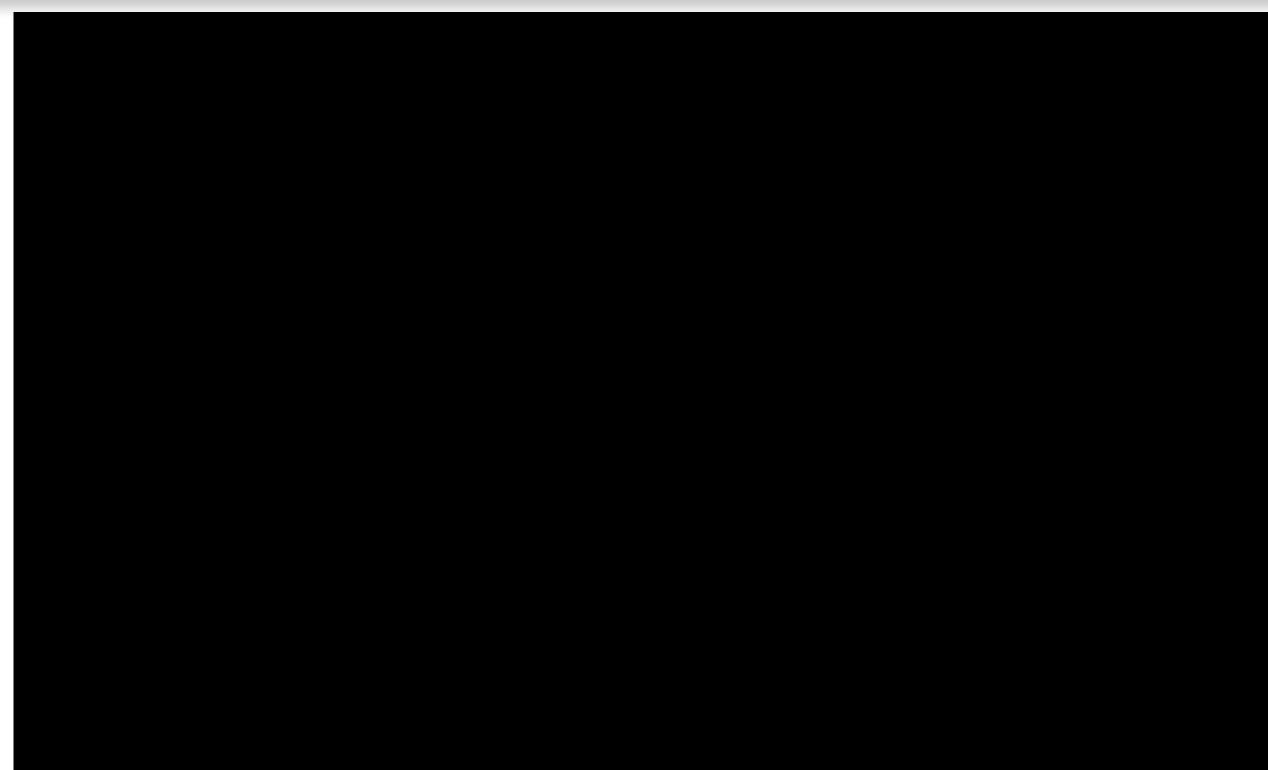
CMB Future - Planck

Launched in May 2009, the ESA Planck satellite, is currently mapping the Universe at high resolution - first cosmology results Jan 2013



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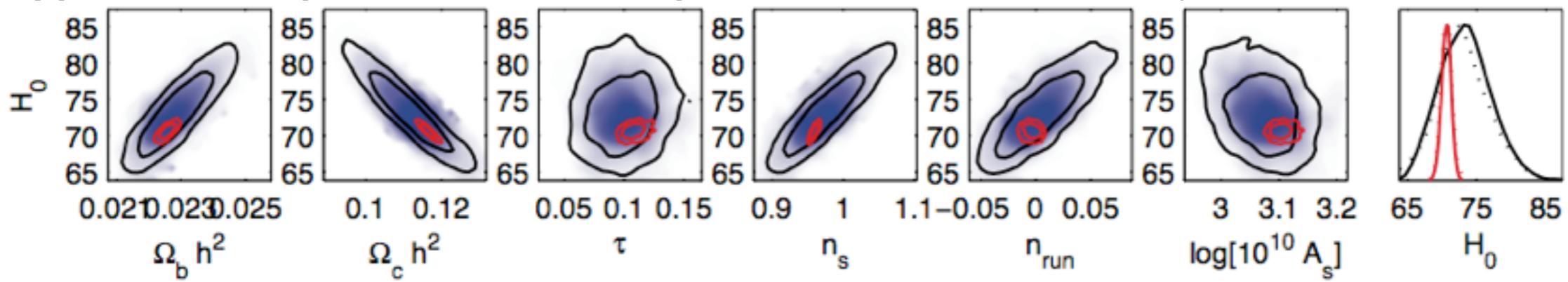
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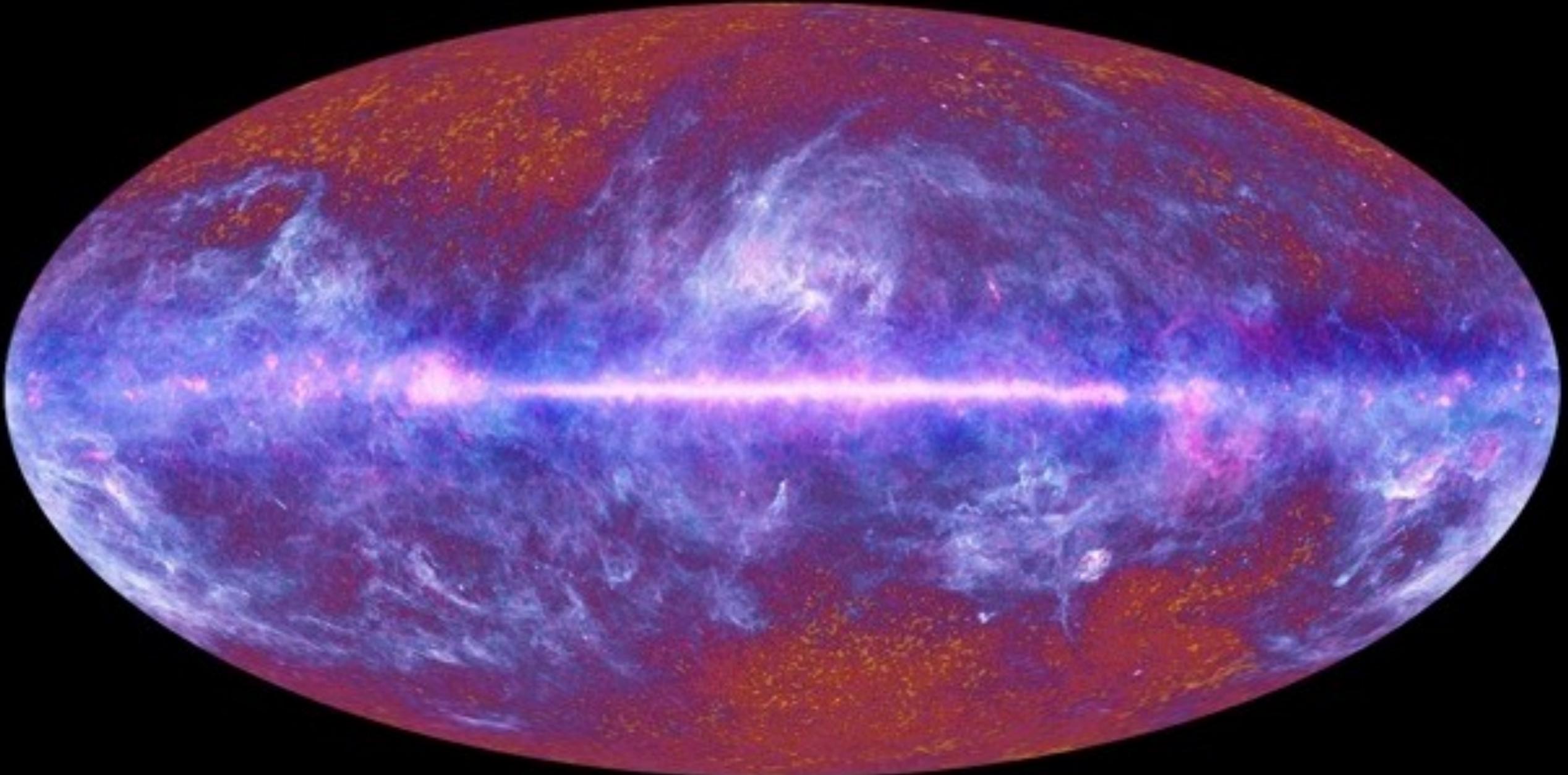
WMAP

PLANCK

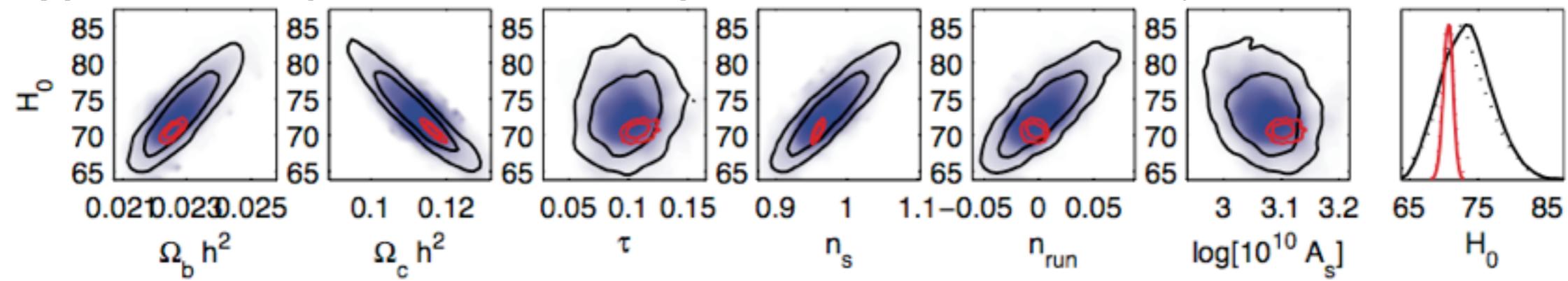
Typical example with Hubble parameter likelihoods (4 x smaller errors)



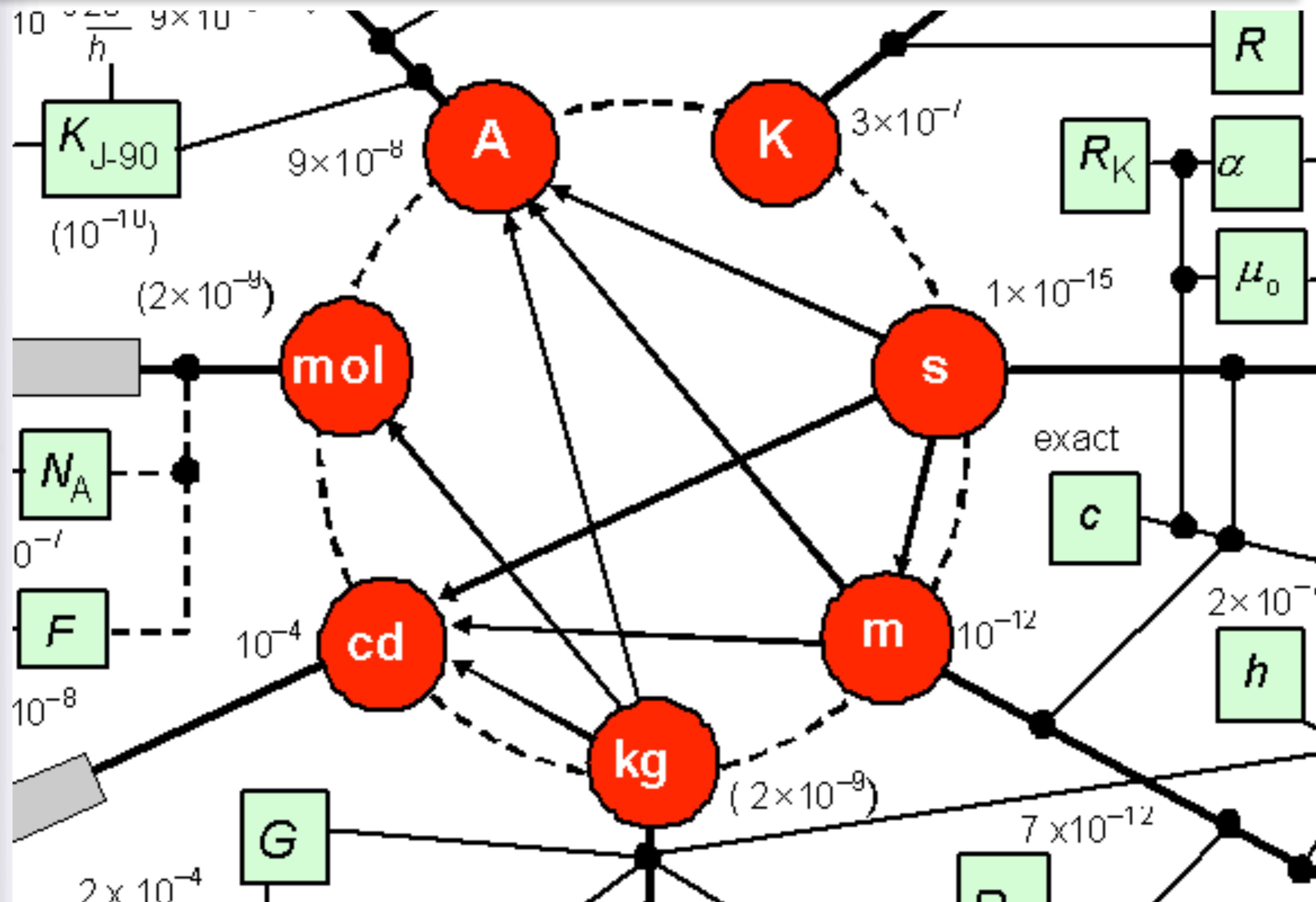
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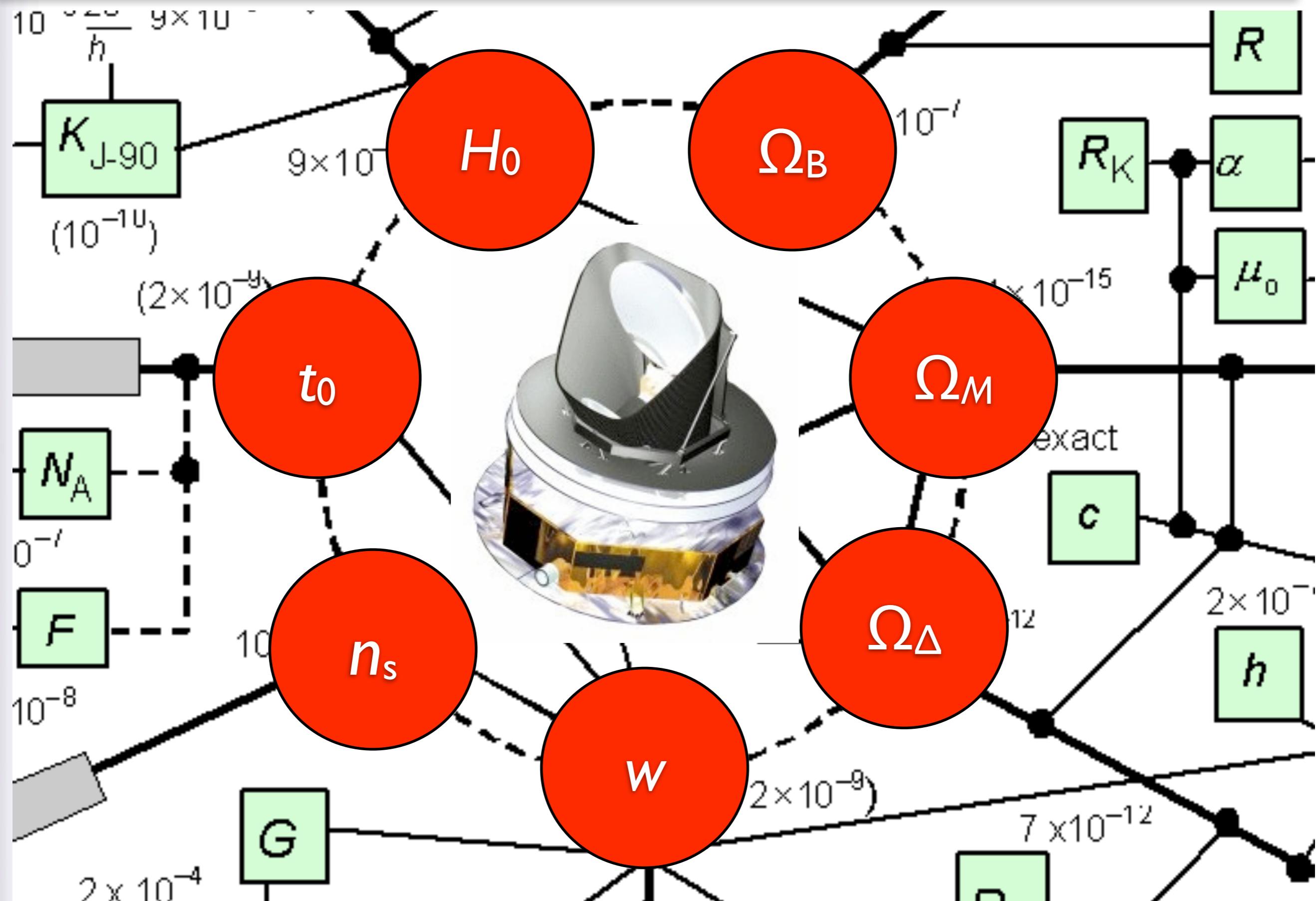
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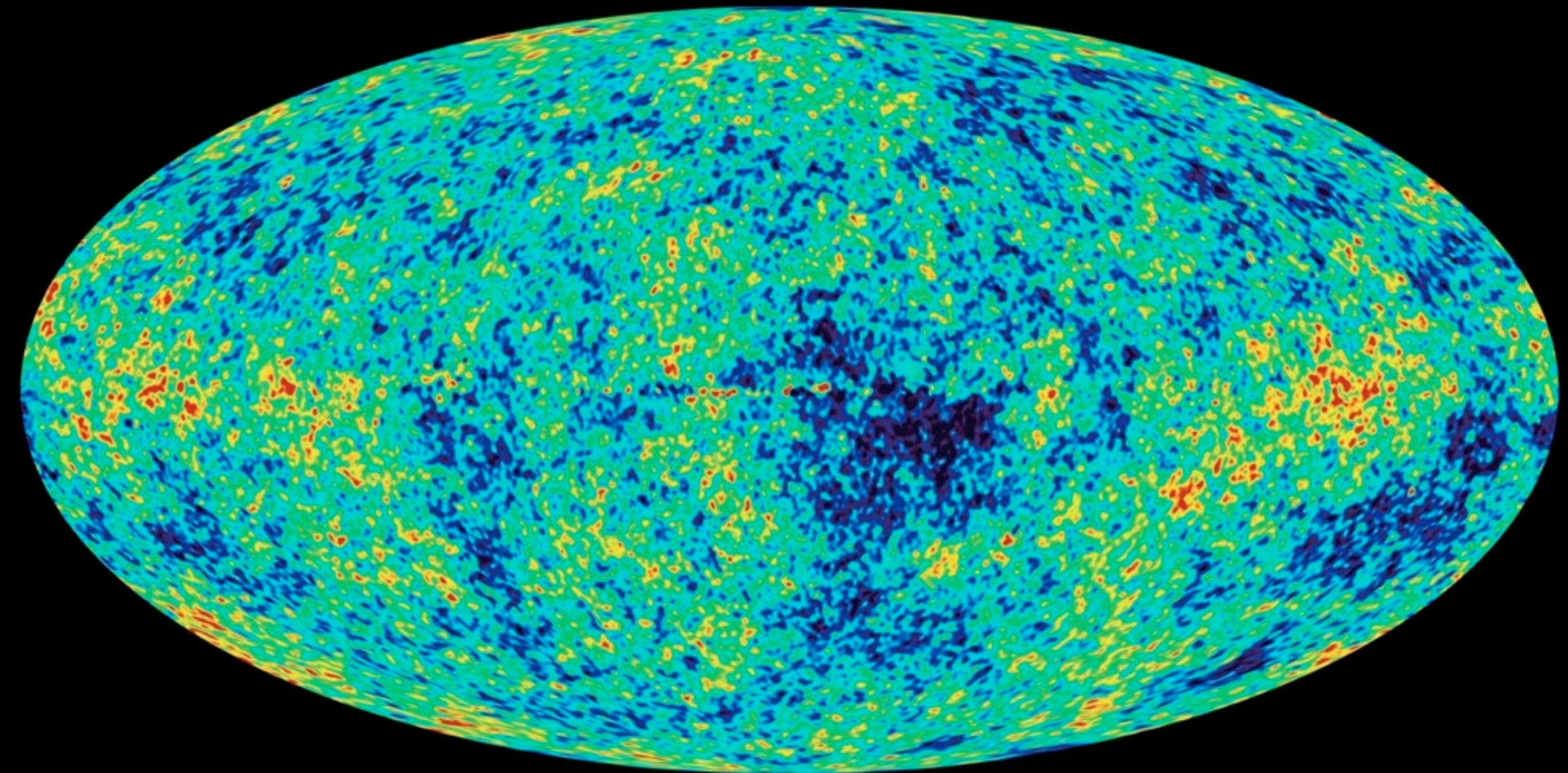
Only a cosmic standards lab?

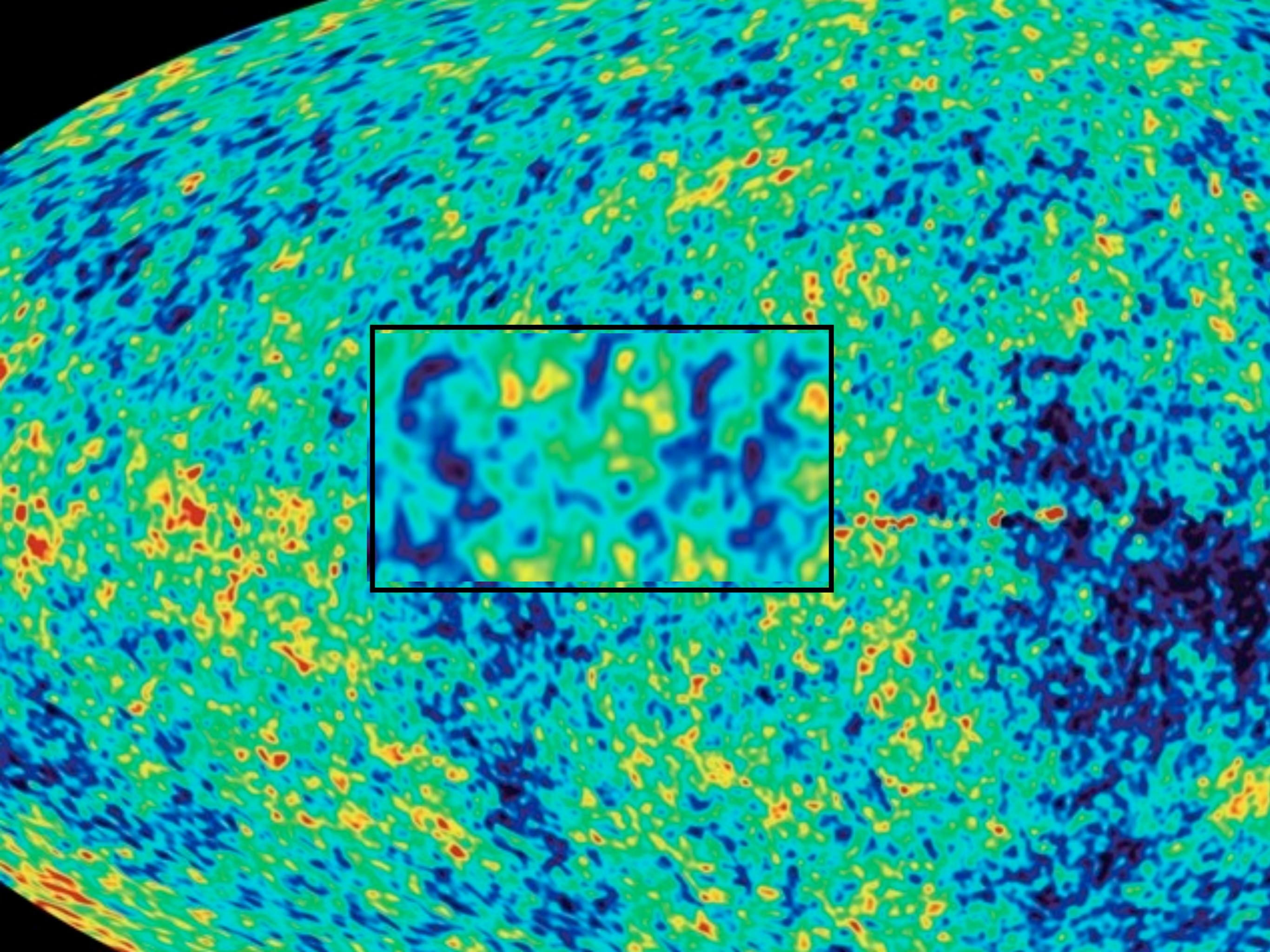


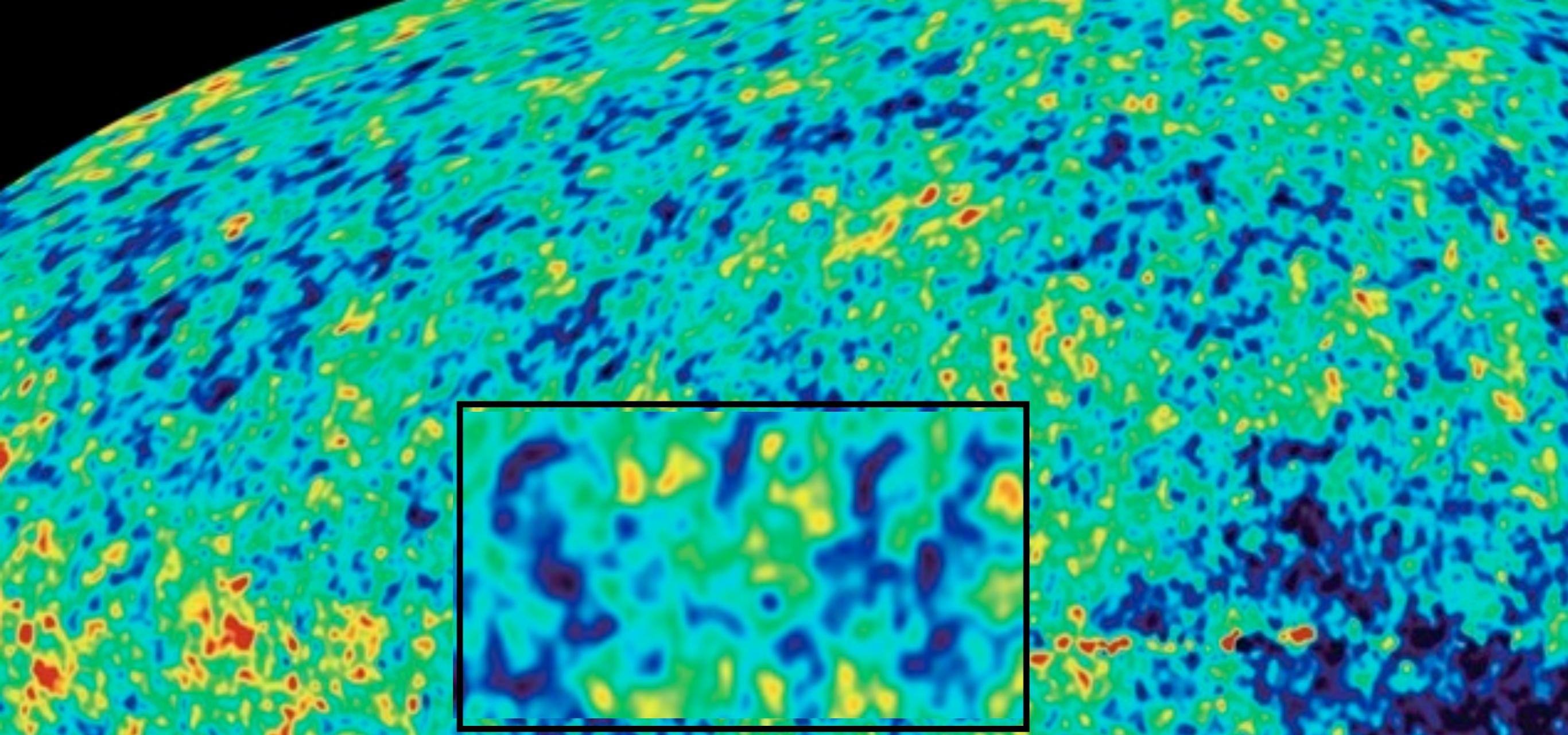
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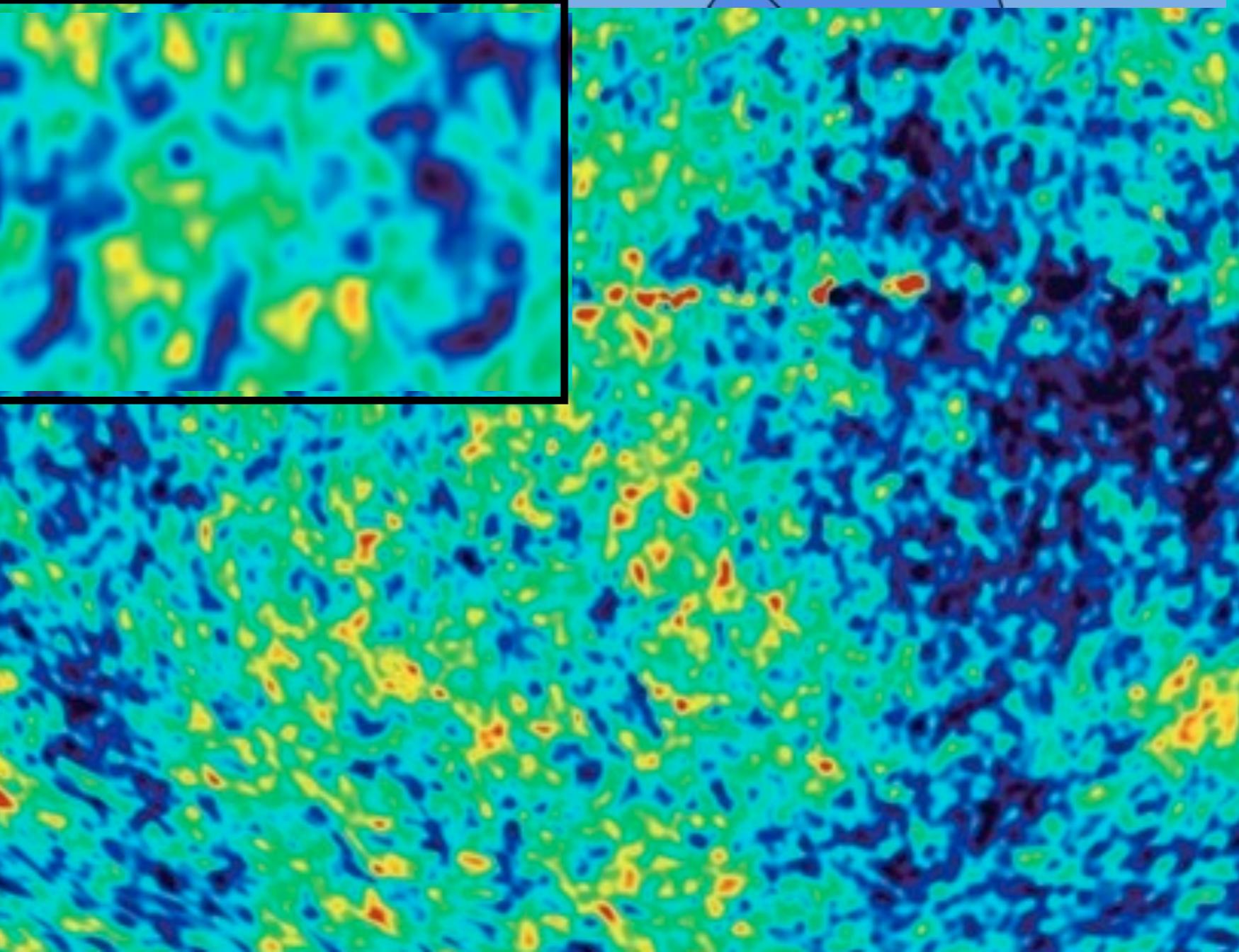
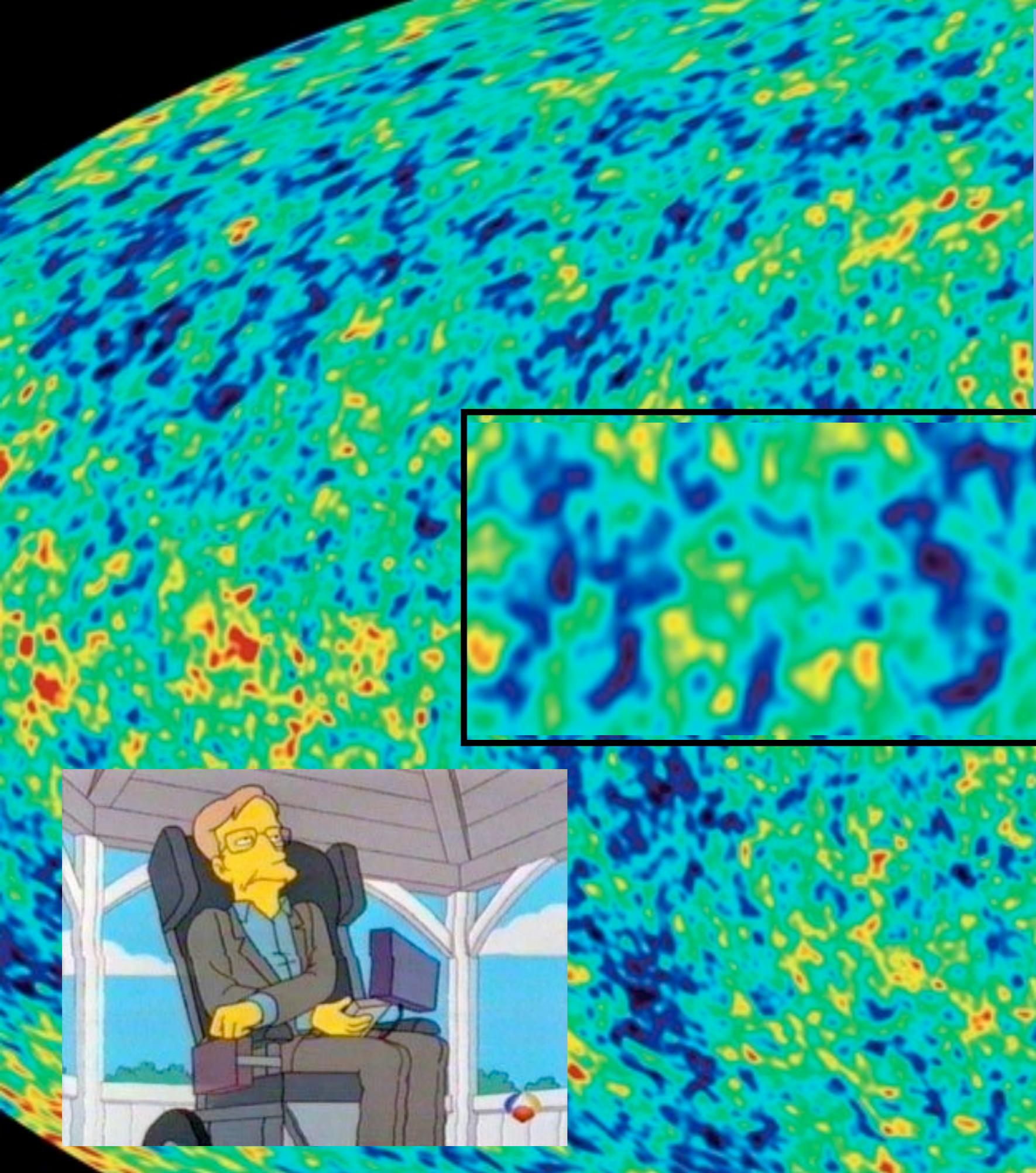


Messages in the WMAP sky?











INFLATION

Inflation is defined as a period of accelerated expansion $\ddot{a} > 0$

The simplest example has equation of state $P = -\rho c^2$

so the energy conservation equation yields $\dot{\rho} = 0 \Rightarrow \rho = \text{const}$

The Friedmann equation gives

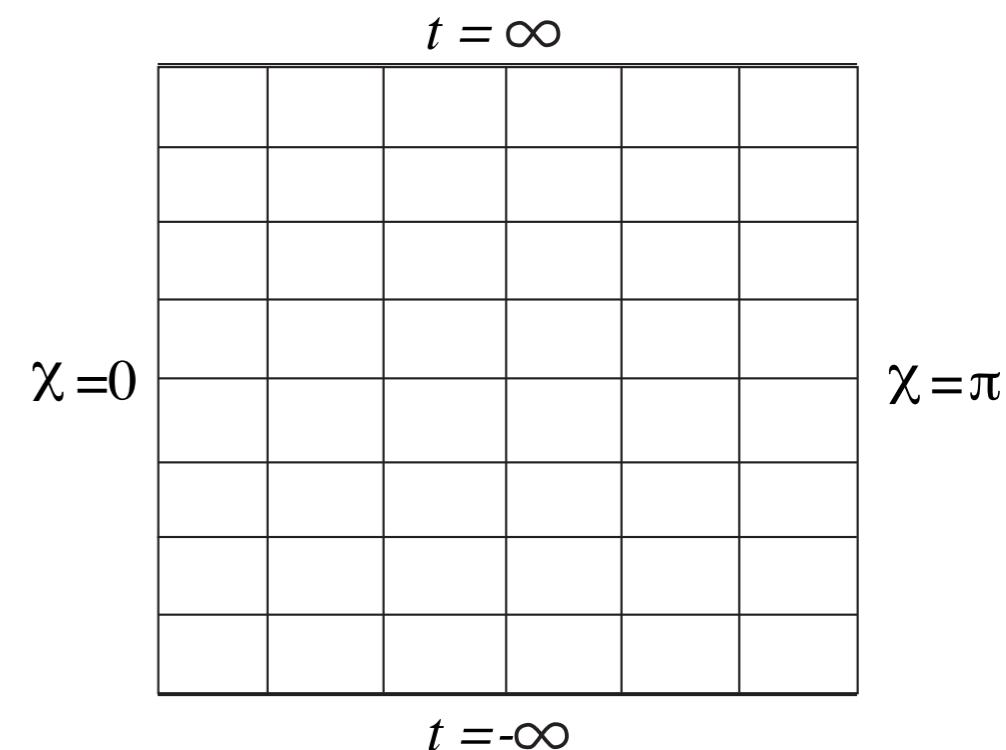
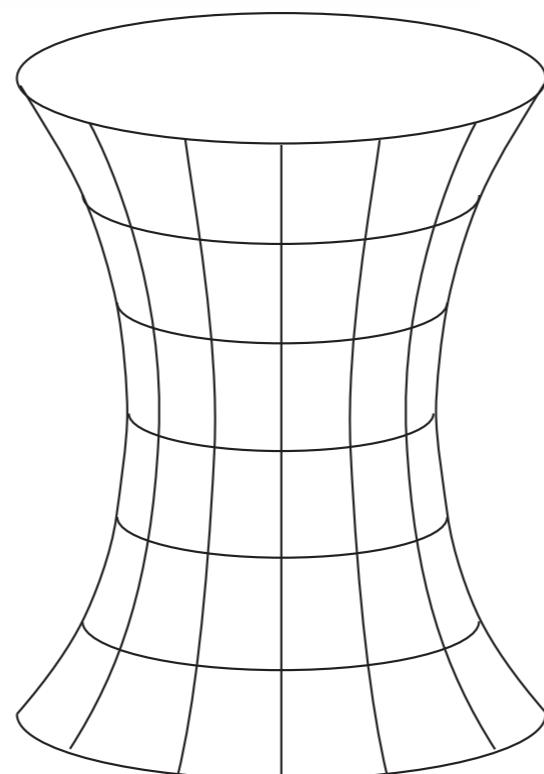
$$\equiv \left(\frac{c^2}{8\pi G} \right) \Lambda$$

$$\left(\frac{\dot{a}}{a} \right)^2 = c^2 \frac{\Lambda}{3} \quad \Rightarrow a(t) = e^{H(t-t_0)}$$

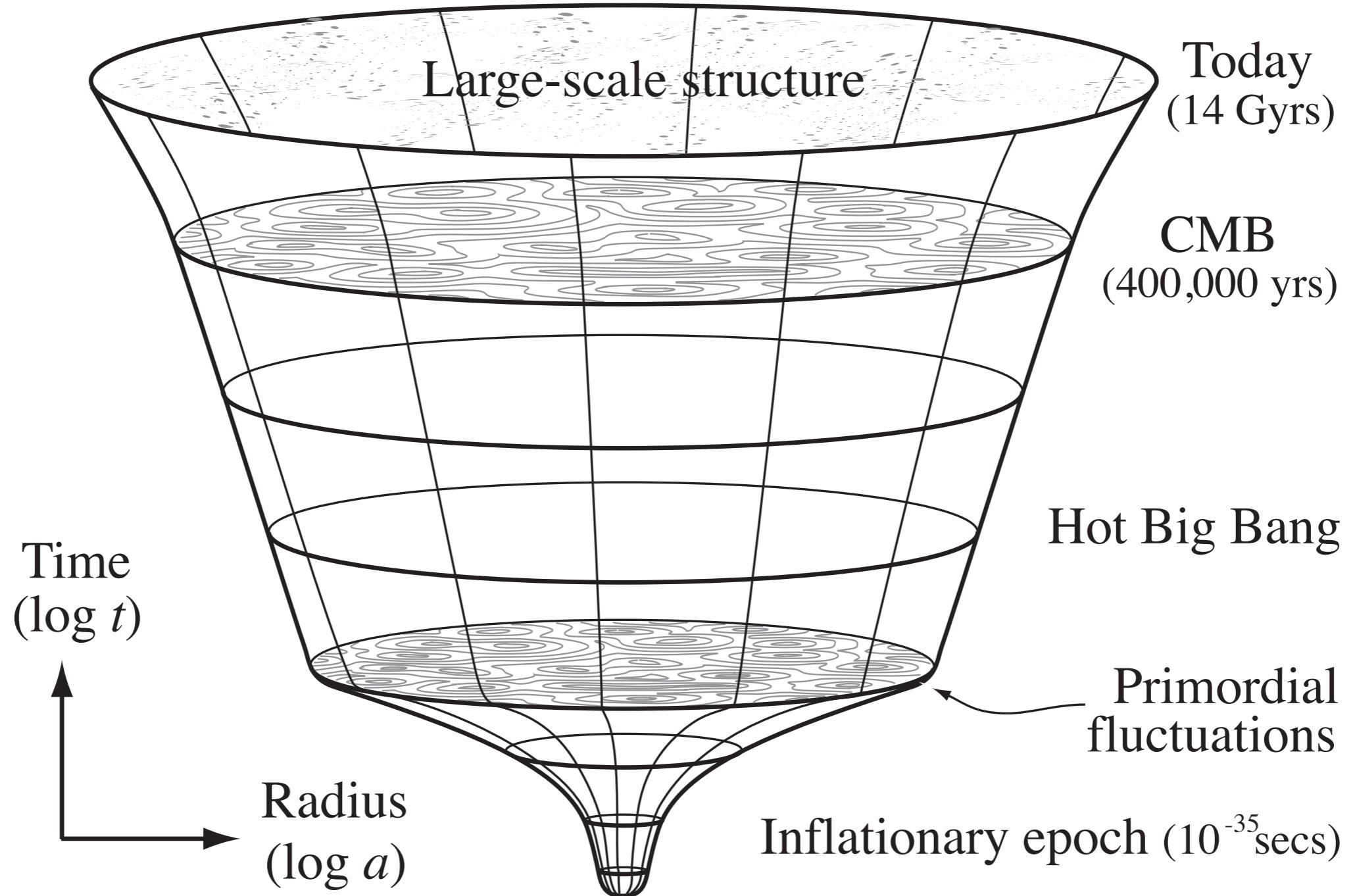
with Hubble parameter

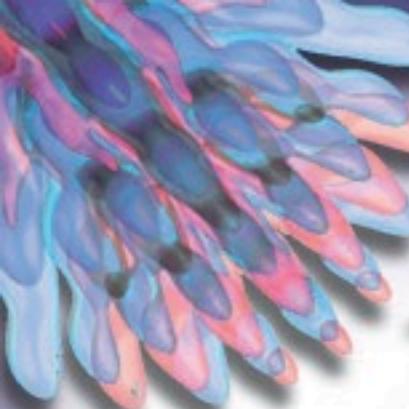
$$H = c \sqrt{\frac{\Lambda}{3}}$$

*De Sitter space
conformal diagram*



Inflationary Hot Big Bang model

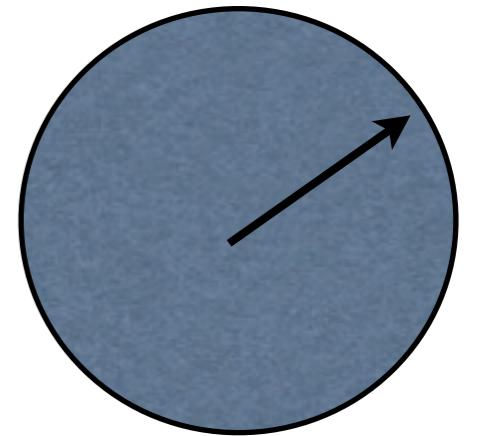




Quantum fluctuations

An accelerating universe has an event horizon

$$d_E(t) = c \int_t^\infty e^{-H_I t'} dt' = c H_I^{-1}$$



Heisenberg's Uncertainty principle with $\Delta x \approx c H_I^{-1}$ implies Δp

$$\Delta E \approx \frac{c}{\hbar} \dot{\phi}^2 \mathcal{V} \approx \frac{c}{\hbar} \left(\frac{\Delta \phi}{\Delta t} \right)^2 (c H_I^{-1})^3 \approx \frac{c^4}{\hbar} (\Delta \phi)^2 H_I^{-1}$$

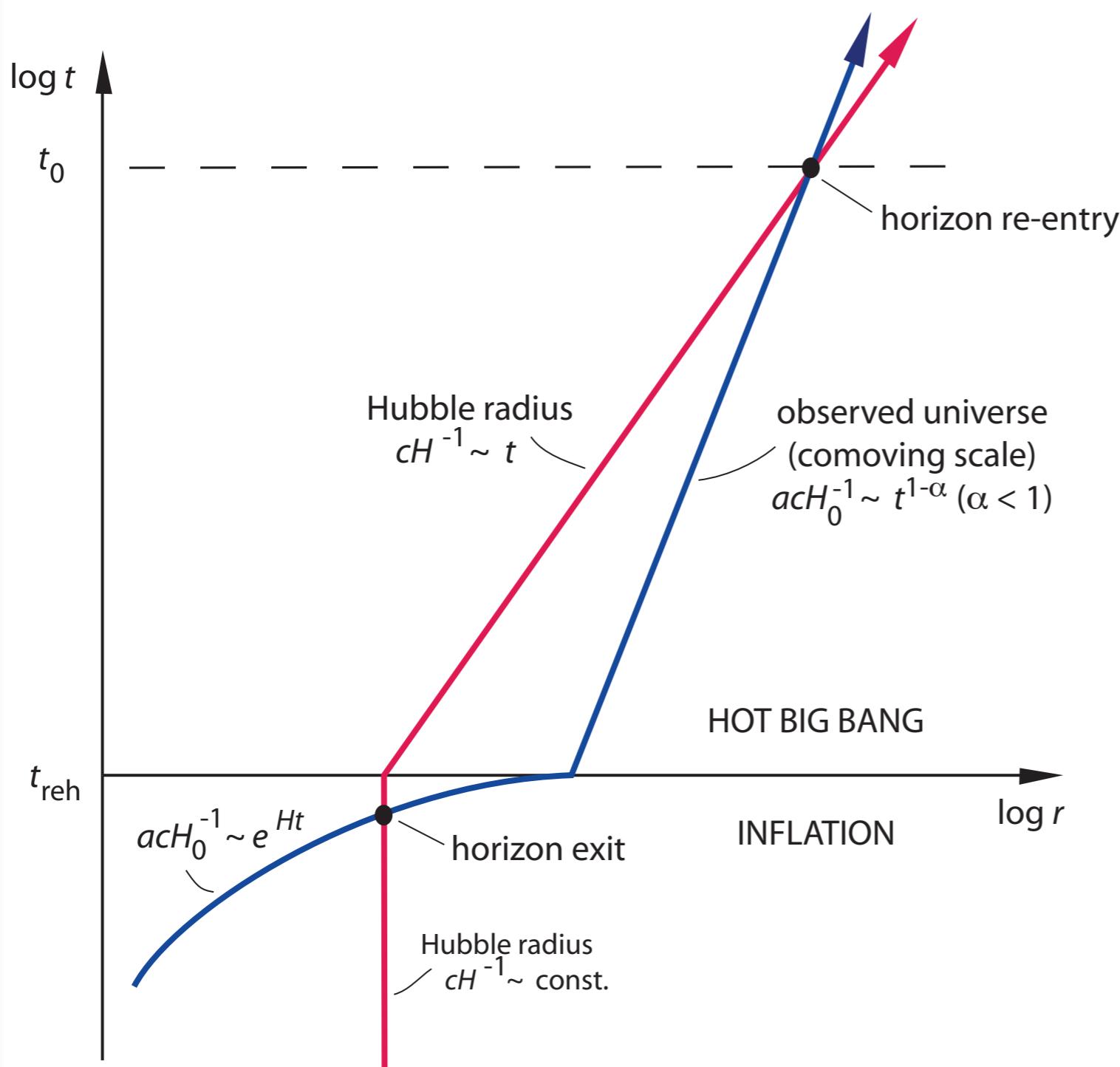
Hence $\Delta E \Delta t \approx (c^4/\hbar)(\Delta \phi)^2 H_I^{-2} \approx \hbar$, implies inflaton

fluctuations

$$(\Delta \phi)^2|_I \approx \left(\frac{\hbar H_I}{c^2} \right)^2$$

These exit the horizon and “freeze” as classical spacetime fluctuations

Perturbation evolution



*The observed universe
(comoving scale evolution):*

*At reheating well outside the
Hubble radius,*

$$\frac{a(t_{\text{reh}})}{a(t_0)} cH_0^{-1} \gg cH^{-1}(t_{\text{reh}})$$

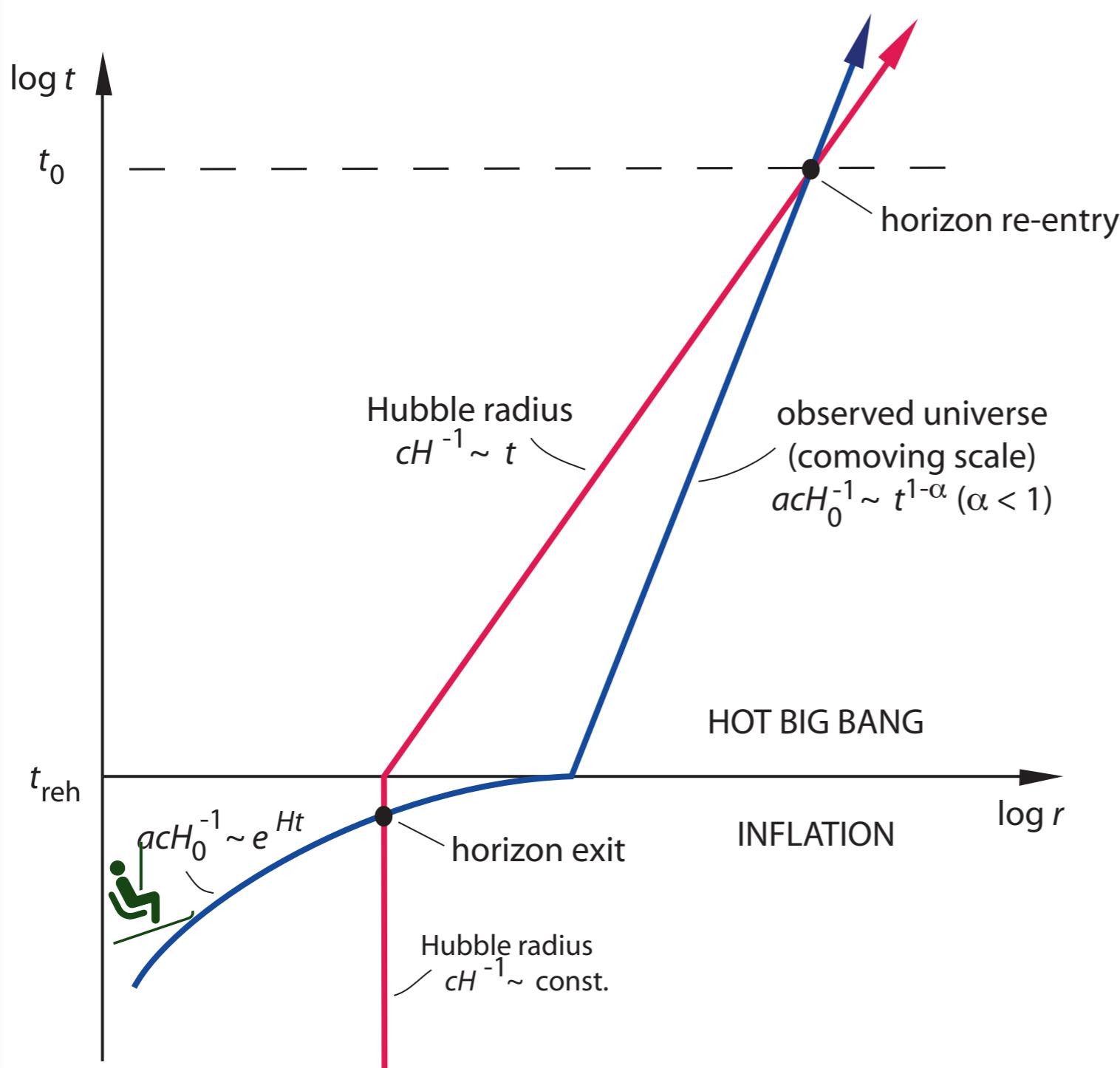
*but matches the Hubble radius
earlier during inflation,*

$$\frac{a(t_I)}{a(t_{\text{reh}})} \frac{a(t_{\text{reh}})}{a(t_0)} cH_0^{-1} = cH_I^{-1}$$

that is, at a time t_I defined by

$$a(t_I)H_I = a(t_0)H_0$$

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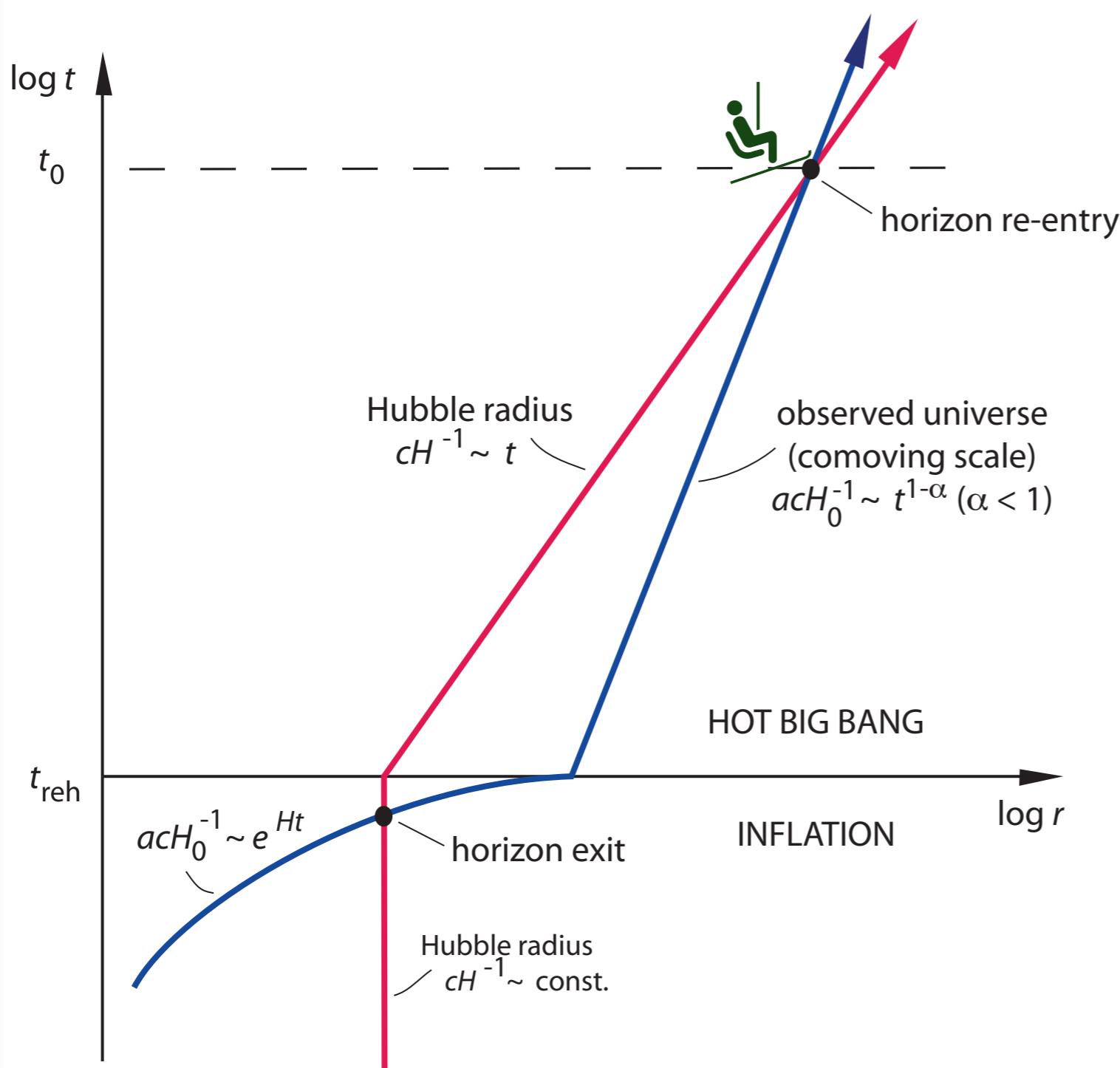
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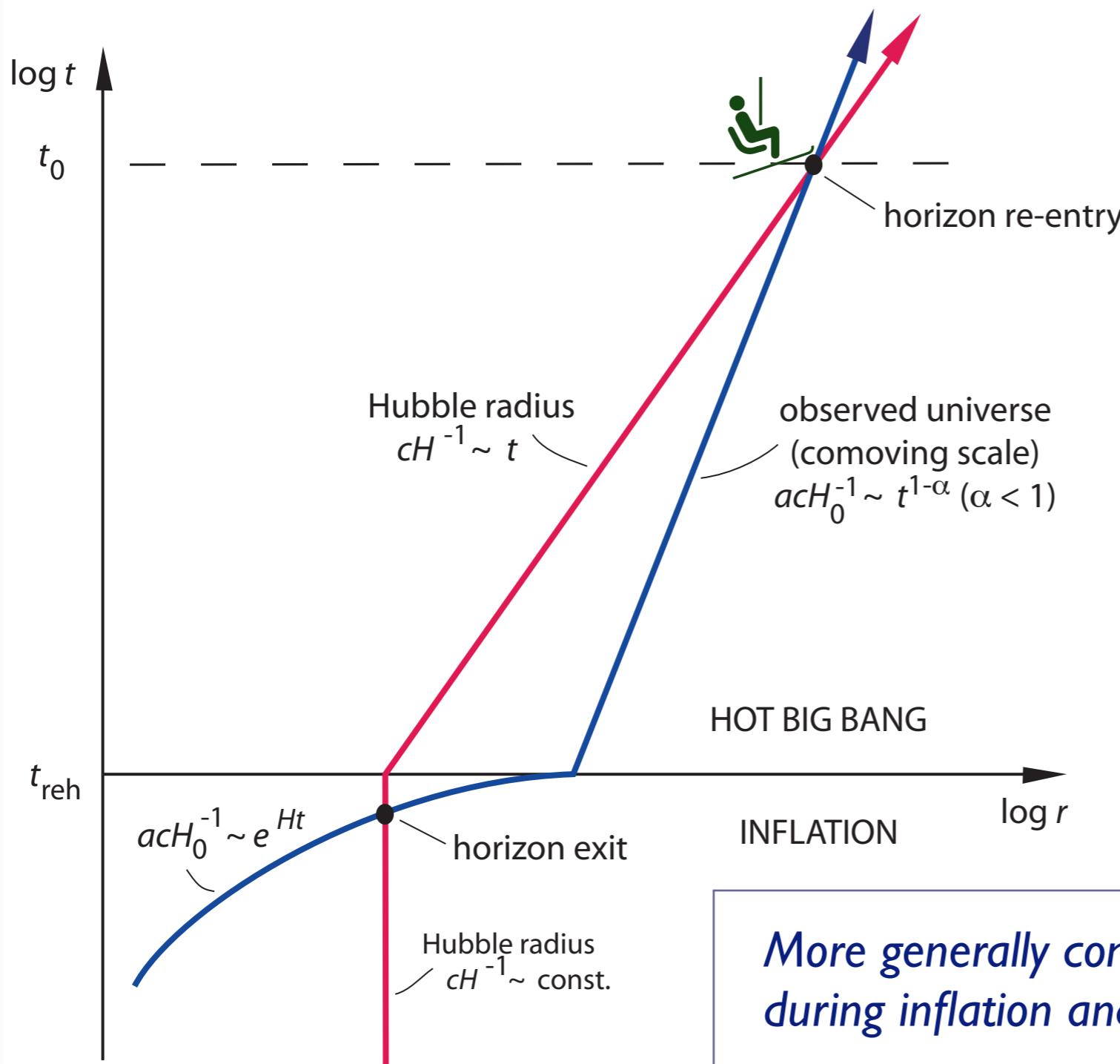
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More generally comoving lengthscale x exits at time t_I during inflation and then re-enters at time t_H :

$$a(t_I)H_I = a(t_H)H(t_H) = cx^{-1}$$

Slow-roll inflation

An extended period of inflation ensues if slow roll conditions apply

$$\dot{\phi}^2 \ll V(\phi) \quad |\ddot{\phi}| \ll |3H\dot{\phi}|, |V'| \quad \leftrightarrow \quad \epsilon, \eta \ll 1$$

or equivalently if the inflaton potential is flat and featureless (approx slow roll params)

$$\epsilon_V = \frac{1}{2} M_{pl}^2 \left(\frac{V'}{V} \right)^2, \quad \eta_V = M_{pl}^2 \frac{V''}{V} \quad \rightarrow \quad \epsilon_V, \eta_V \ll 1$$

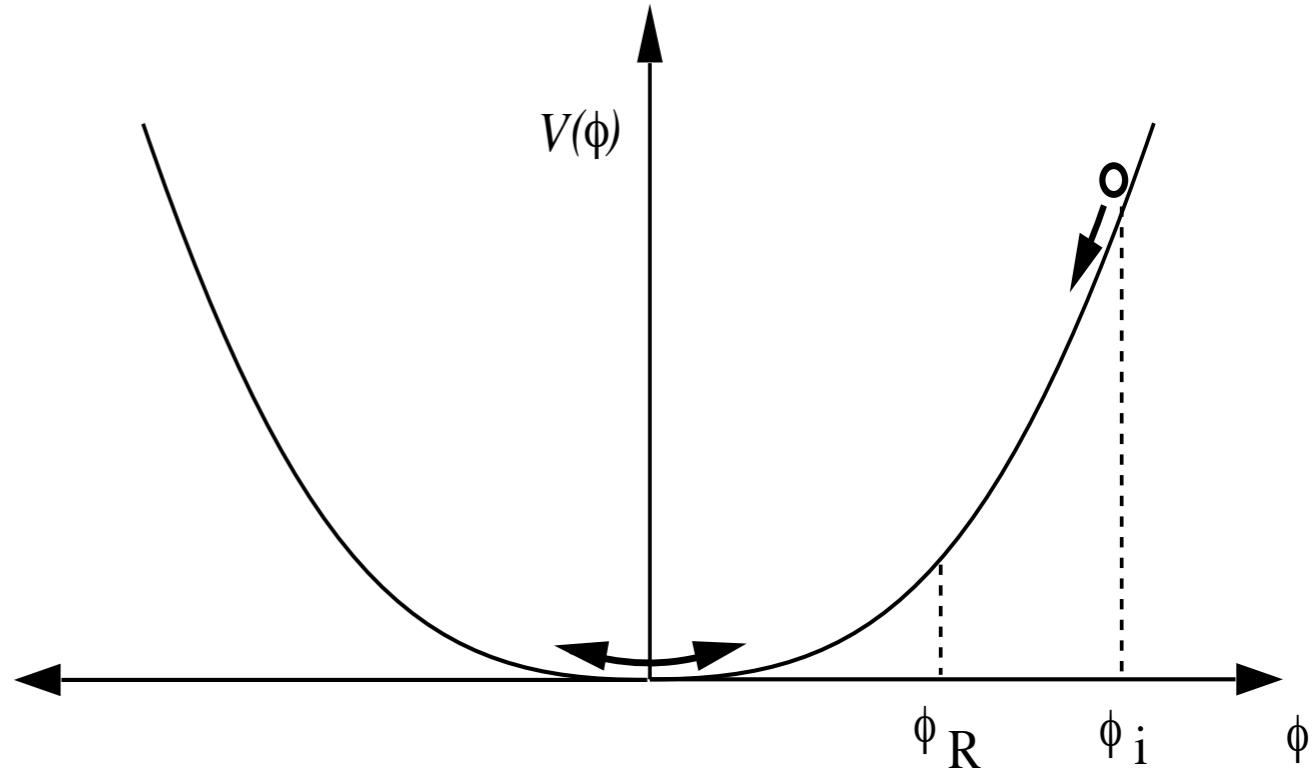
The slow roll equations then admit quasi-exponential expansion

$$H^2 = \frac{1}{3M_{pl}^2} V(\phi), \quad 3H\dot{\phi} = -\frac{dV}{d\phi}$$

Inflation ends and reheating occurs:

$$\frac{1}{2}\dot{\phi}_R^2 \approx V(\phi_R) \quad \text{or} \quad \epsilon(\phi_R) = 1$$

The total number of e-foldings is



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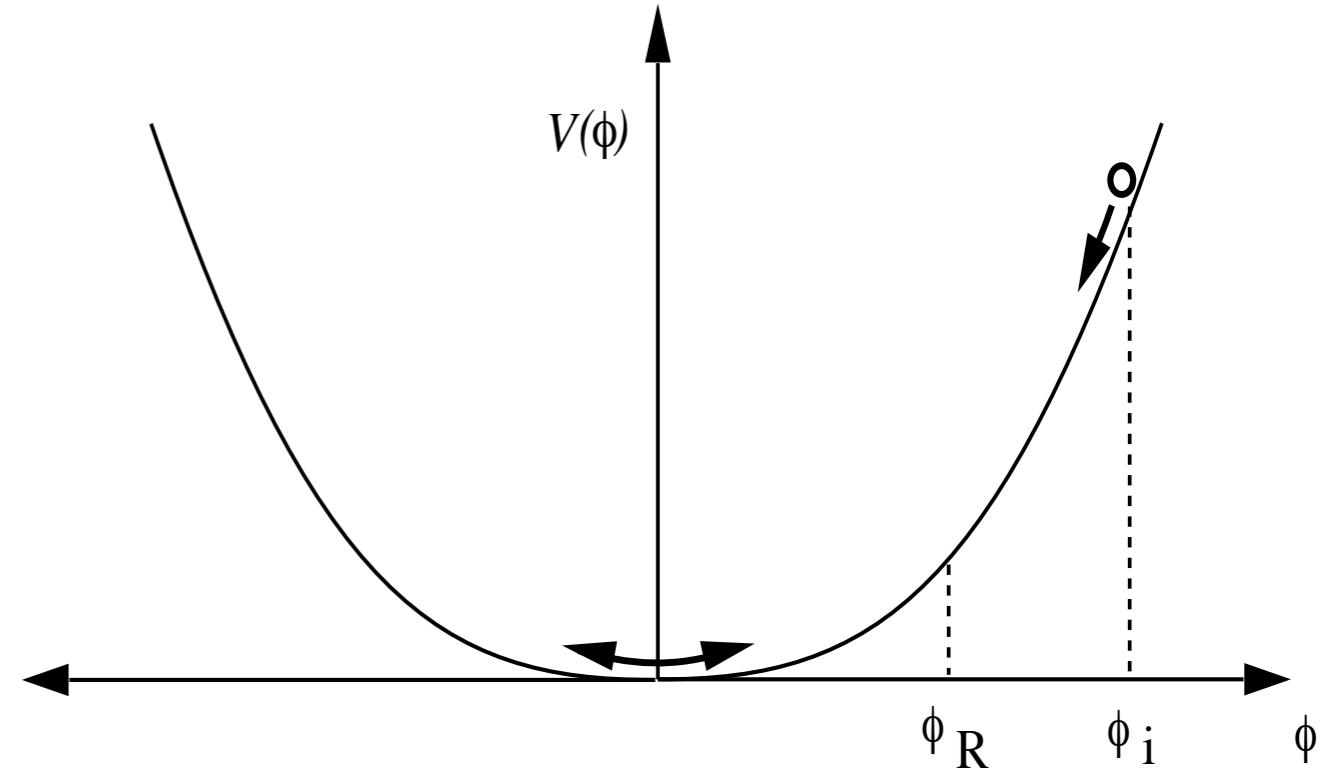
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$$\begin{aligned} \mathcal{N} &= \ln[a(t_i)/a(t_R)] = \int_{t_i}^{t_R} H dt = \int_{\phi_i}^{\phi_R} \frac{H}{\dot{\phi}} d\phi \\ &\approx \frac{1}{M_{pl}^2} \int_{\phi_i}^{\phi_R} \frac{V}{V'} d\phi \end{aligned}$$



Slow roll spectral index

Amplitude of scalar perturbations for slow roll

$$\Delta_\zeta^2(k) = \frac{1}{24\pi^2} \frac{H^2}{M_{pl}^2} \frac{1}{\epsilon} \Big|_{k=aH} \quad \longrightarrow \quad \Delta_\zeta^2(k) \approx \frac{1}{24\pi^2} \frac{V}{M_{pl}^2} \frac{1}{\epsilon_V} \Big|_{k=aH}$$

Spectral index in the slow roll approximation

$$n_s - 1 \equiv \frac{d \ln \Delta_\zeta^2}{d \ln k} \quad \longrightarrow \quad n_s - 1 \approx 2\eta - 4\epsilon \approx 2\eta_V - 6\epsilon_V$$

so for simple free field potential

$$\epsilon_V \approx \eta_V \approx \frac{1}{\mathcal{N}} \text{ for } V(\phi) = \frac{1}{2} m^2 \phi^2 \quad \longrightarrow \quad n_s \approx 1 - \frac{2}{\mathcal{N}} \approx 0.96$$

and more generally

$$V(\phi) = \Lambda^4 \left(\frac{\phi}{\mu} \right)^p \quad \longrightarrow \quad n_s \approx 0.98 - \frac{p}{100}$$

Aside: Hamilton-Jacobi slow-roll params

$$\epsilon = -\frac{\dot{H}}{H^2} = \frac{1}{2M_{pl}^2} \frac{\dot{\phi}^2}{H^2} \quad \eta = -\frac{\ddot{\phi}}{H\dot{\phi}}$$

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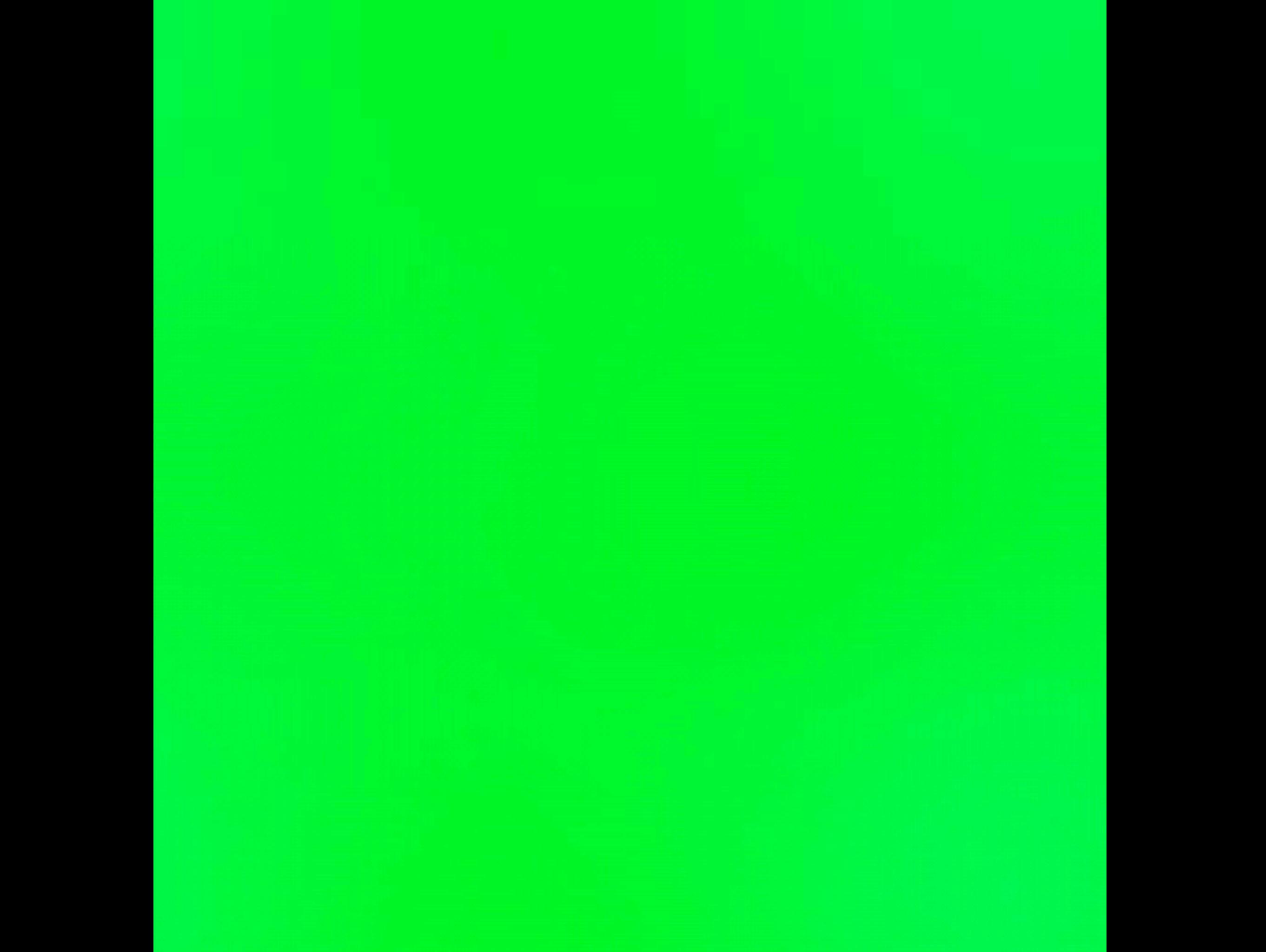
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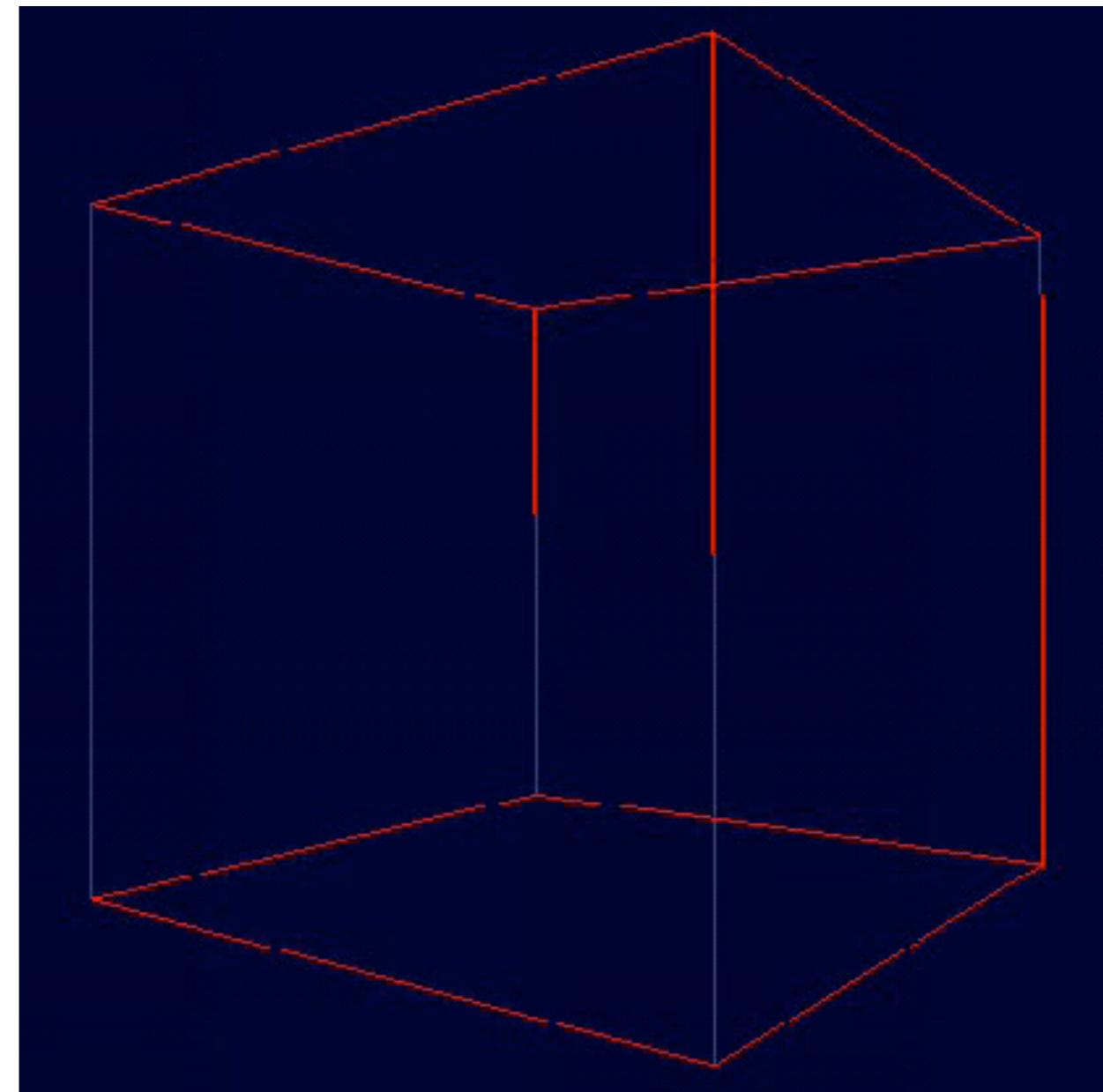
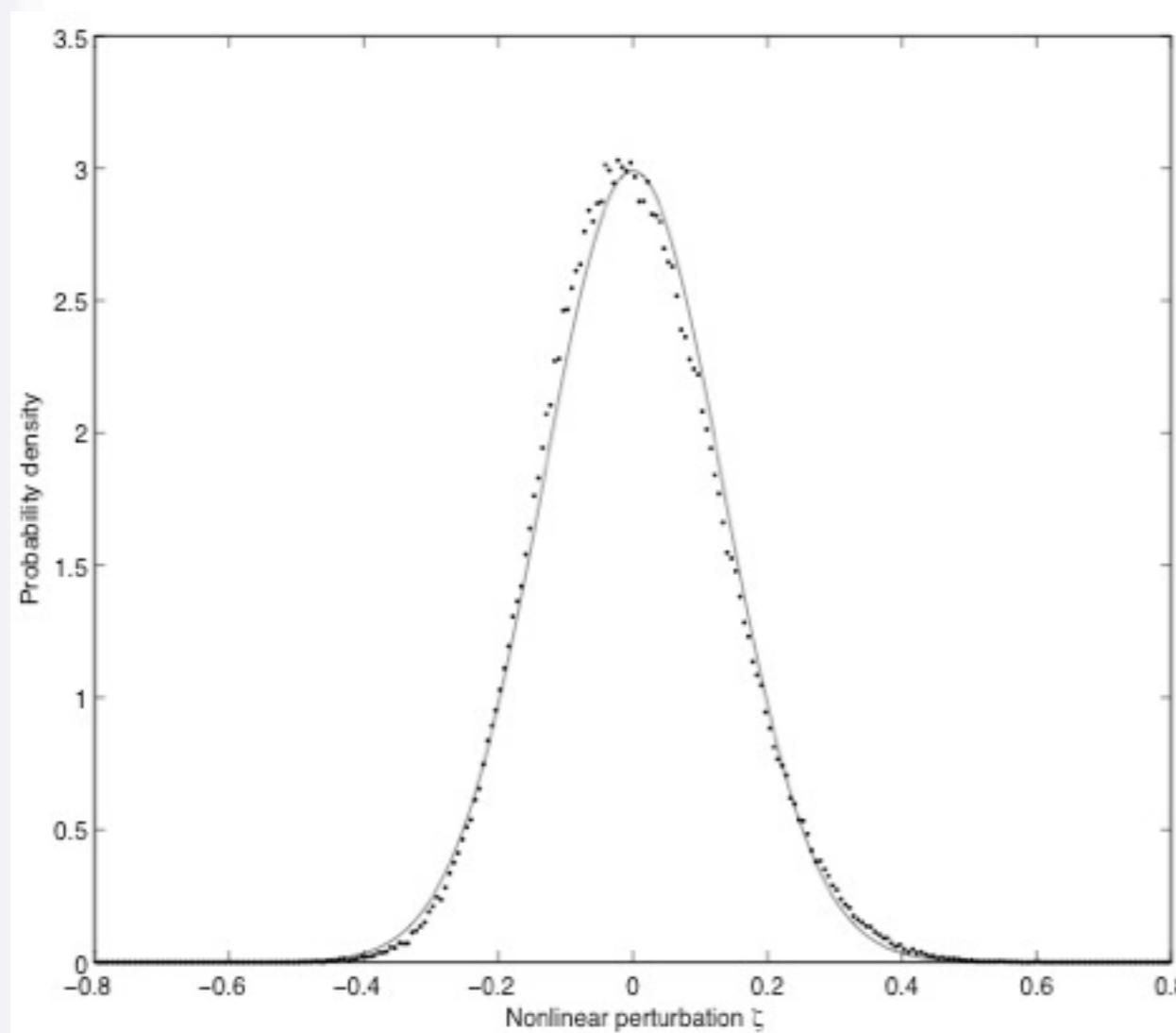
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$$\eta = -\frac{\ddot{\phi}}{H\dot{\phi}}$$



Probability density function

Numerical results ...



Small NG signal for single field inflation

Significant f_{NL} for multifield inflation (see later)

Ongoing stochastic simulations/theory (Funakoshi & EPS, 2011)



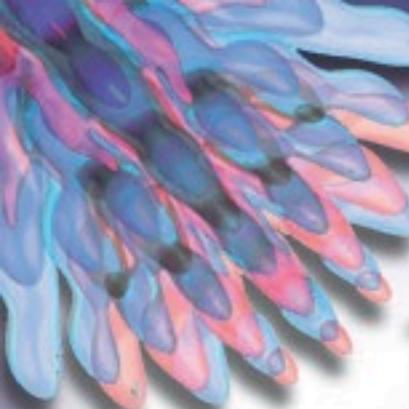
Randomness primer

Tossing a coin:

Heads

Tails





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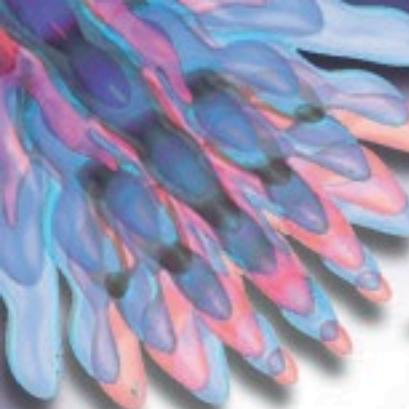
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Pascal's triangle

N = 0							1
N = 1					1		1
N = 2				1	2		1
N = 3			1	3	3		1
N = 4		1	4	6	4		1
N = 5	1	5	10	10	5		1
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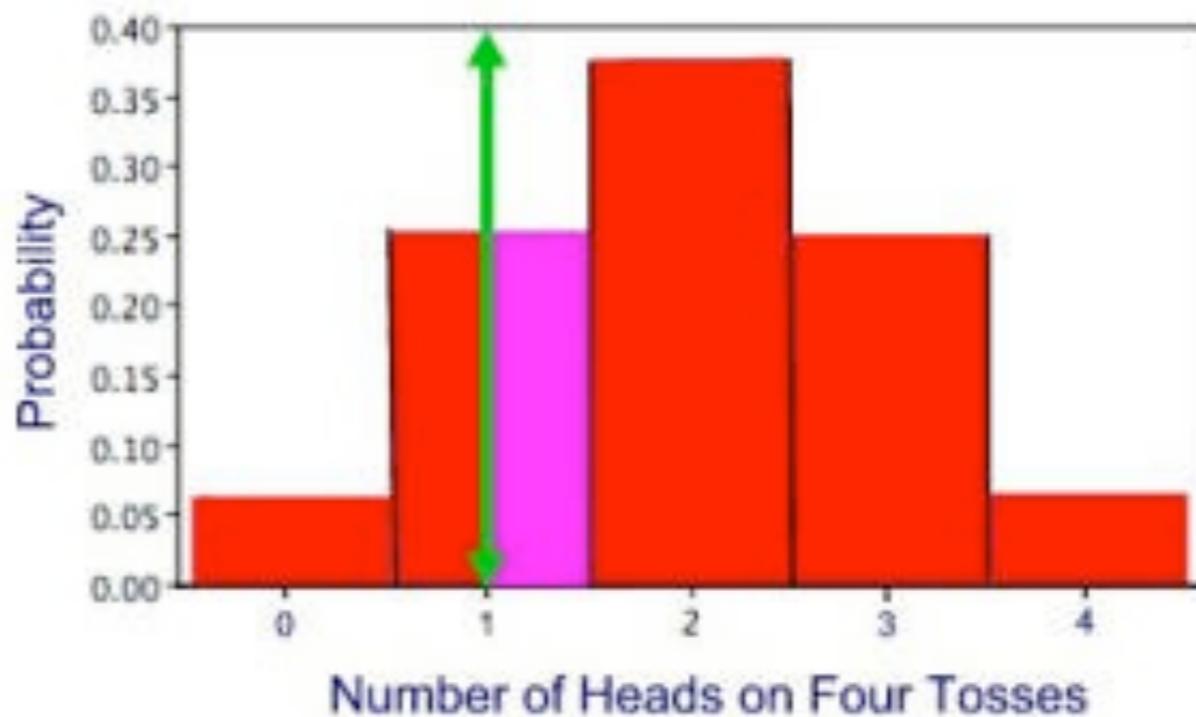
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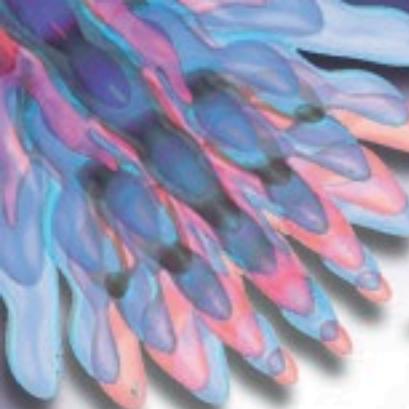
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Plot of binomial distn for different no. of tosses:





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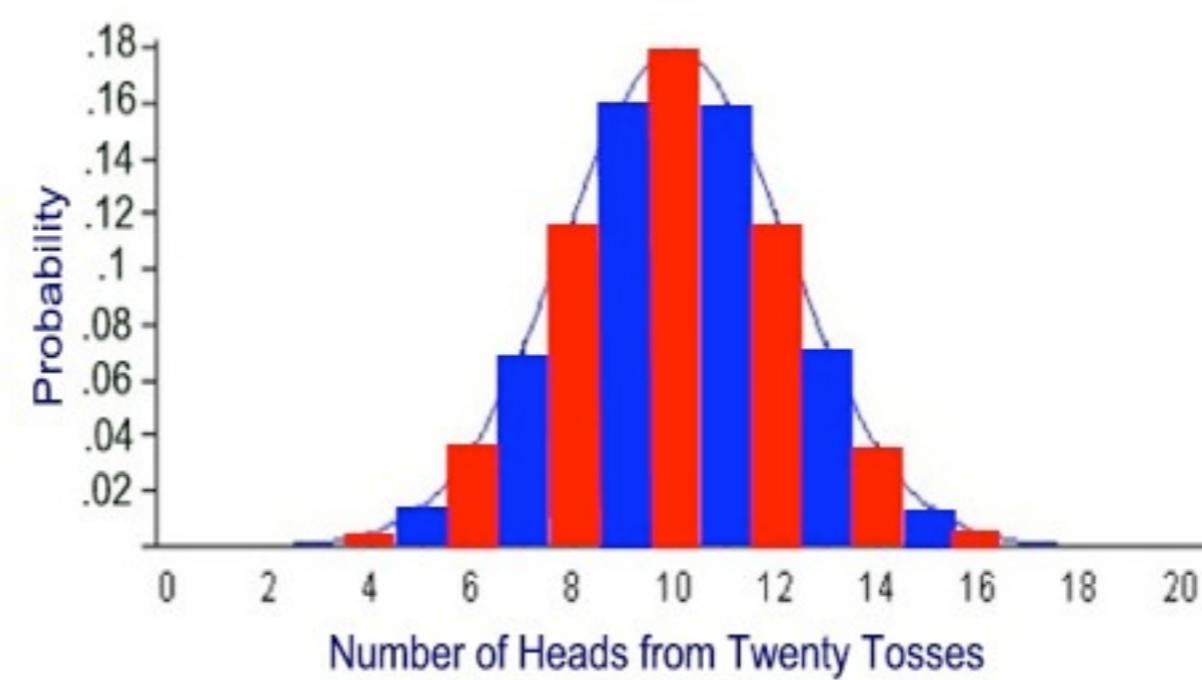
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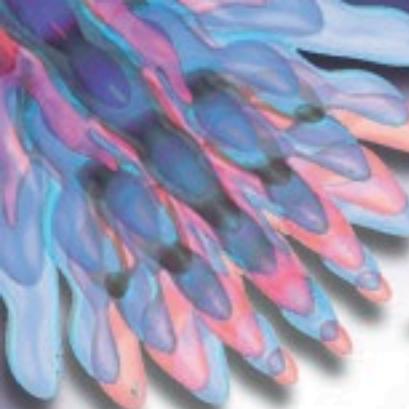


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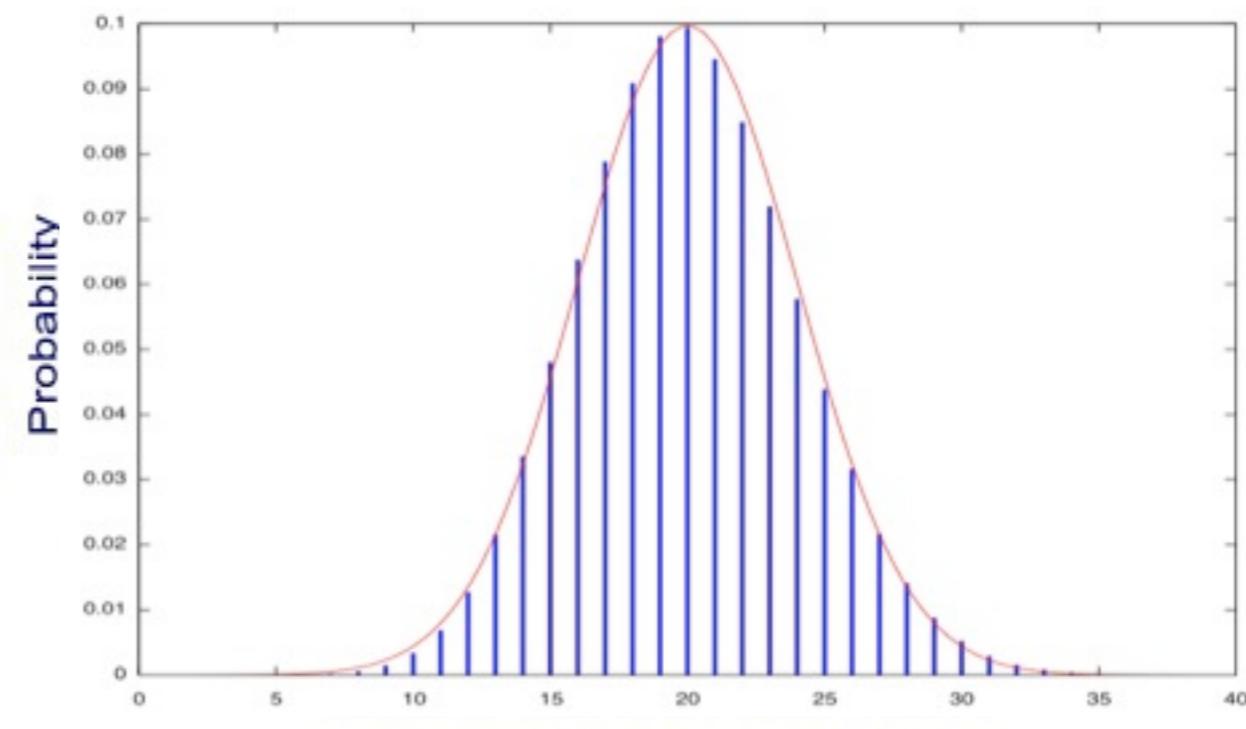
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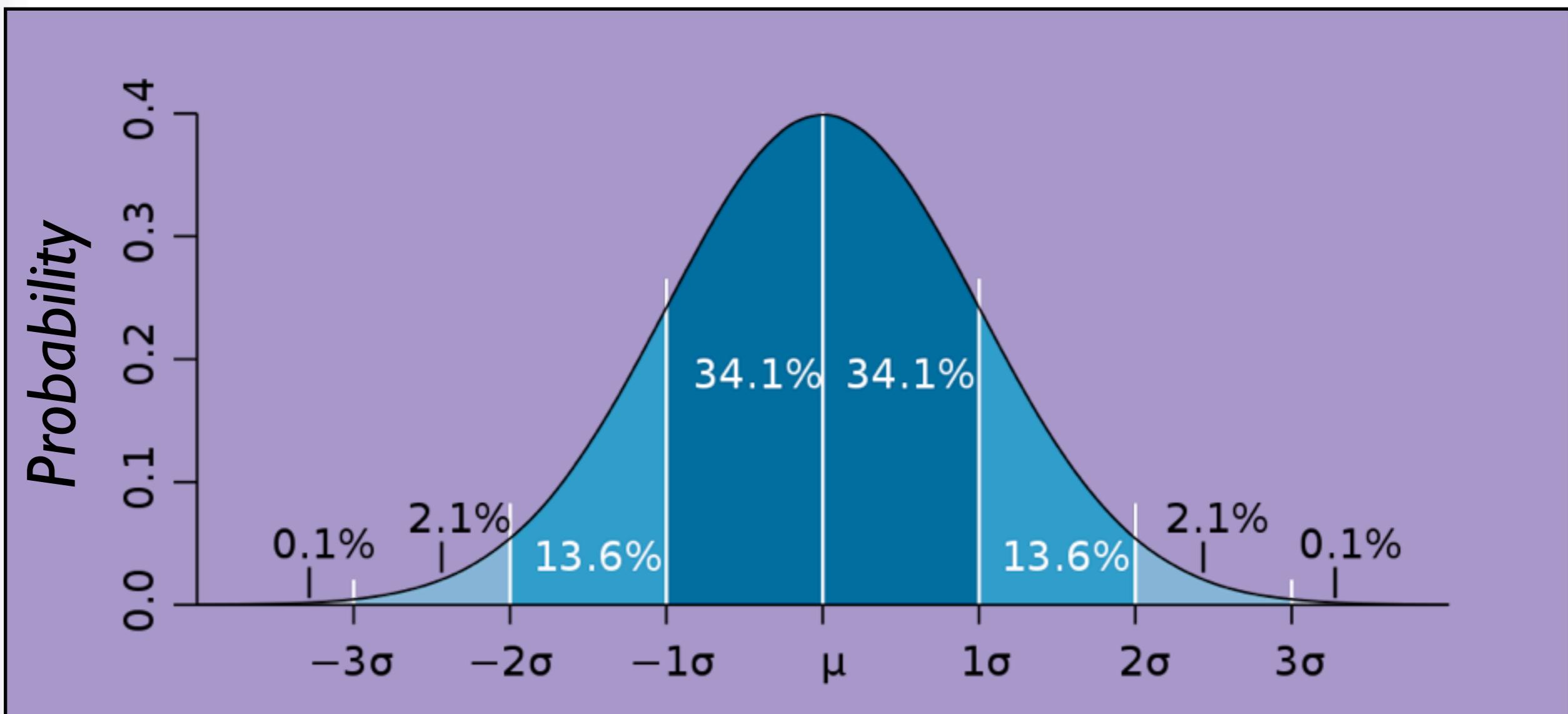
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Plot of binomial distn for different no. of tosses:

Approaches “bell curve” or normal distribution (when rescaled)



Normal or Gaussian distribution



Determined only by mean μ and standard deviation

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2}$$

Central limit theorem: any independent random process

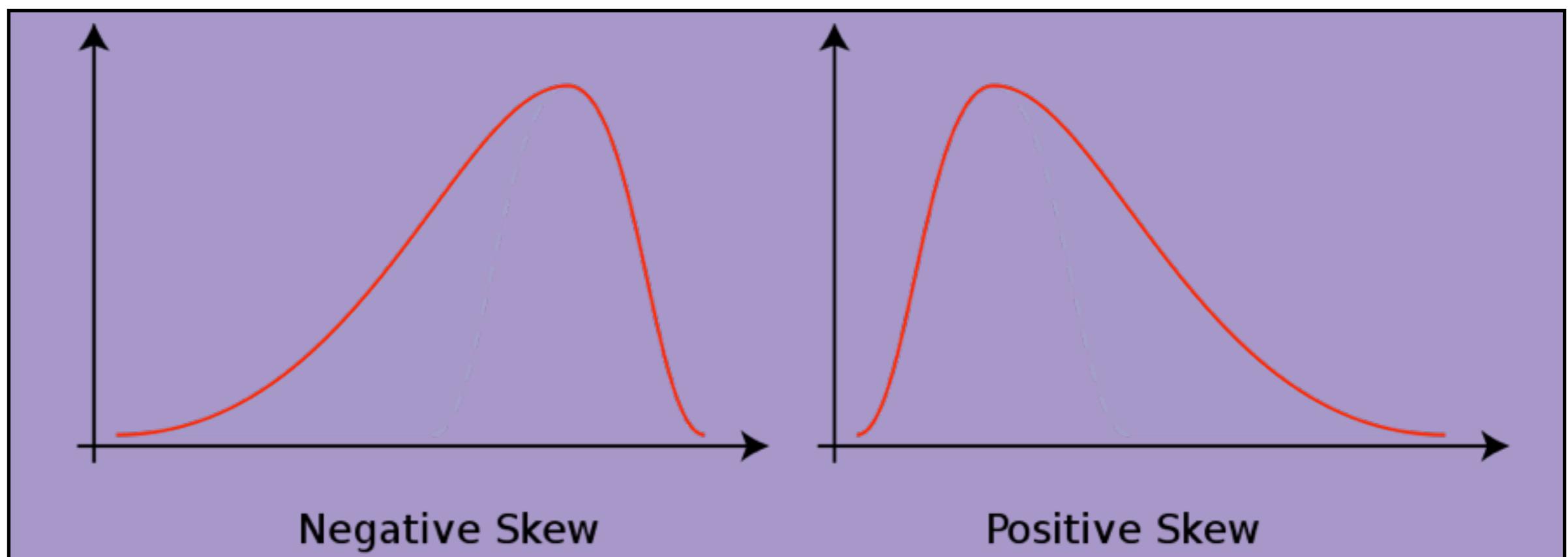
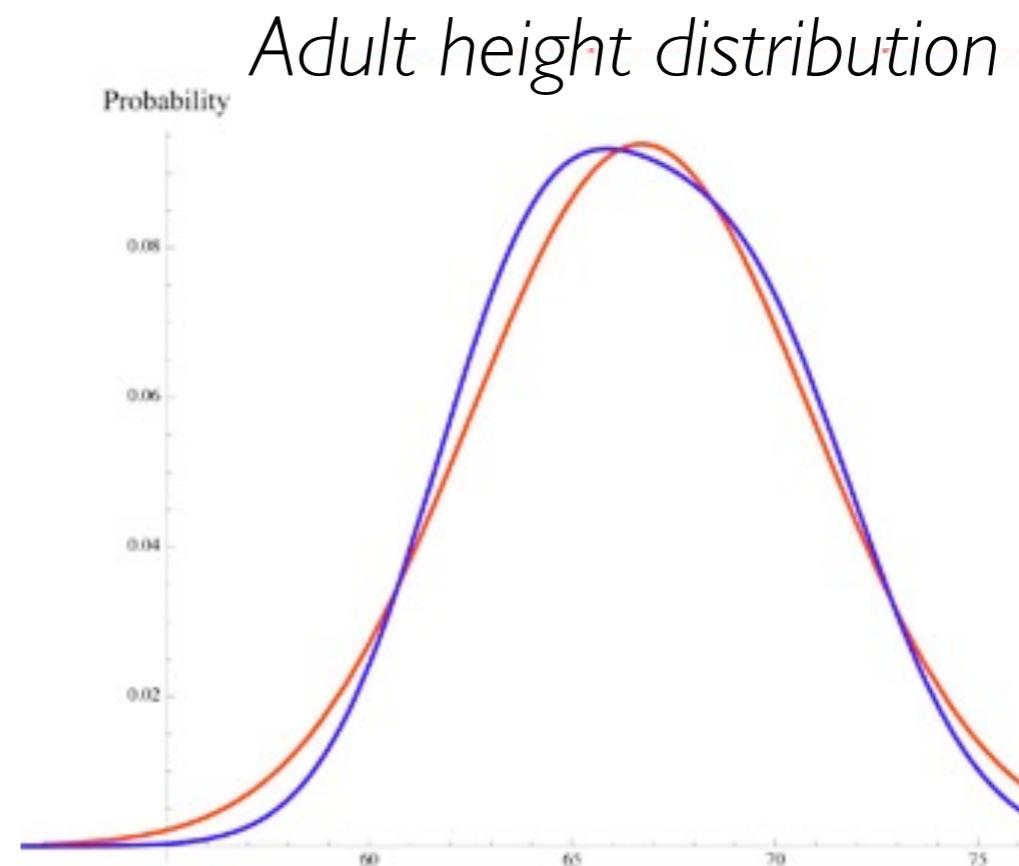


Gauss 1809

Non-randomness: Skewness

The skewness measures cubic deviation from normal distribution:

$$\gamma_1 \equiv \left\langle \left(\frac{\Delta T}{T} \right)^3 \right\rangle$$



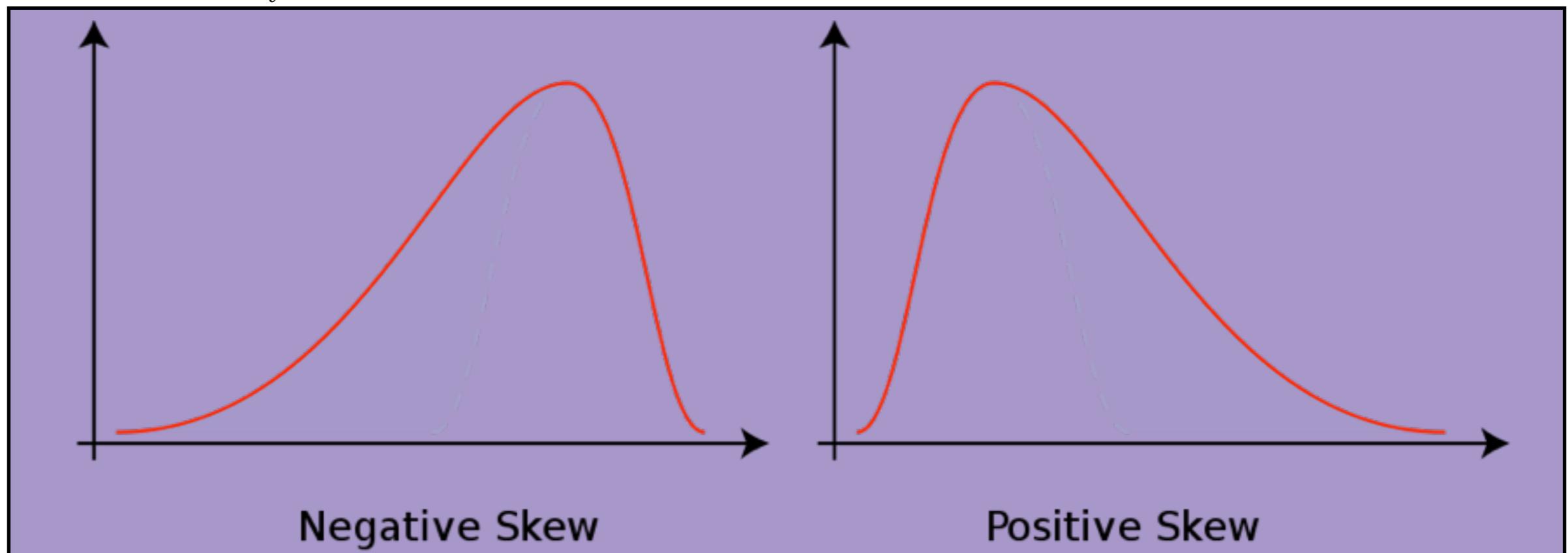
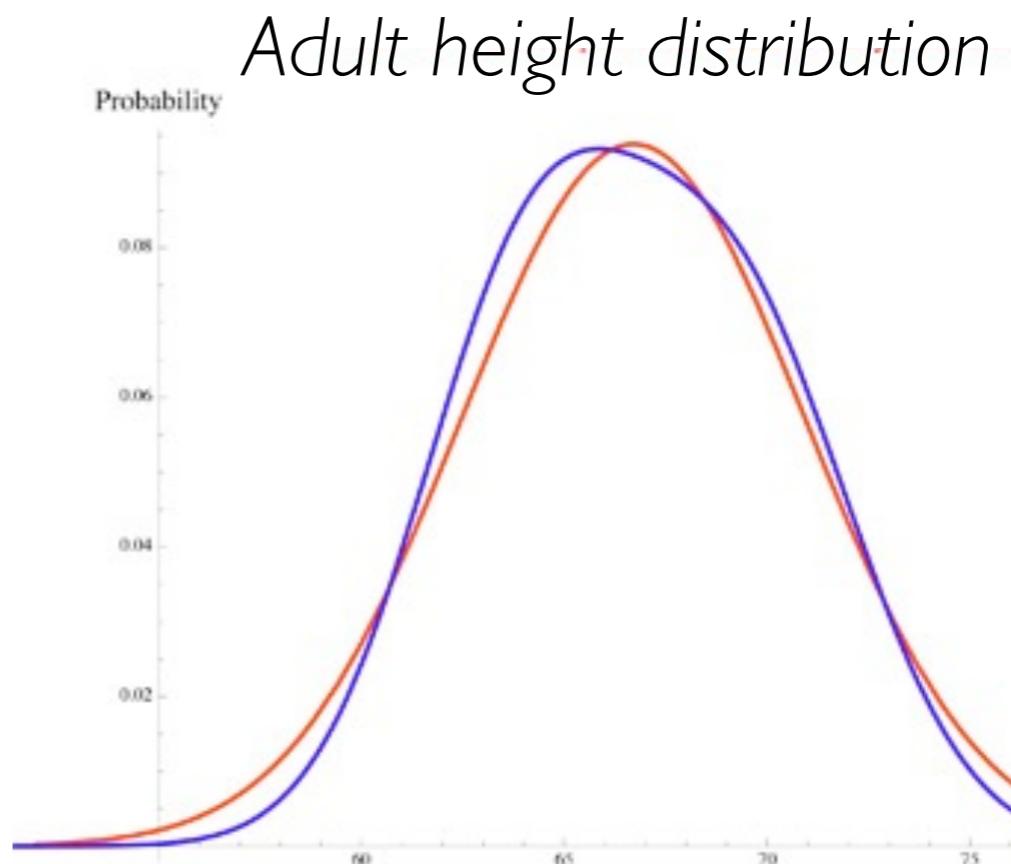
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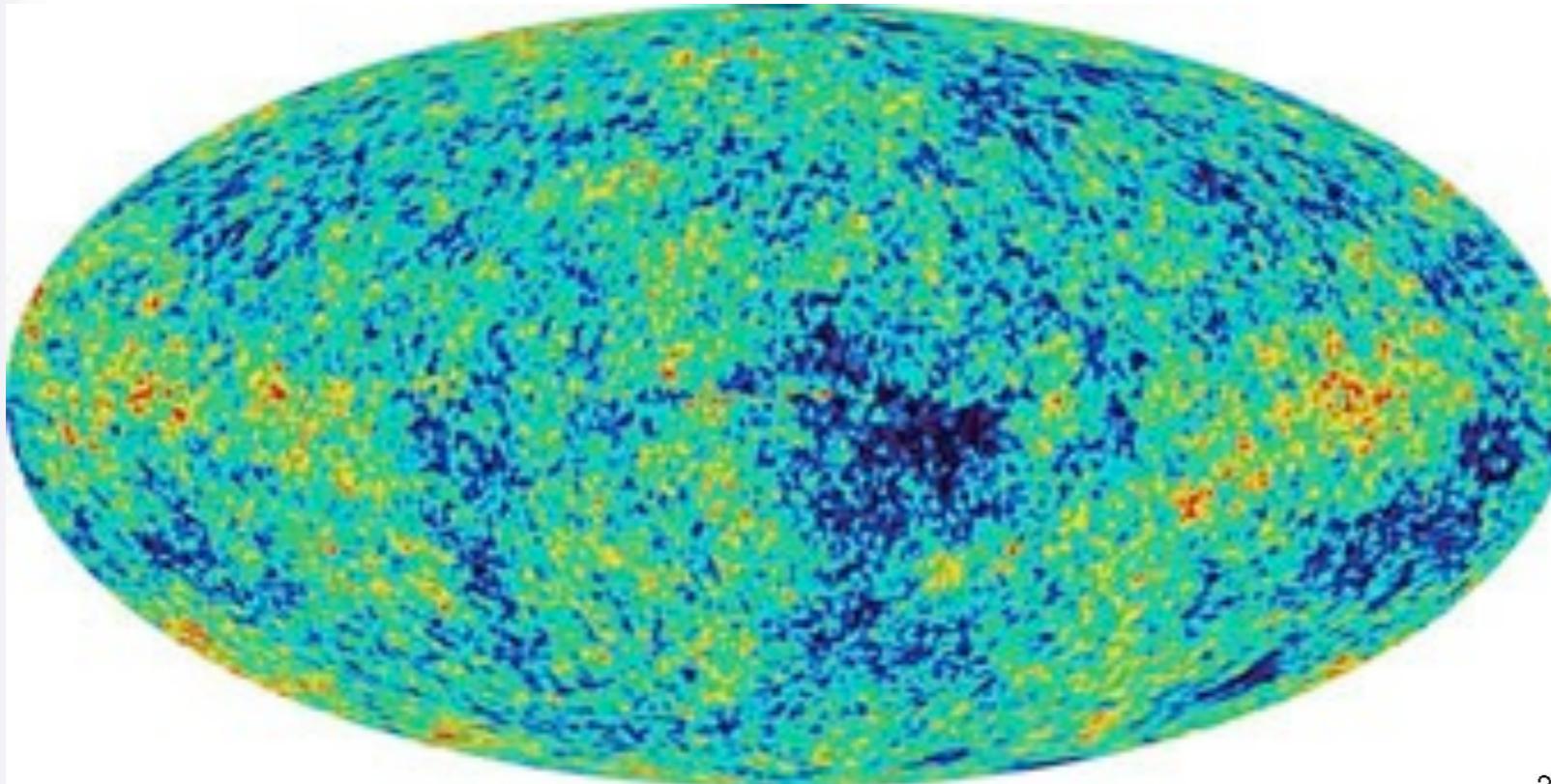
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For the CMB this is the sum over a 3D object - the bispectrum $b_{l_1 l_2 l_3}$

$$\gamma_1 = \frac{1}{4\pi} \sum_{l_i} h_{l_1 l_2 l_3}^2 b_{l_1 l_2 l_3}$$



Non-randomness: the Bispectrum



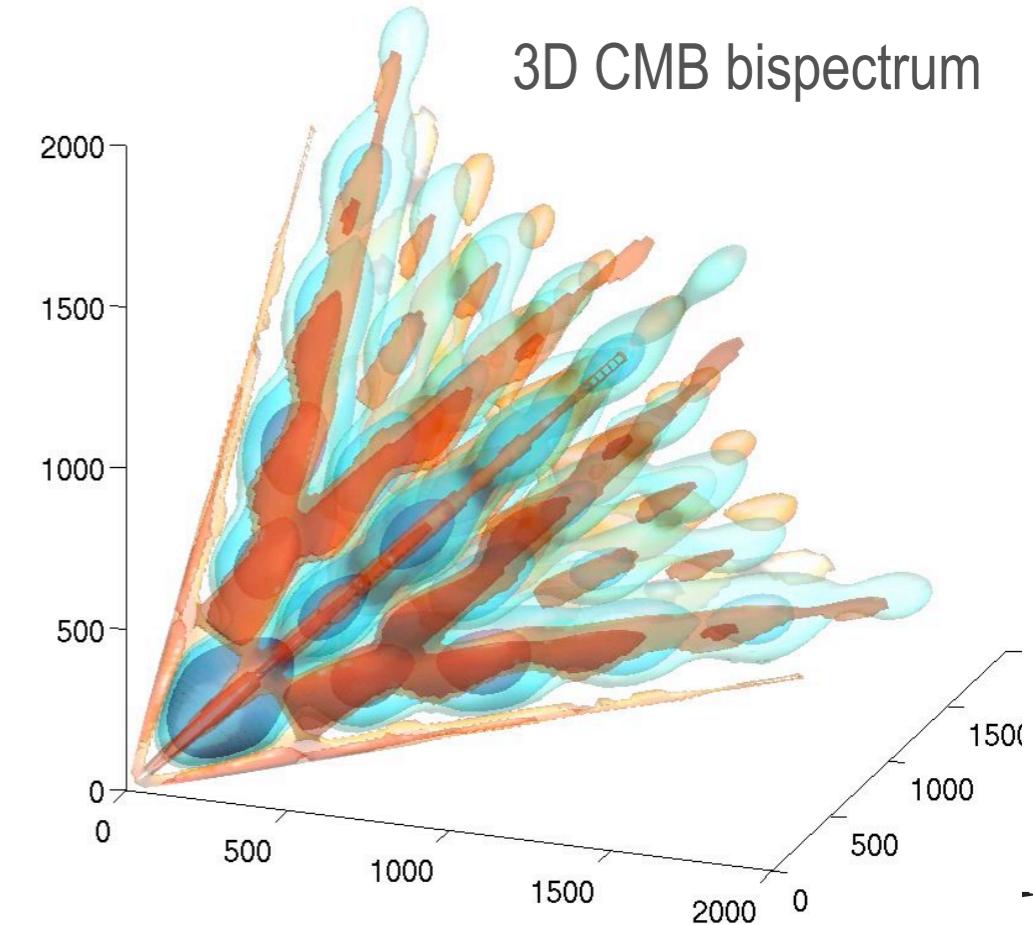
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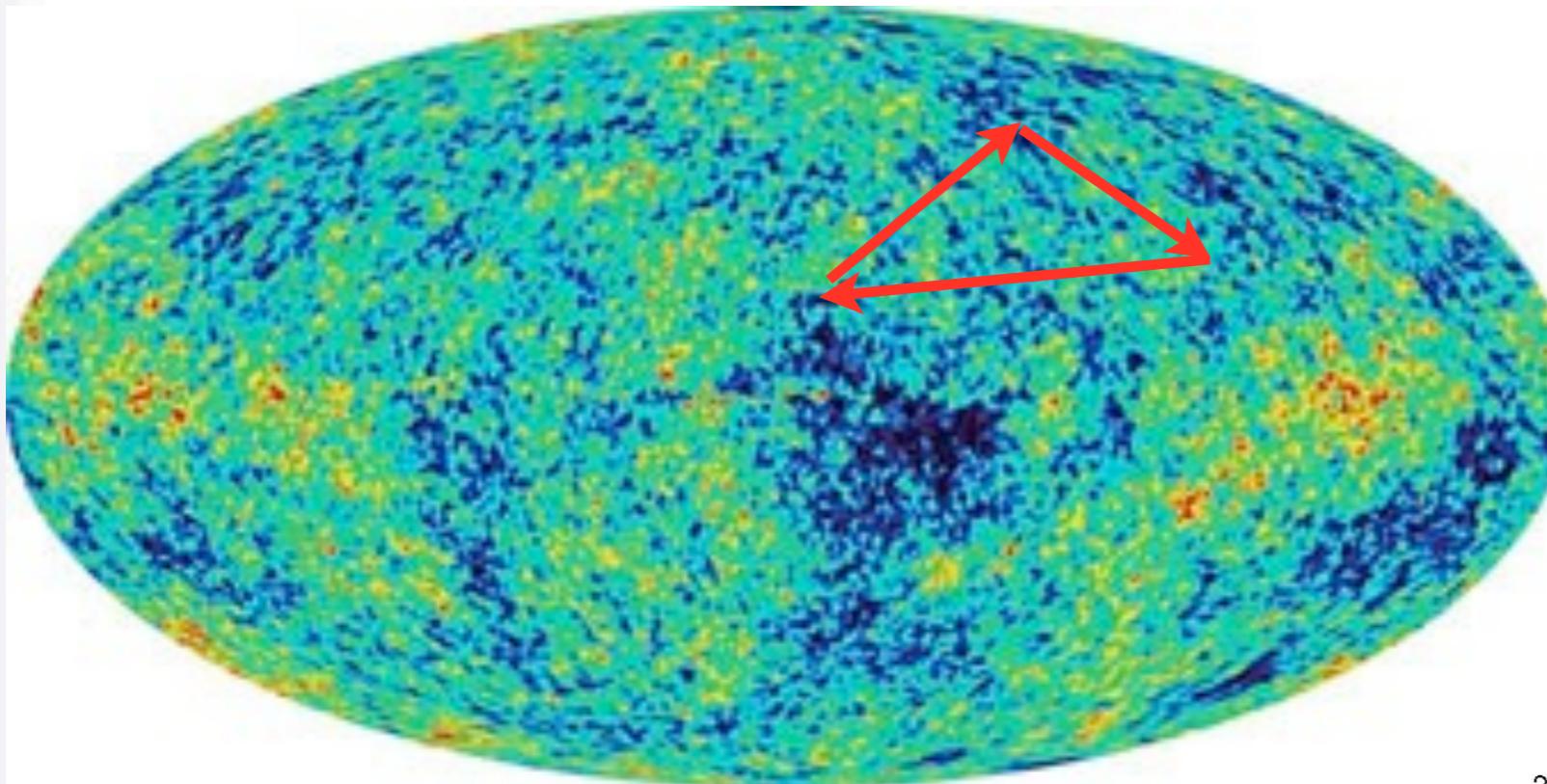
is then a 3D object constructed from the a_{lm} 's with multipoles l , making triangles (the new coords)

$B_{l_1 l_2 l_3}$ exists on a tetrahedron

We search for (isotropic) non-random correlations at three points, i.e. summing over similar triangles with side lengths representing 3 coords.



Non-randomness: the Bispectrum



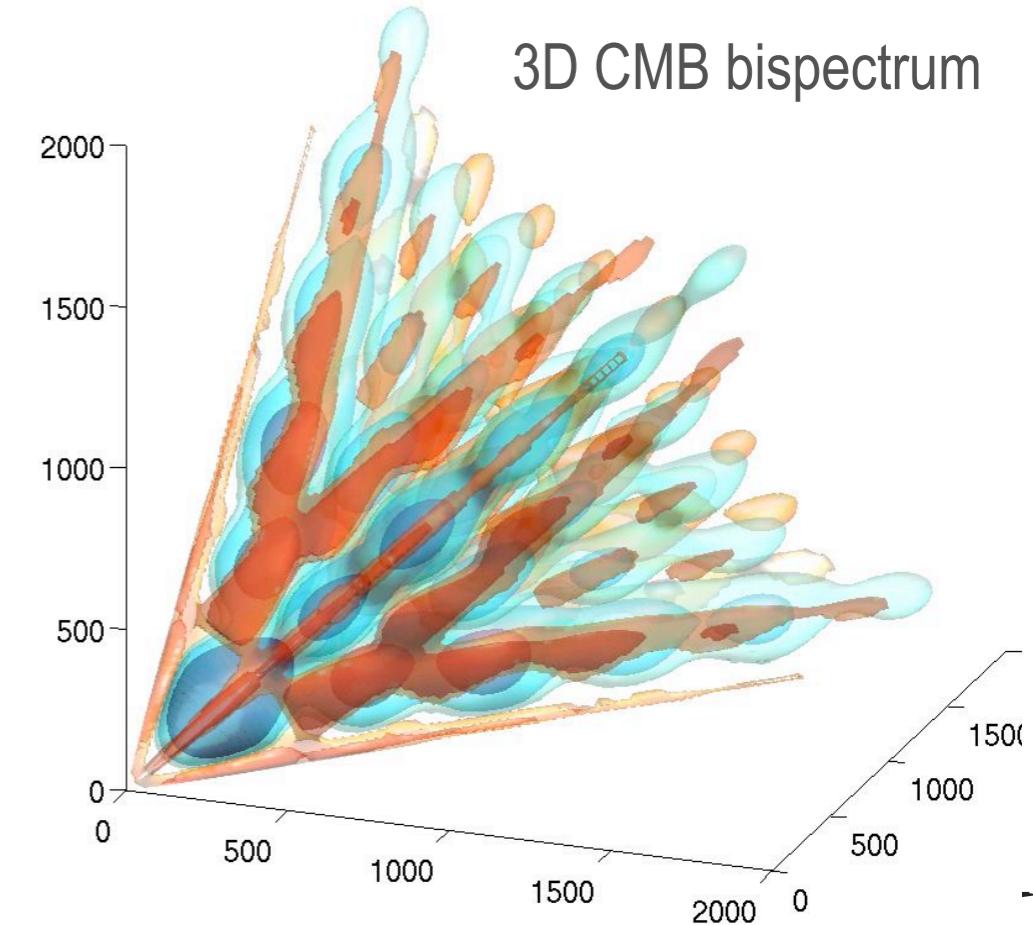
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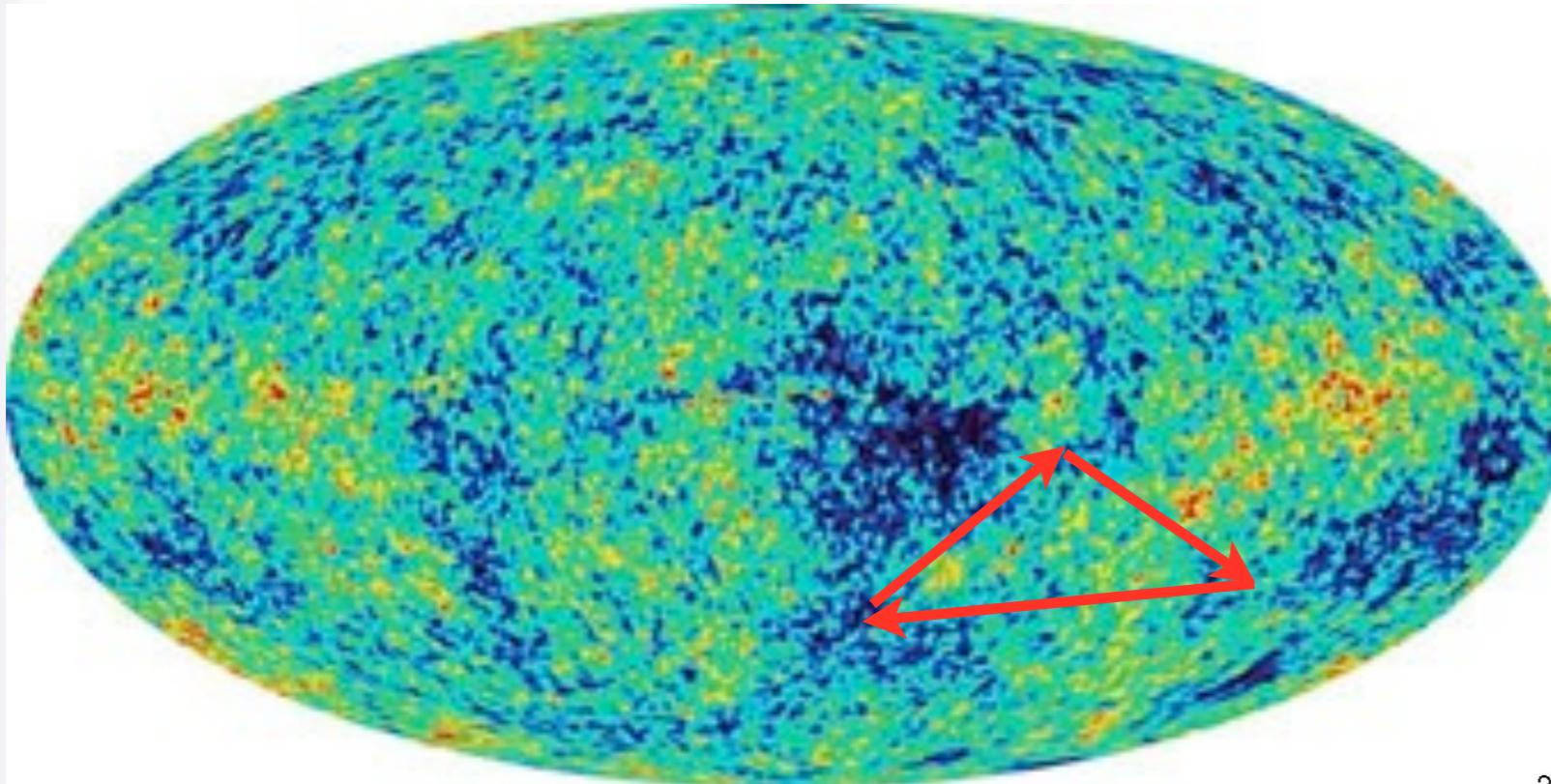
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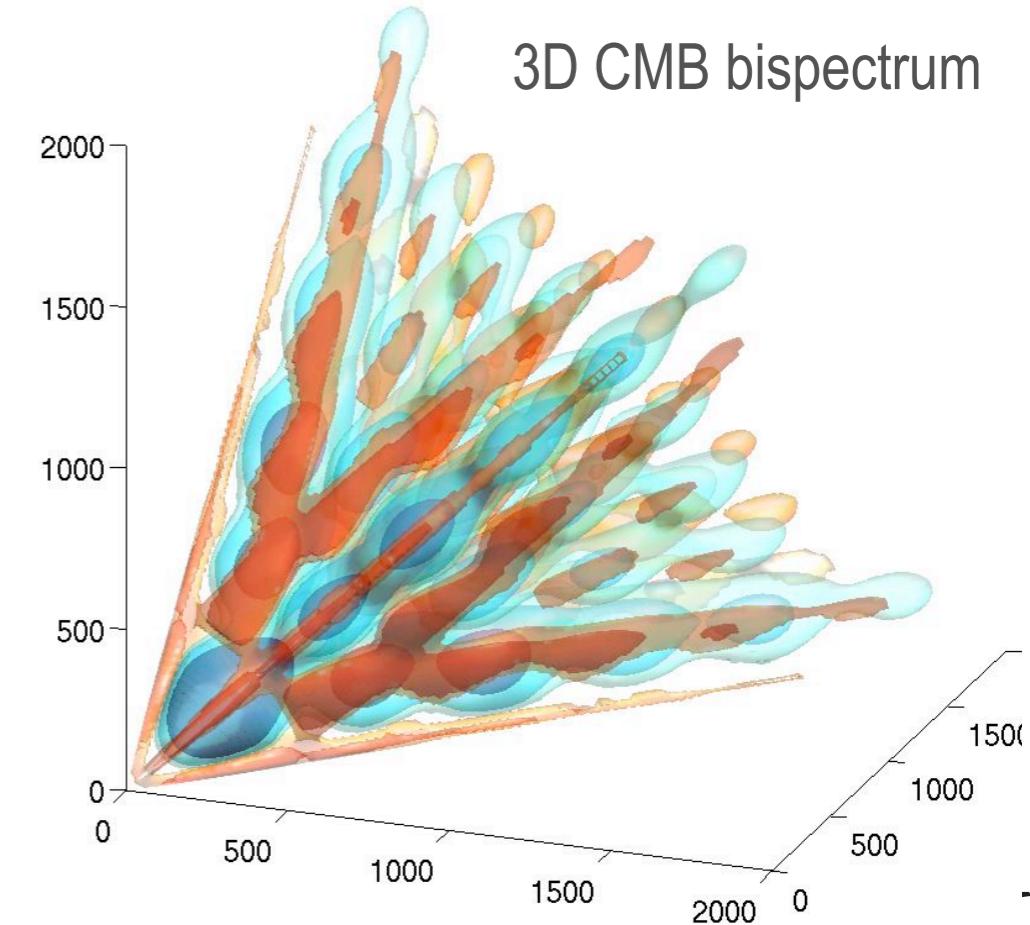
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A Non-Gaussian Universe?

Gaussianity

$$\hat{\Phi}_{\text{lin}} = \Phi_{\text{lin}} \hat{a}^\dagger + \Phi_{\text{lin}}^* \hat{a} \Rightarrow \text{Gaussian with } \langle \hat{\Phi} \hat{\Phi} \hat{\Phi} \rangle = 0$$

Non-Gaussianity (*local model with parameter f_{NL}*)

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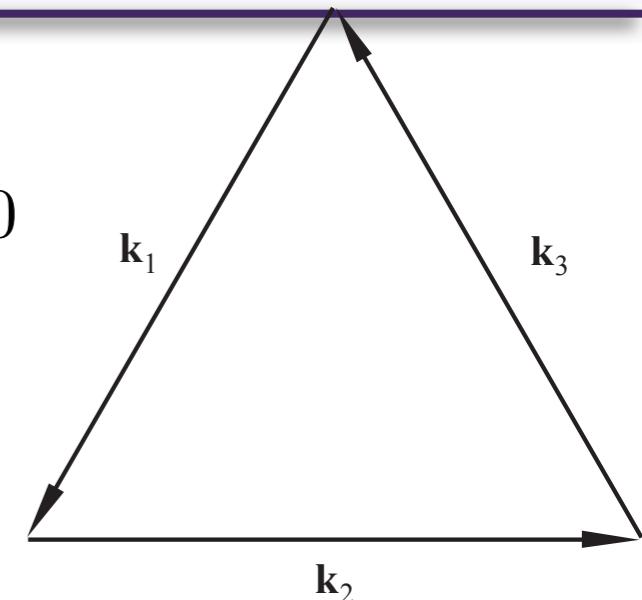
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Perturbative expectations - inflation

Perturbative higher order correlators $f_{NL} \sim O(1)$

Leading order term $\langle \Phi^{(1)} \Phi^{(1)} \Phi^{(2)} \rangle \approx \langle \Phi^{(1)} \Phi^{(1)} (\Phi^{(1)} * \Phi^{(1)}) \rangle$

Second-order gravitational perturbations also $\sim O(1)$



Non-perturbative models - e.g. cosmic strings

Poor suppression of higher orders with signals $f_{NL} \sim P(k)^{-1/2} \sim 10^5 \gg 1$

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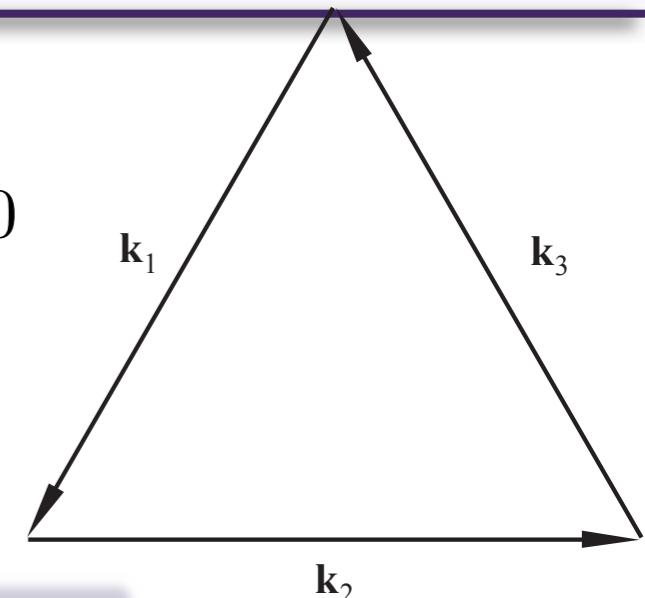
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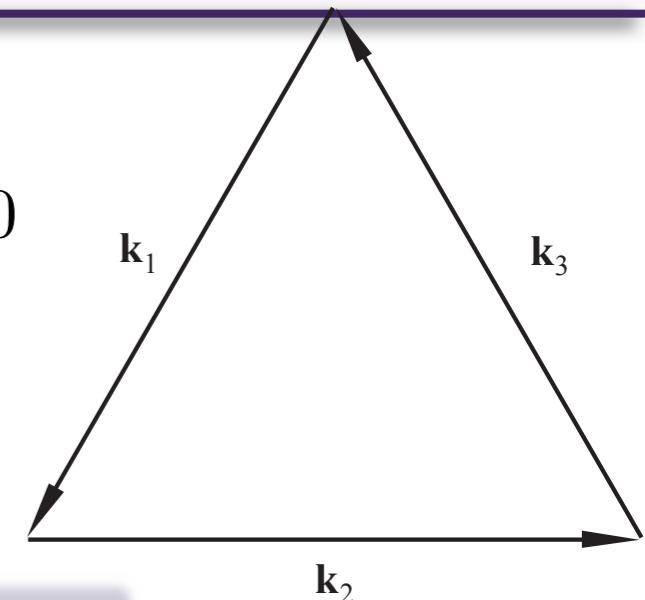
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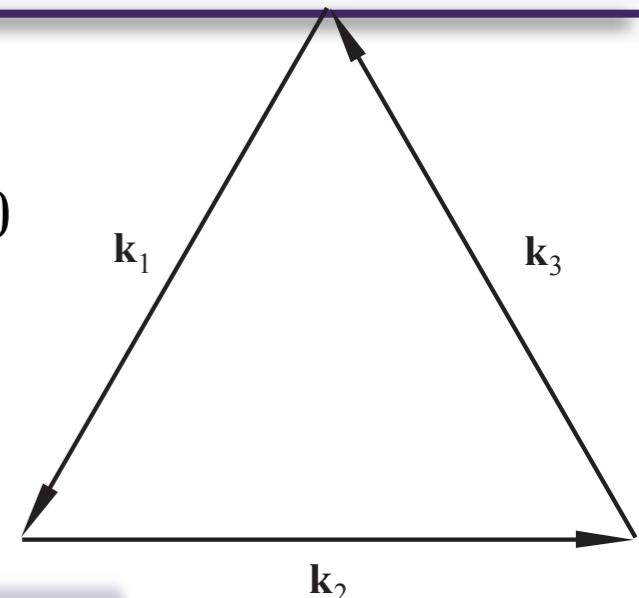
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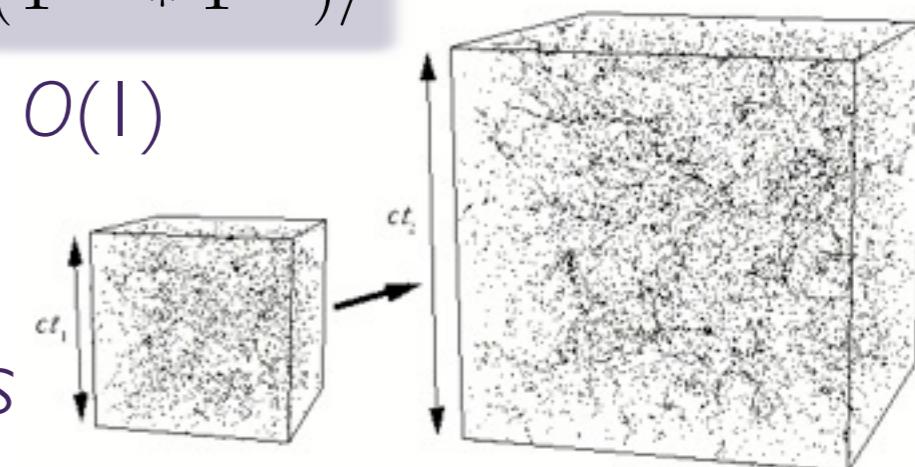


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Exercise 1: Local non-Gaussianity

Assume that higher order corrections to the linear pert. solution are

$$\zeta(\mathbf{x}) = \zeta^{(1)}(\mathbf{x}) + \zeta^{(2)}(\mathbf{x}) + \dots = \zeta^{(1)} + f_{\text{NL}} \left((\zeta^{(1)})^2 - \langle (\zeta^{(1)})^2 \rangle \right) + \dots$$

where the linear solution obeys $\langle \zeta^{(1)}(\mathbf{k}) \zeta^{(1)}(\mathbf{k}') \rangle = (2\pi)^3 P(k) \delta(\mathbf{k} + \mathbf{k}')$

Using the convolution theorem, show that second-order solution can be expressed in the form:

$$\zeta^{(2)}(\mathbf{k}) = \int \frac{d^3 \mathbf{k}'}{(2\pi)^3} \int \frac{d^3 \mathbf{k}''}{(2\pi)^3} \left[\zeta^{(1)}(\mathbf{k}') \zeta^{(1)}(\mathbf{k}'') - \langle \zeta^{(1)}(\mathbf{k}') \zeta^{(1)}(\mathbf{k}'') \rangle \right] \delta(\mathbf{k} - \mathbf{k}' - \mathbf{k}'')$$

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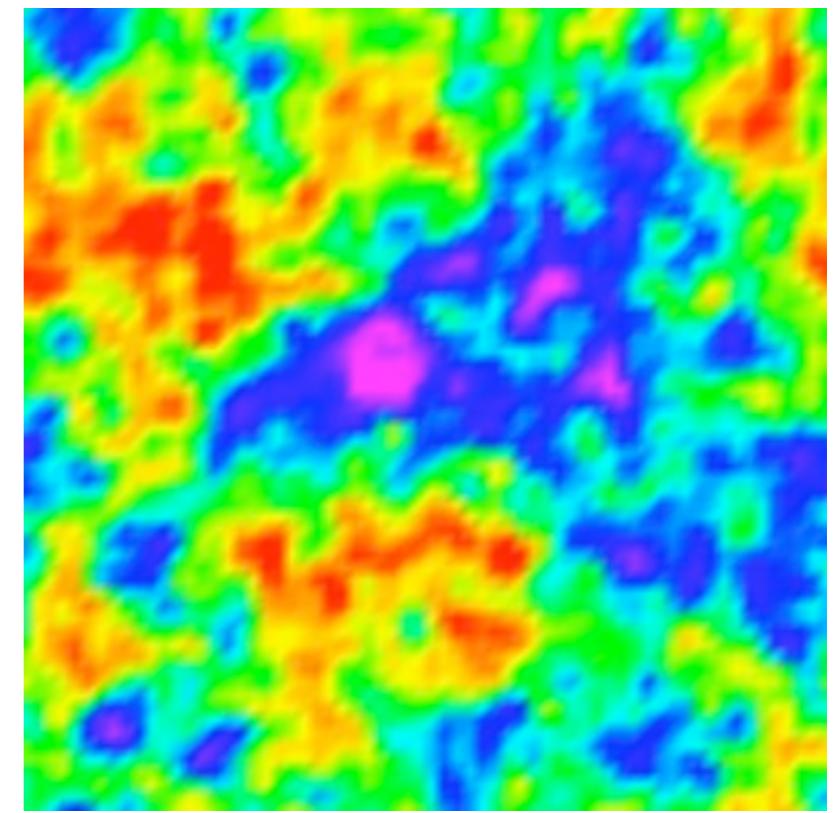
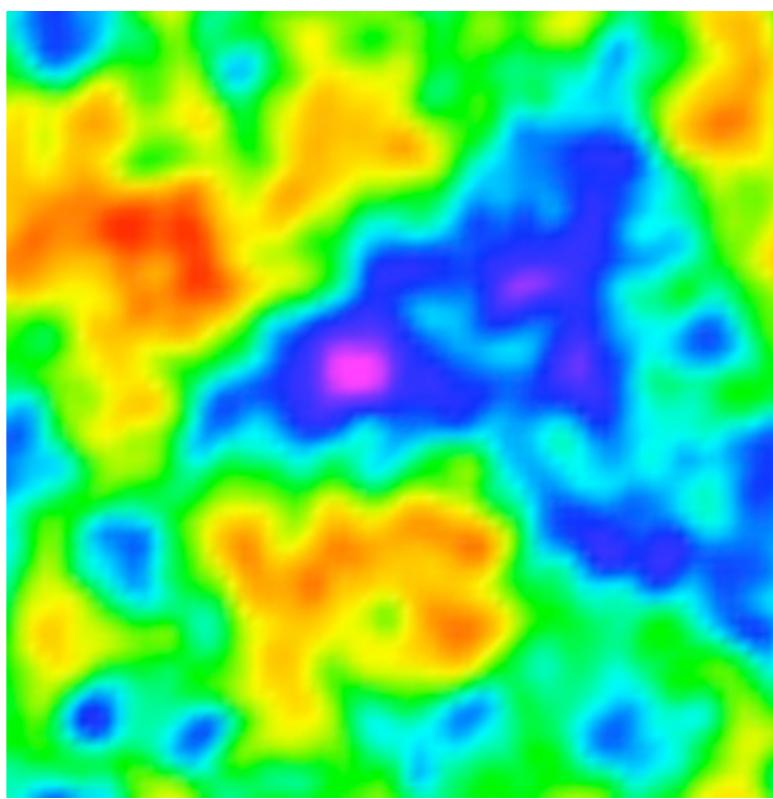
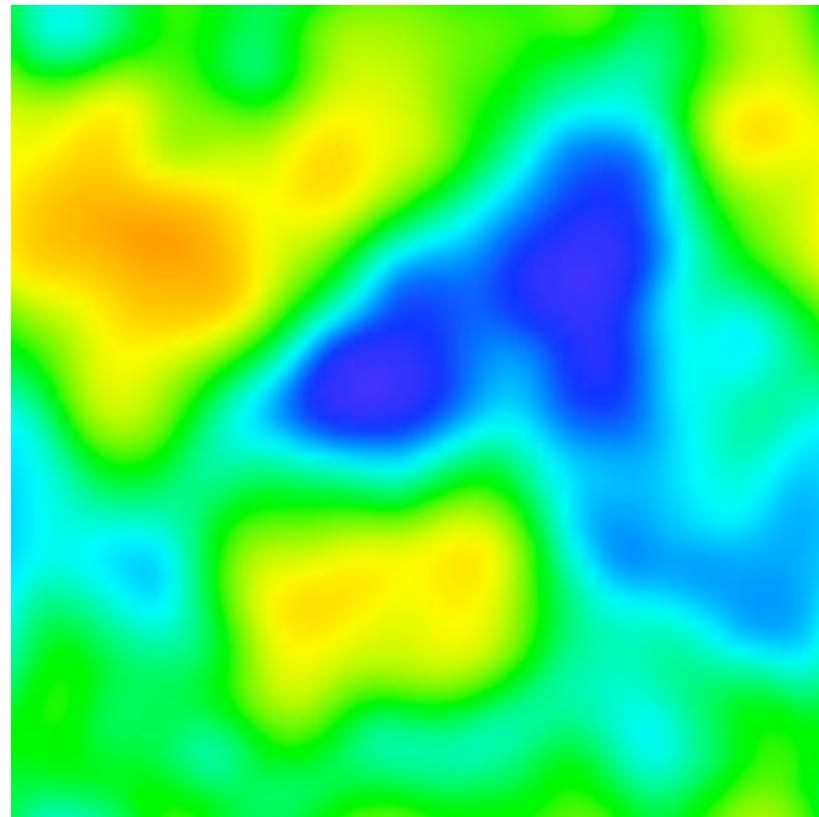
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Origin of local non-Gaussianity

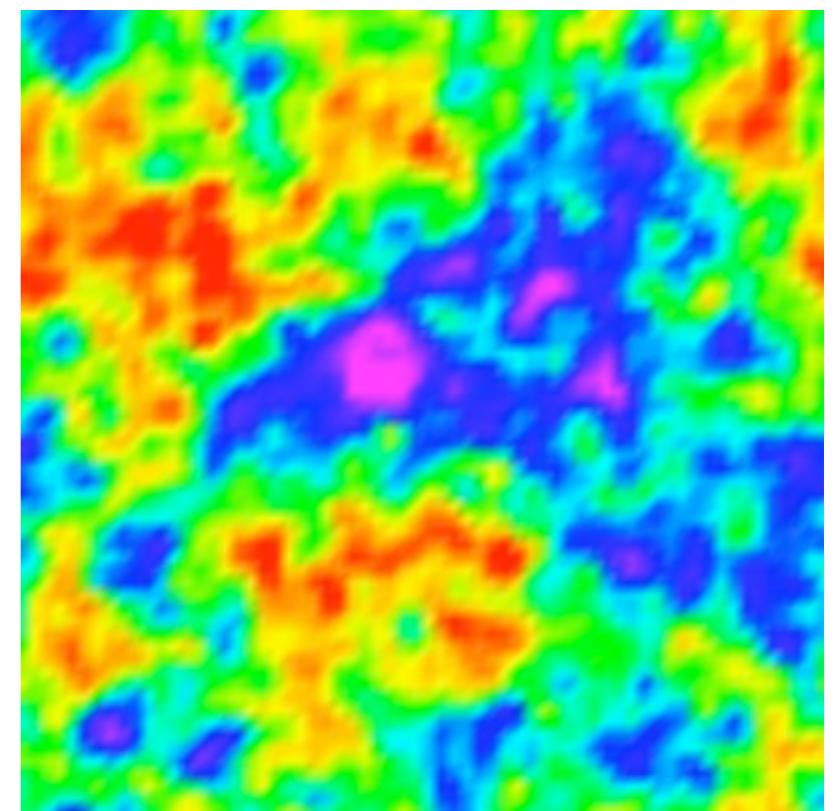
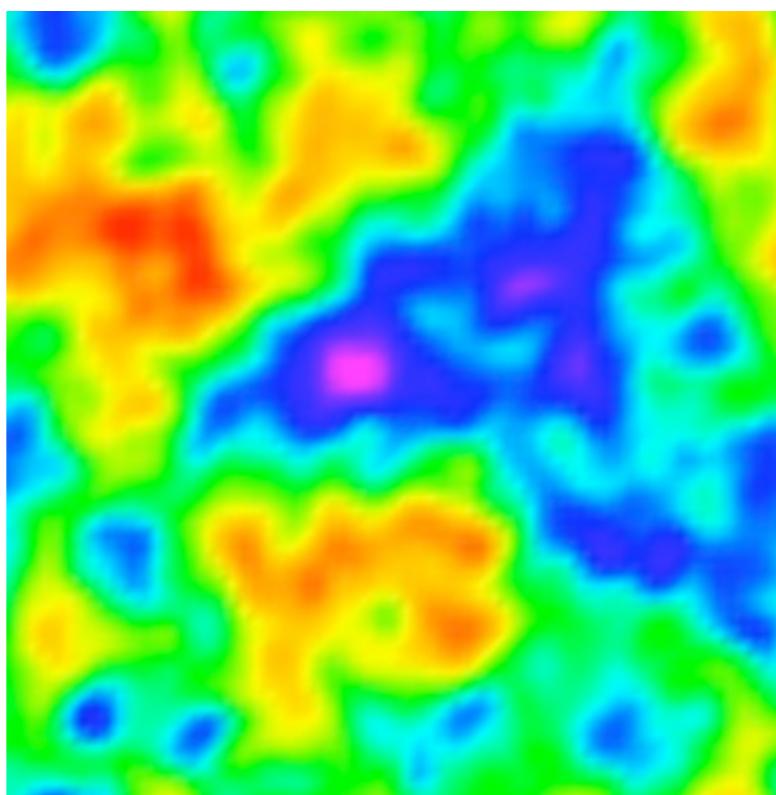
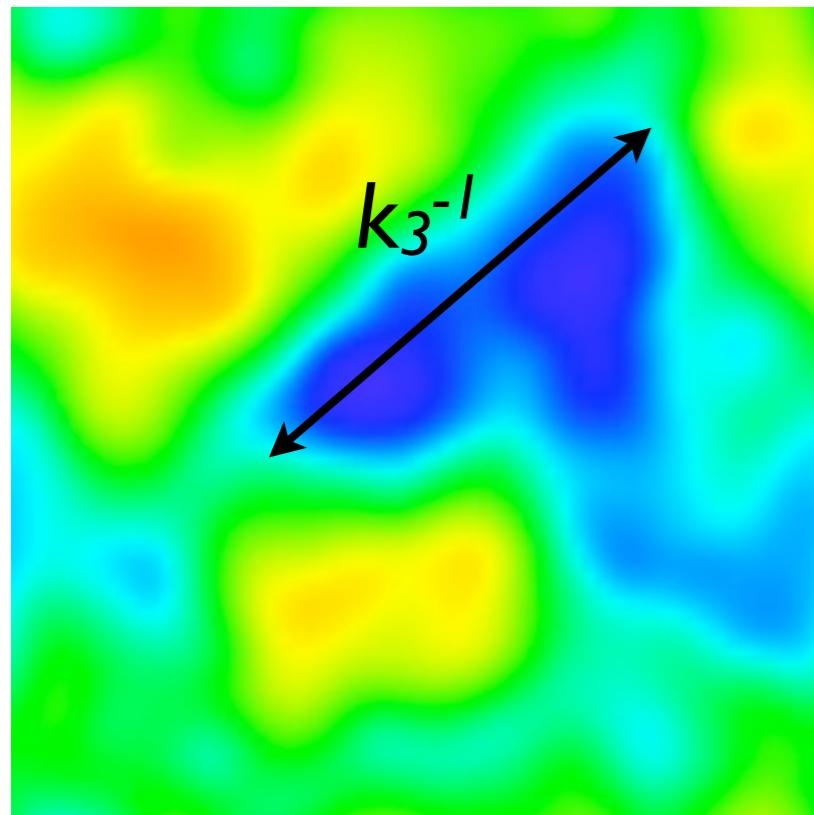


Time →

Fixed comoving volume simulations

Origin of local non-Gaussianity

*Long wavelength k_3
freezes out*

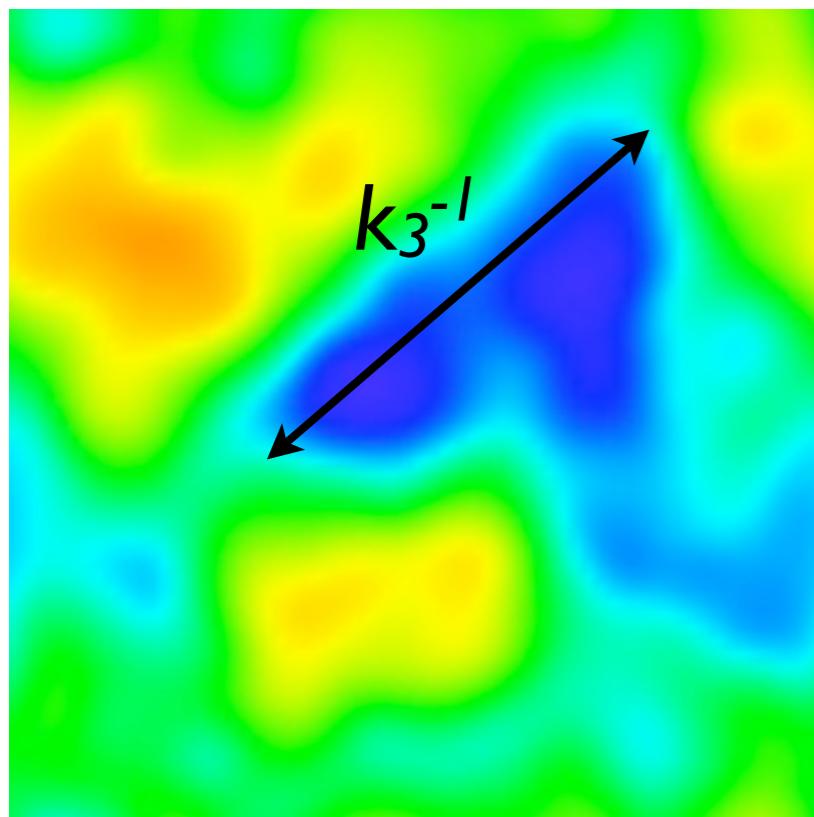


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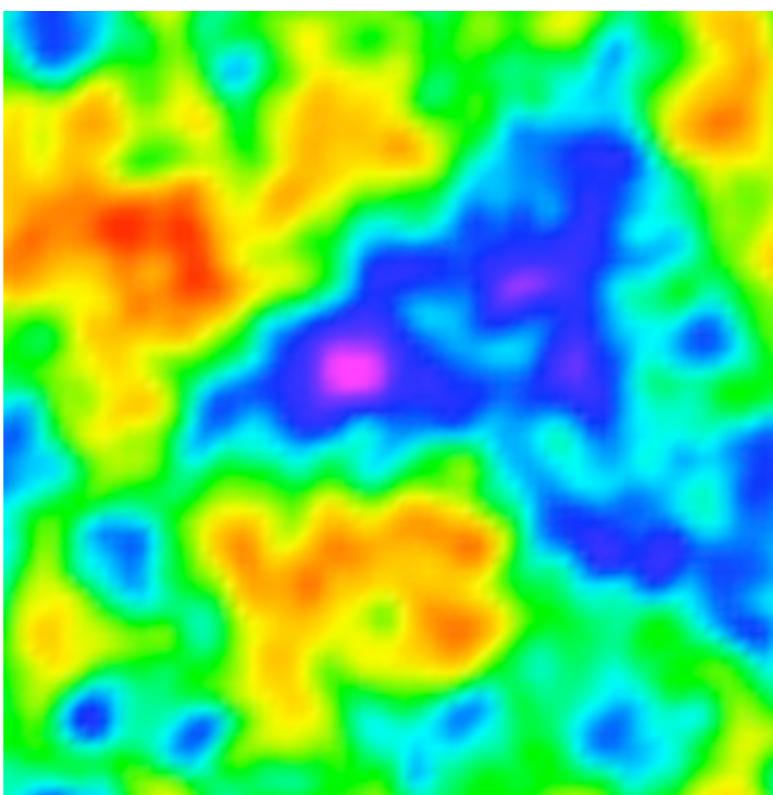
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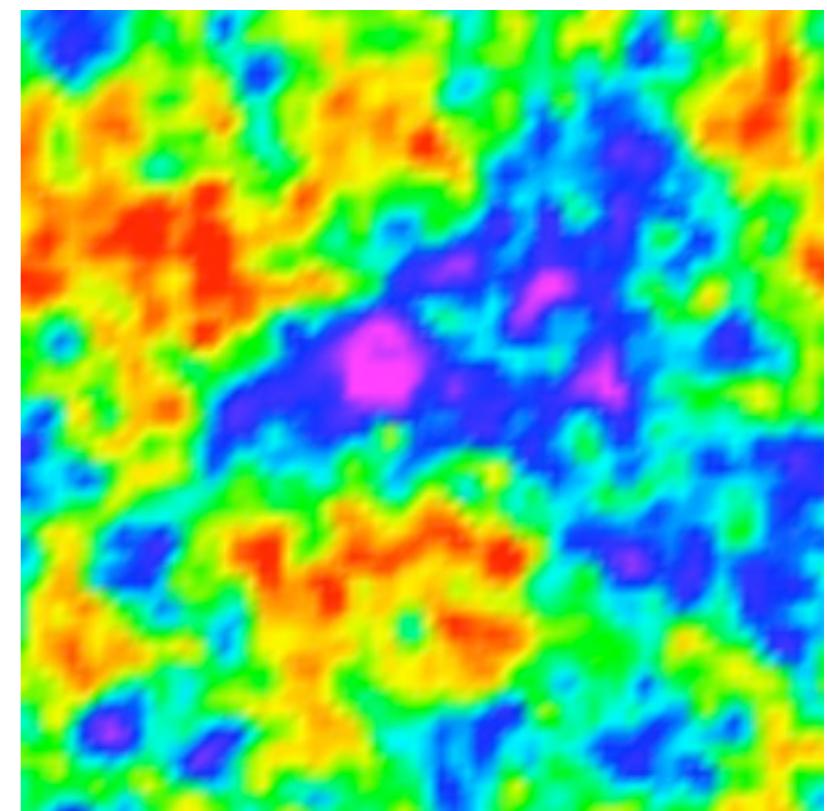
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*Intervening
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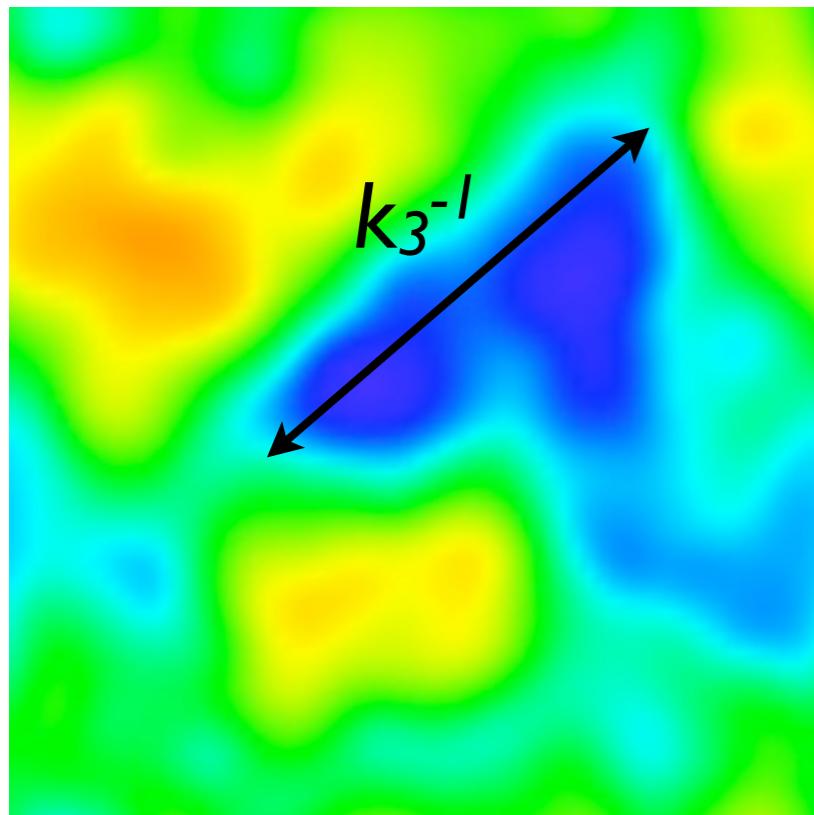
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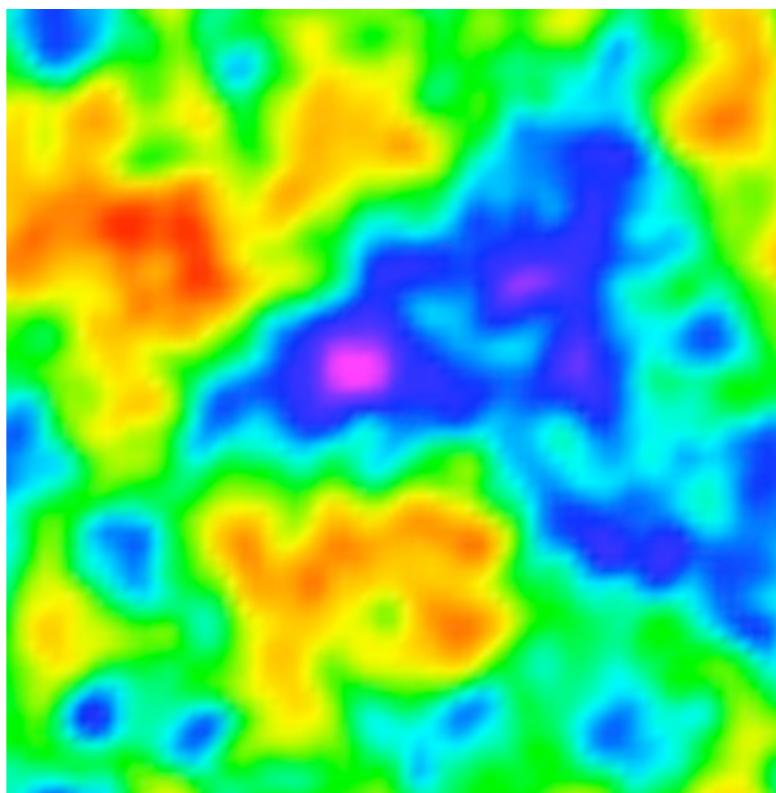
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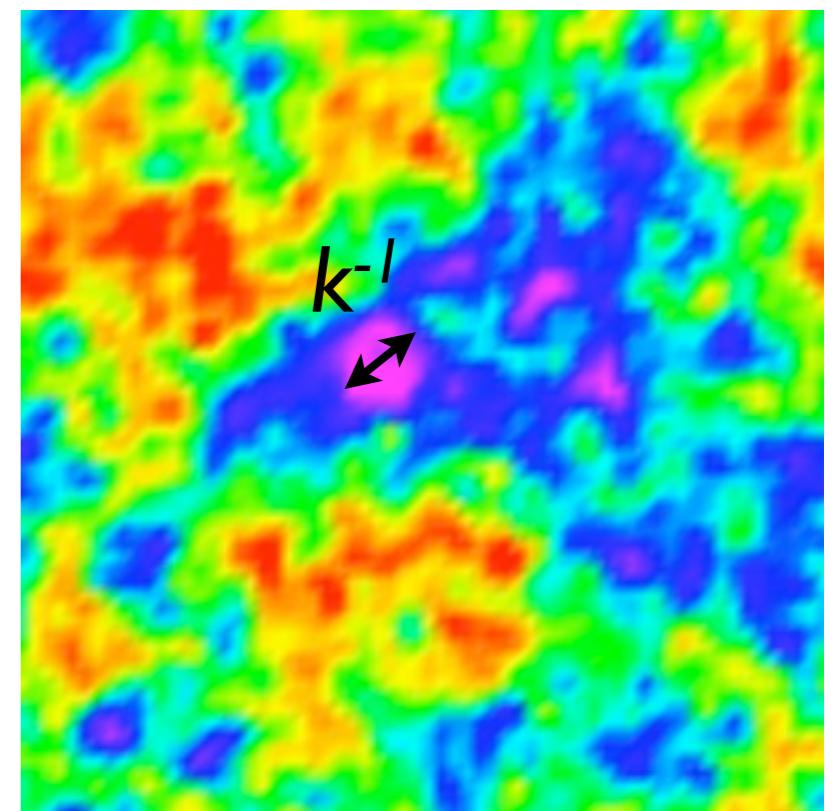
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*Short wavelengths k
influenced by k_3*



Time

Fixed comoving volume simulations

Local Non-Gaussianity

Single field inflation non-Gaussianity (Maldacena, 2003)

*Simple route to NG,
Rigopoulos, et al, 2004*

Nonlinearity for $\zeta_i = \frac{1}{a\sqrt{2\epsilon}} Q_i$ only in prefactor of source $S_i[Q_{\text{lin}}]$

$$S_i = \frac{-\kappa}{2a\sqrt{\epsilon}} \int \frac{d^3 k}{(2\pi)^{3/2}} \dot{W}(k) i k_i e^{ik \cdot x} Q_{\text{lin}}(k) \alpha(k) + \text{c.c.}$$

Expand to leading order ...

$$\partial_i \epsilon = -2\epsilon(2\epsilon - \eta) \longrightarrow \zeta_i^{(2)} = \frac{(4\epsilon - 2\eta) \zeta^{(1)}}{1 - n_s} S_i^{(1)}$$

Squeezed limit with
 $k_3 \ll k_1, k_2 = k$

$$\zeta_i^{(2)} = \frac{(4\epsilon - 2\eta) \zeta^{(1)}}{1 - n_s} S_i^{(1)}$$

↑ ↑ ↑
 2nd order
solution Long mode k_3
background Short mode
 k source

Schematically 2nd order solution is
yields 3-pt correlator or bispectrum:

$$\zeta^{(2)}(k) \sim (4\epsilon - 2\eta) P(k)^{1/2} P(k_3)^{1/2}$$

$$\langle \zeta^{(1)}(k) \zeta^{(1)}(k_3) \zeta^{(2)}(k) \rangle \sim (4\epsilon - 2\eta) P(k) P(k_3) \sim (4\epsilon - 2\eta) \frac{1}{k^3 k_3^3} \frac{H^2}{\epsilon} \Big|_{k_3} \frac{H_k^2}{\epsilon_k} \Big|_k$$

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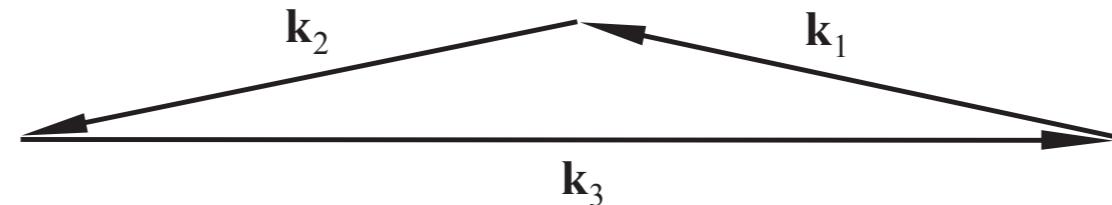
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Primordial Bispectrum

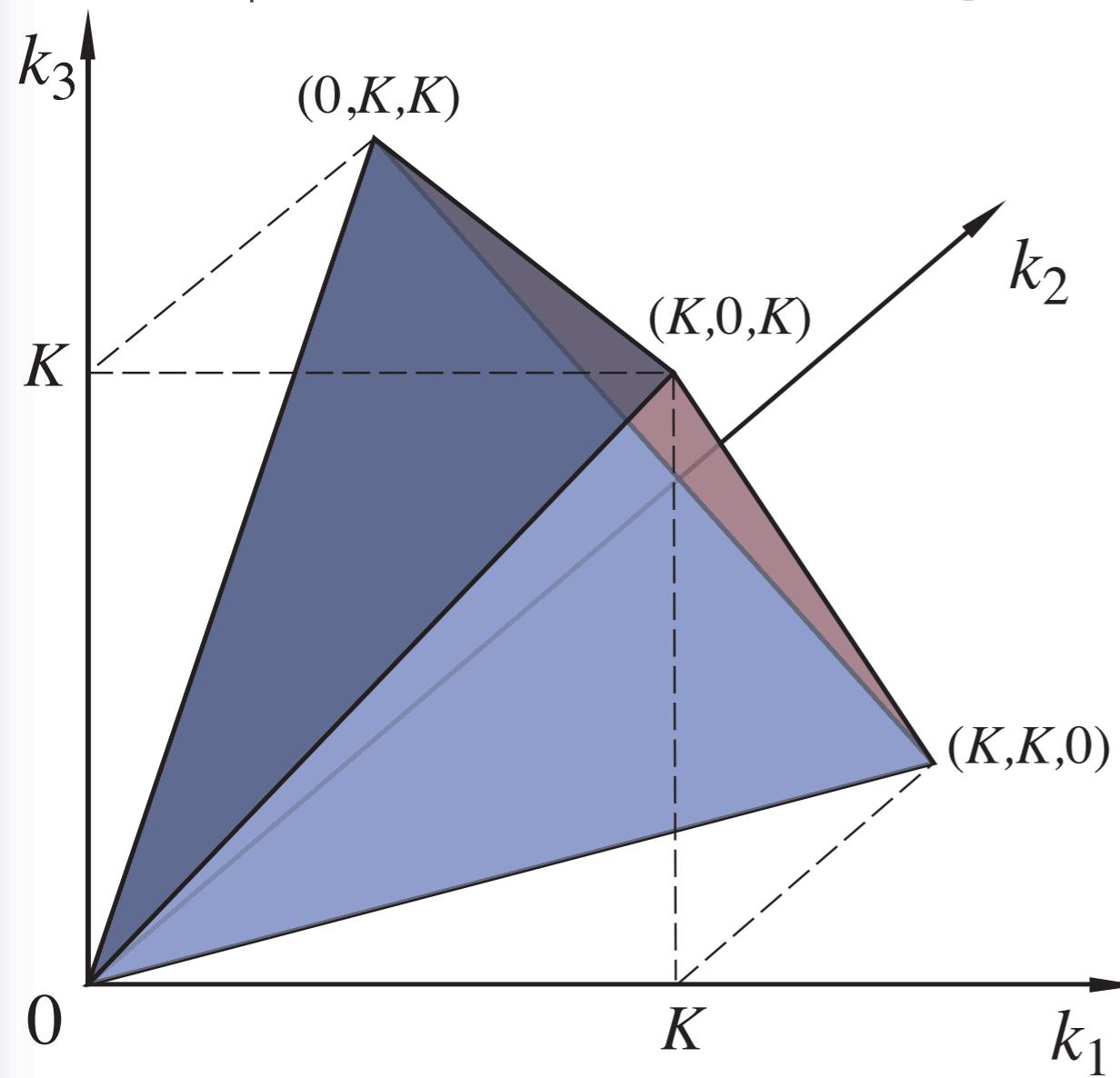
Defined by

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

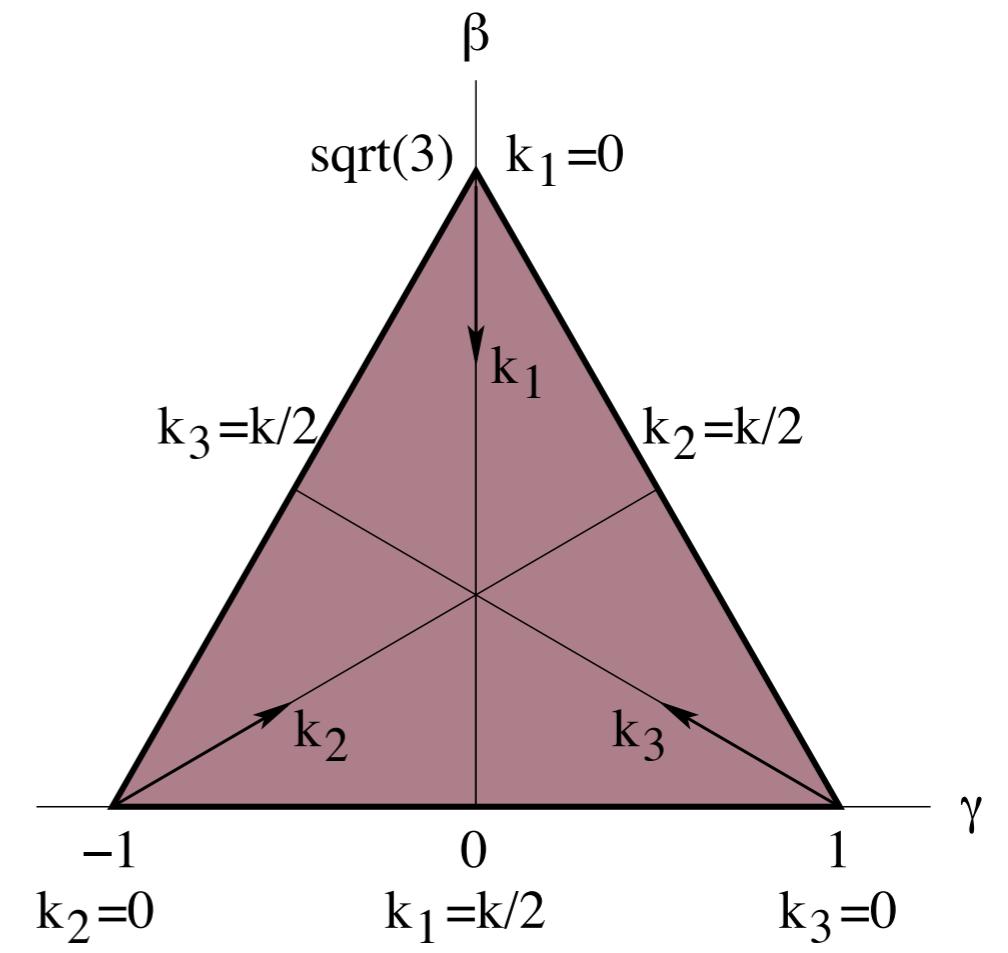
Note the triangle condition



Imposes **3D tetrahedral bispectrum domain** for the wavenumbers k_1, k_2, k_3



Slices $\tilde{k} \equiv \frac{1}{2}(k_1 + k_2 + k_3) = \text{const.}$



Local shape function

There is an overall k^{-6} scaling in the bispectrum, so define

$$S(k_1, k_2, k_3) = \frac{1}{N} (k_1 k_2 k_3)^2 B(k_1, k_2, k_3)$$

which is the (scale-invariant) shape function

For the local $B(k_1, k_2, k_3) = P(k_1)P(k_2) + P(k_2)P(k_3) + P(k_3)P(k_1)$

$$S(k_1, k_2, k_3) \approx \frac{1}{3} \left(\frac{k_1^3 + k_2^3 + k_3^3}{k_1 k_2 k_3} \right)$$

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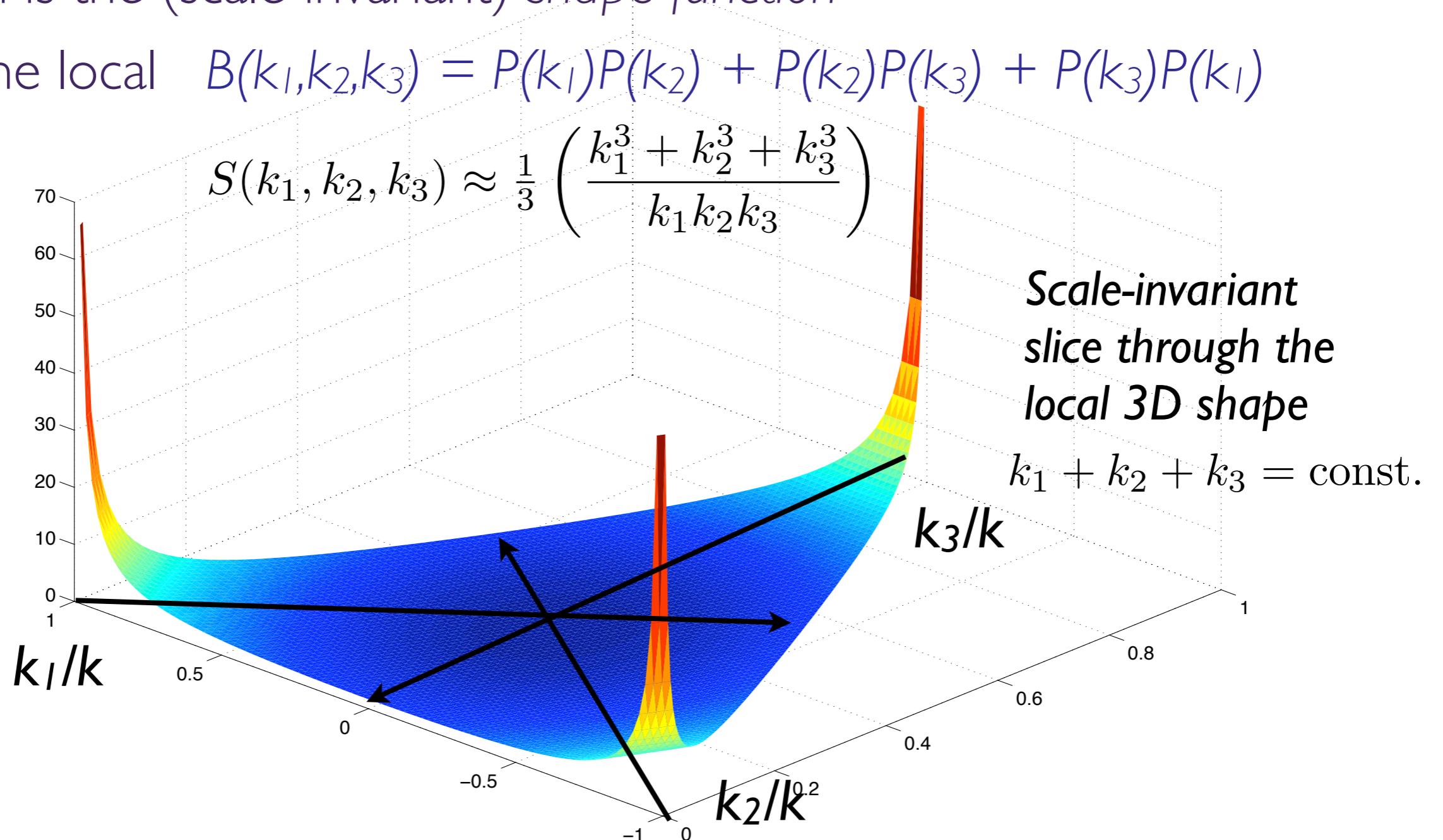
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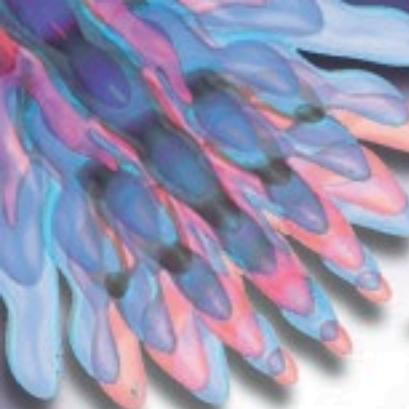
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No-Go Theorem

Simple inflation models cannot generate observable non-Gaussianity:

- single scalar field
- canonical kinetic terms
- always slow roll
- ground state initial vacuum
- standard Einstein gravity

Non-Gaussianity is arguably the most stringent test of the standard picture

But simple inflation model-building faces rigorous challenges in fundamental theory (e.g. *eta problem* and *super-Planckian field values*). Many new ideas/solutions violate these conditions!

Inflation in string theory

Motivated by deficiencies of inflation

UV completeness sensitivity

Super-Planckian fields

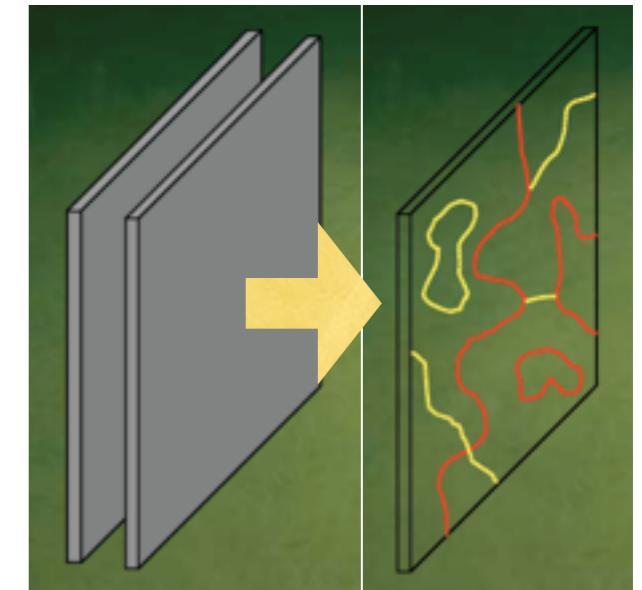
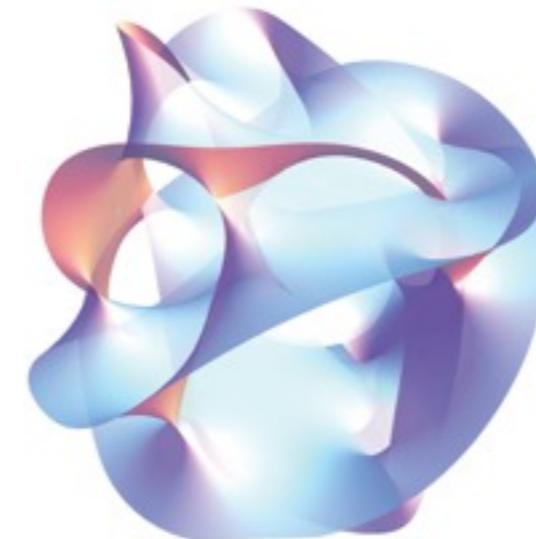
Eta problem - quantum corrections

Challenges to effective field theory

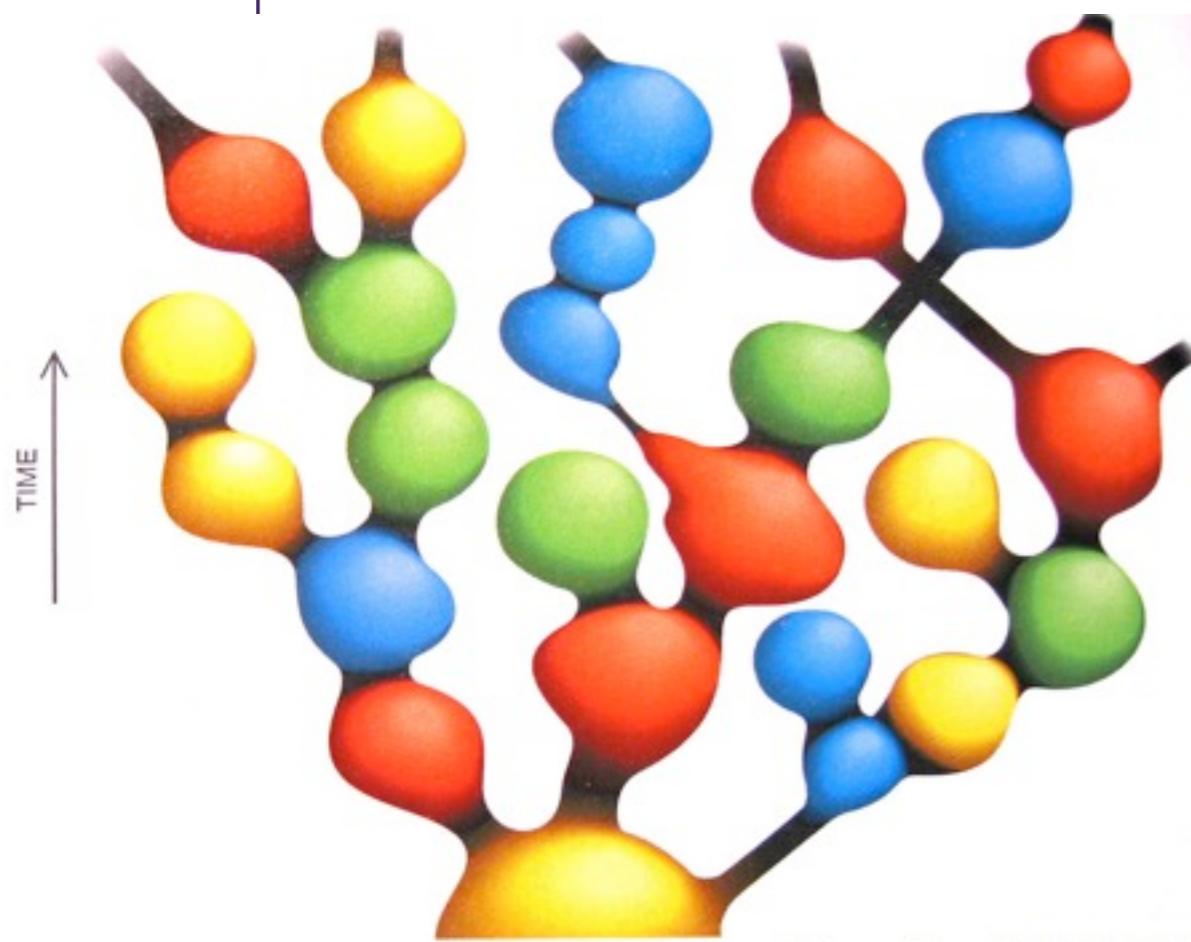
Rich structure

warped branes, axions, eternal inflation

landscape ...



Dvali & Tye, 2000
Sarangi & Tye, 2002
KKLMMT, 2003



Multiple field inflation

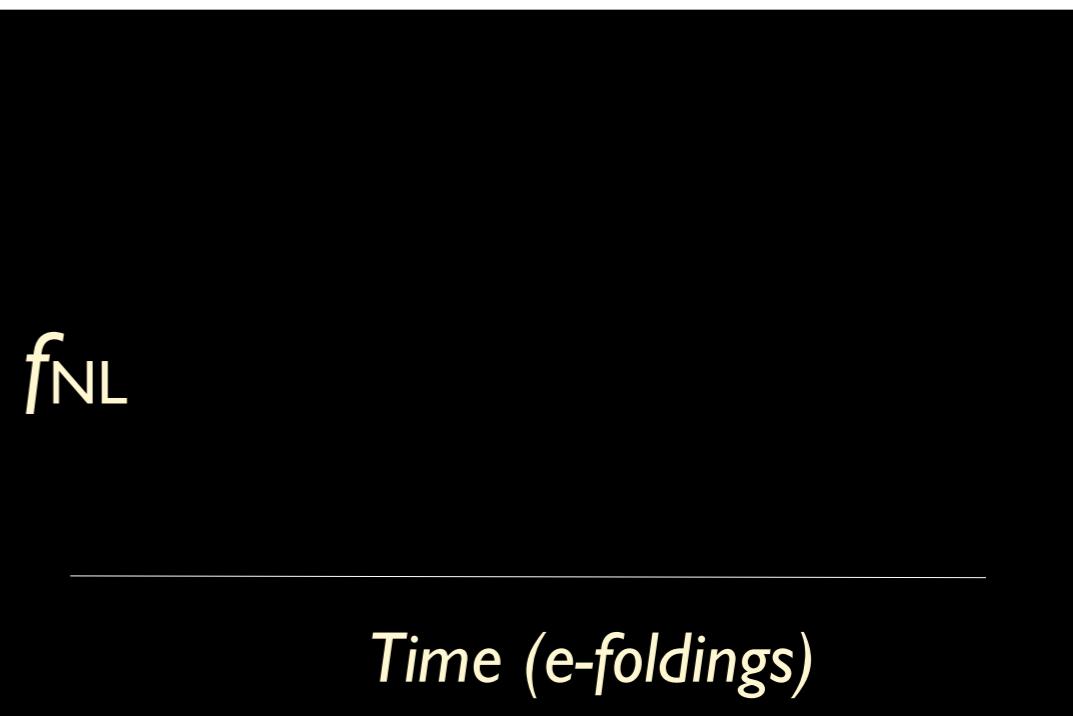
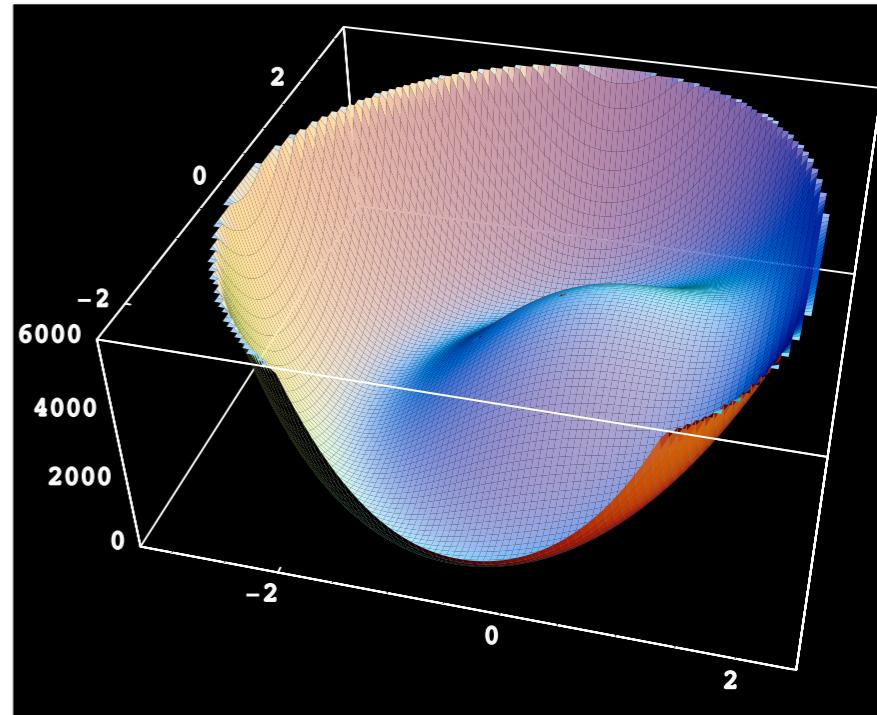
Non-Gaussianity from interacting potentials

$$V(\phi_1, \phi_2) = \frac{1}{4}\lambda(\phi_1^2 + \phi_2^2 - m^2)^2 + \nu(\phi_1 + m)^3$$

Significant final f_{NL} ingredients:

- corner turning
- nontrivial potential
- or breakout (hybrid models)

*Rigopoulos, EPS, van Tent 05, 06
see also Vernizzi & Wands 06,
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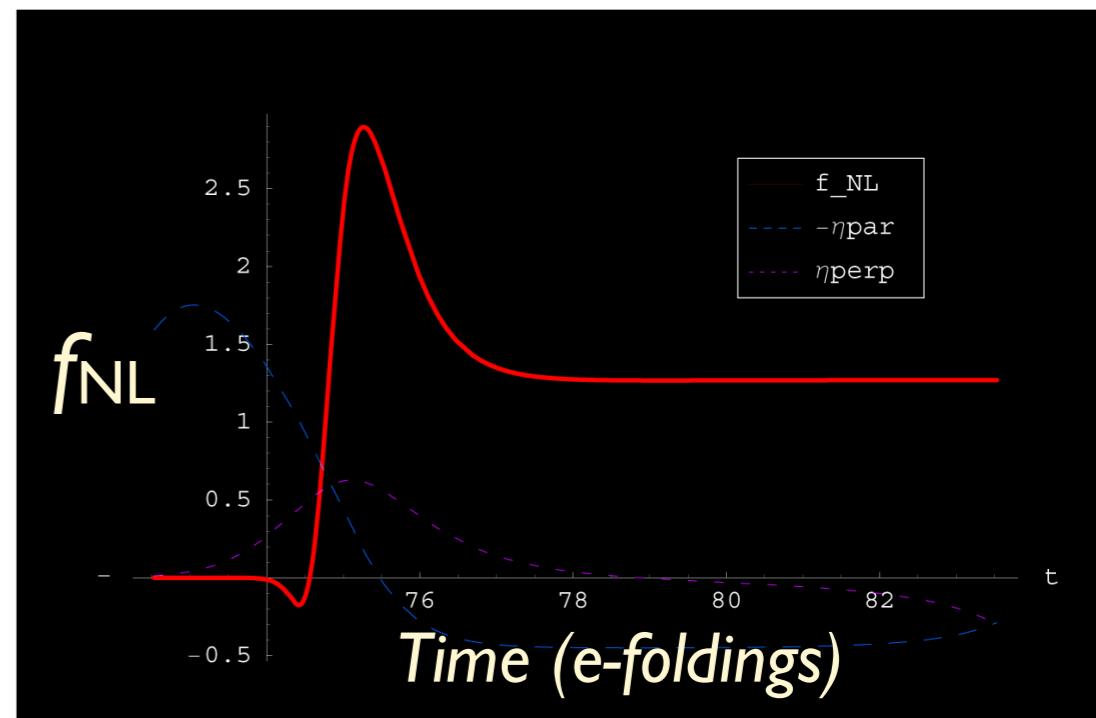
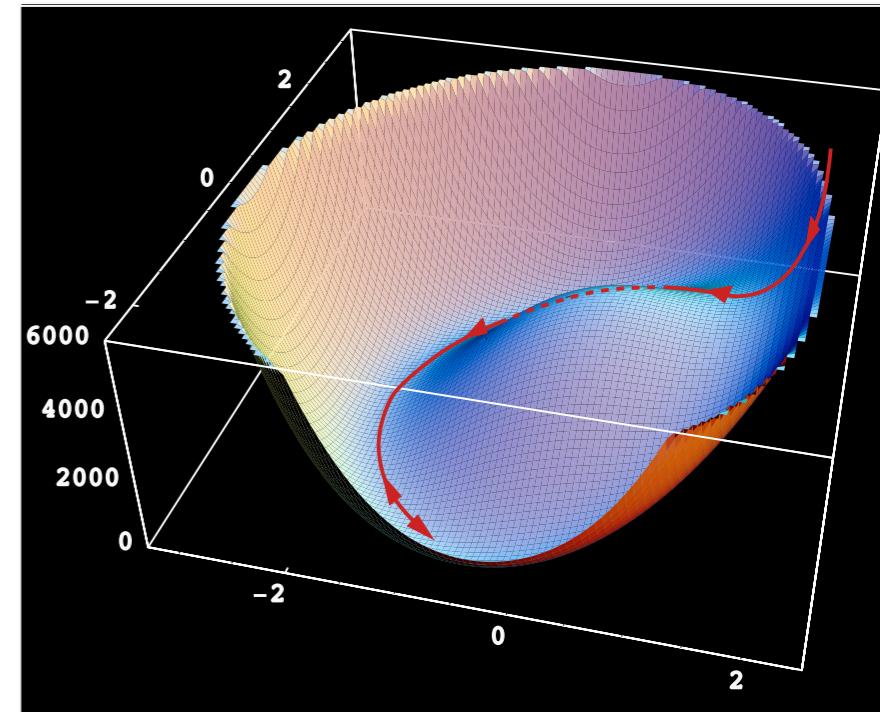
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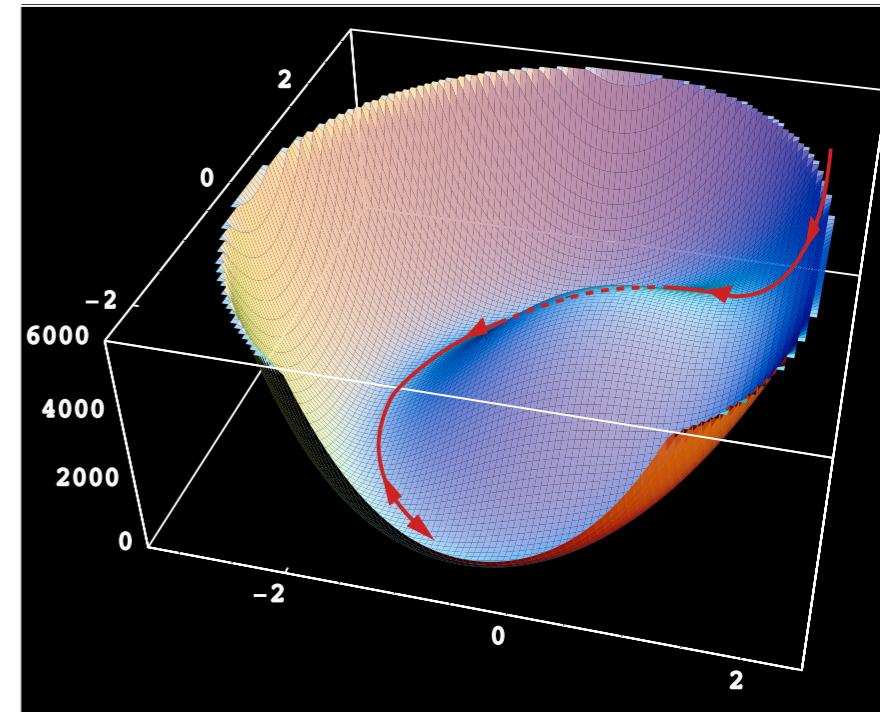
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Large NG from multifield inflation

Curvatons - non-participating, later dominant

e.g. *Linde & Mukhanov 96; Lyth & Wands 01; Moroi & Takahashi 01*

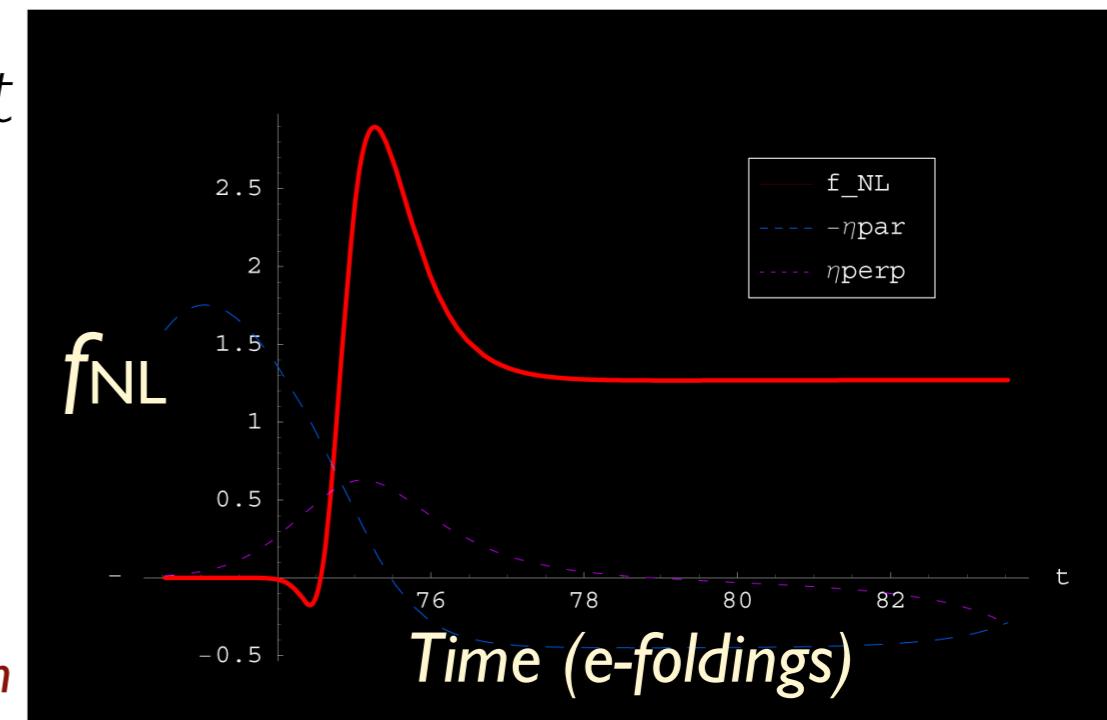
Simple curvatons local, but corrections:

e.g. *Fasiello et al, 11; Lerner et al, 11; Byrnes et al, 11.*

Corner-turning multifield models

Slow-roll: *Byrnes, Choi, Hall, 08a, 08b; see also Seery & Lidsey, 04*

General: *RSvT, 07; Peterson & Tegmark, 10; Meyer & Sivanandam*



Hybrid or waterfall inflation models

Barnaby & Cline, 06; Naruko & Sasaki, 08; Mulryne et al 11.



Non-Gaussian Sources

Multiple field inflation

- Complex corner-turning dynamics during multifield inflation
e.g. Rigopoulos et al 2005; Vernizzi & Wands, 2006; Byrnes, 2010
- End of inflation, reheating and preheating
e.g. Enqvist et al 2005; Rajantie et al, 2008; Kofman et al, 2009;
- Curvatons - post-inflation eqn of state modification
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Higher derivative kinetic terms

- K-inflation, DBI inflation - modified sound speed
e.g. Silverstein & Tong 2003; Alishaha et al 2004; Chen et al 2006

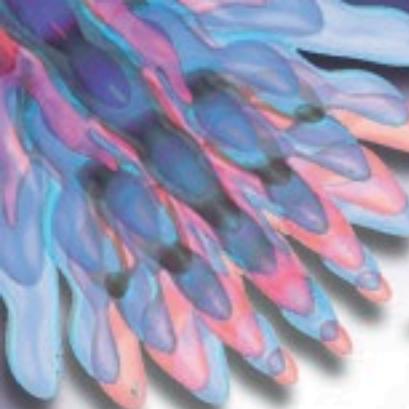
Excited initial states - trans-Planckian effects

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Feature and periodic models (non slow roll)

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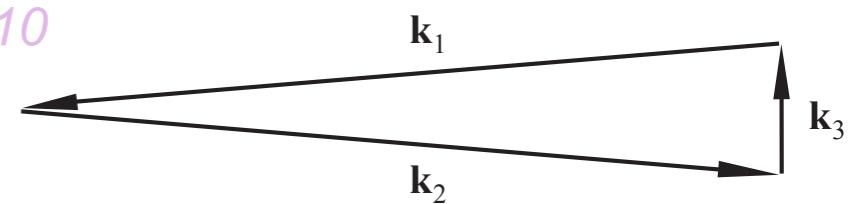
Secondary anisotropies - e.g. ISW, cosmic strings



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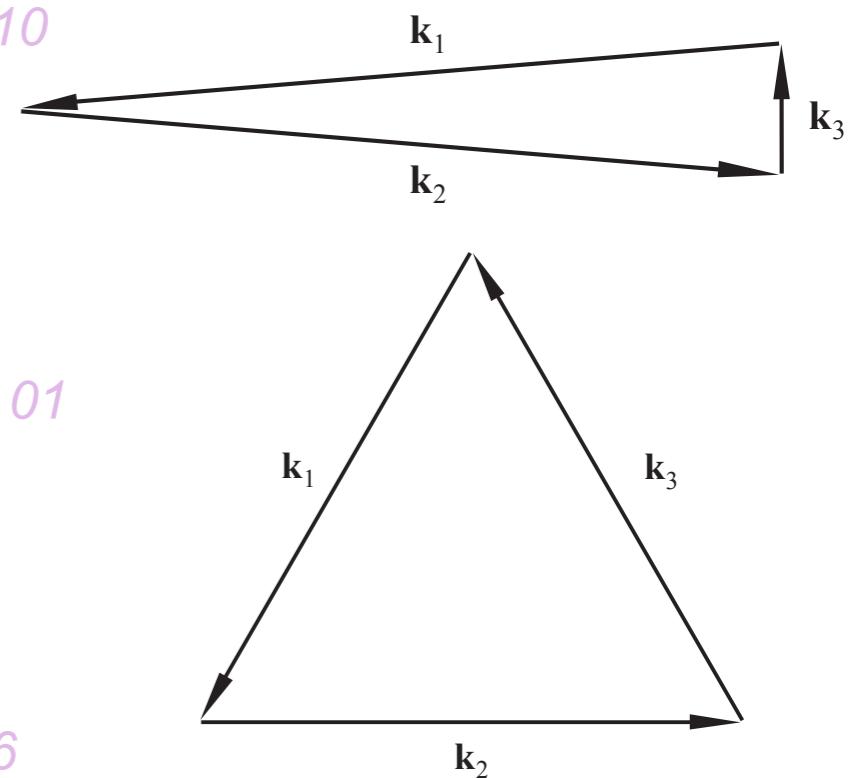
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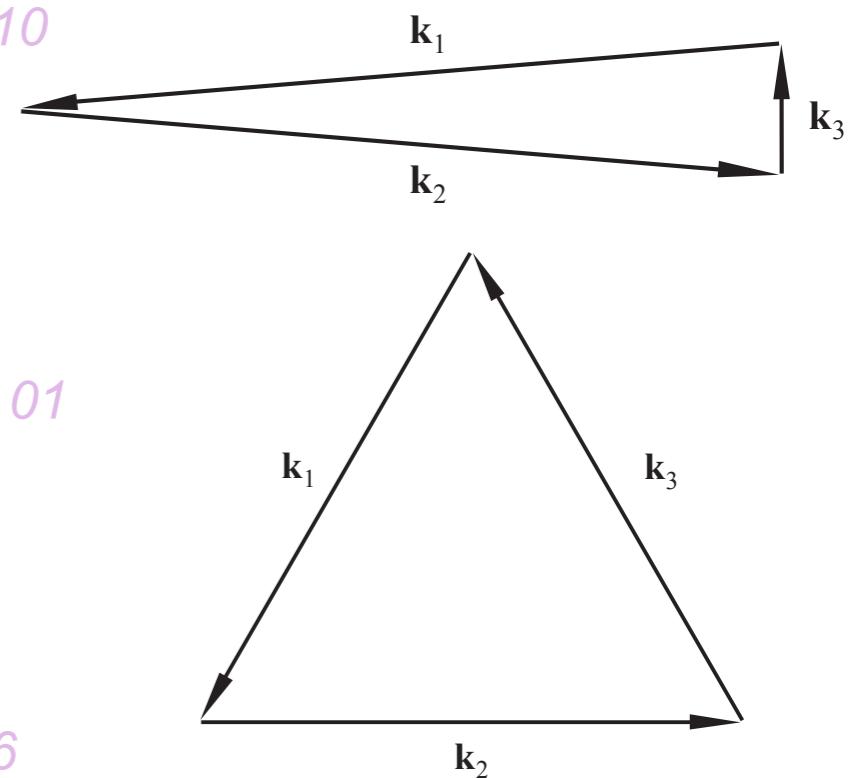
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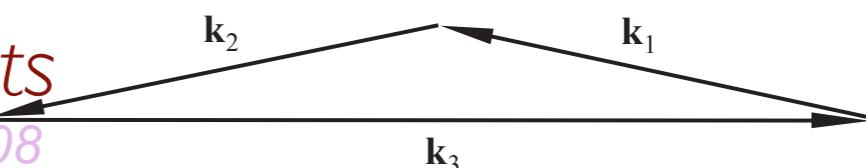


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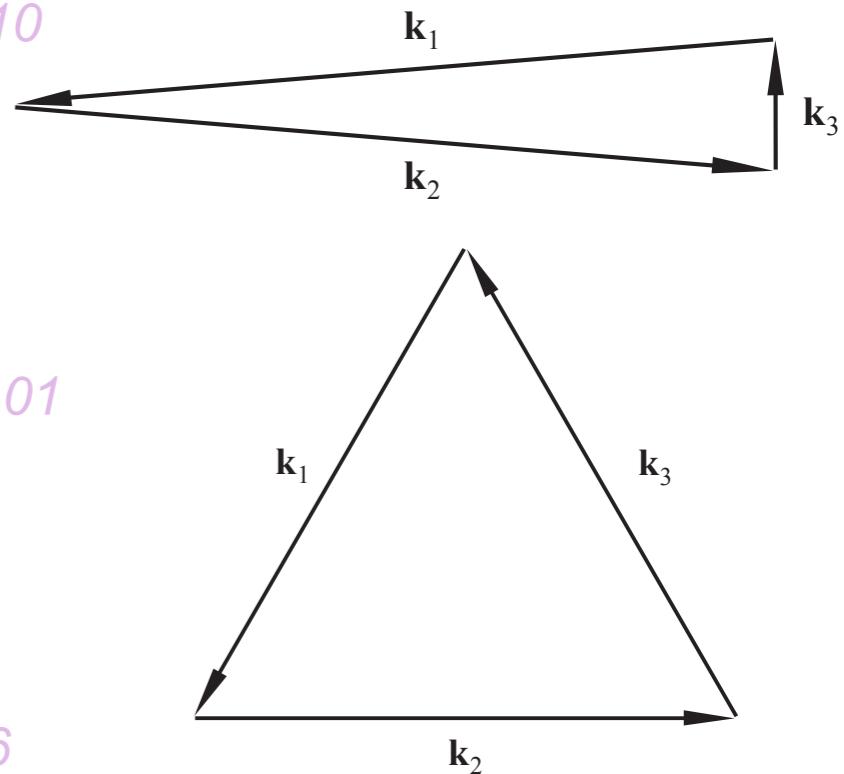
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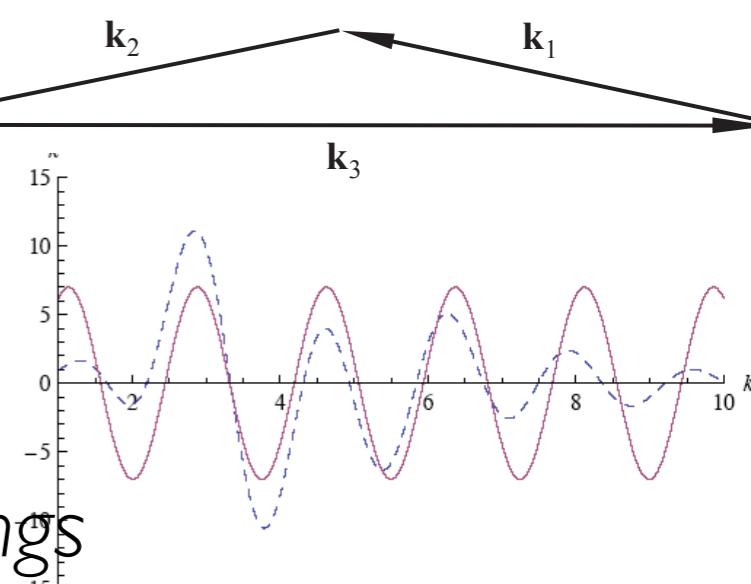
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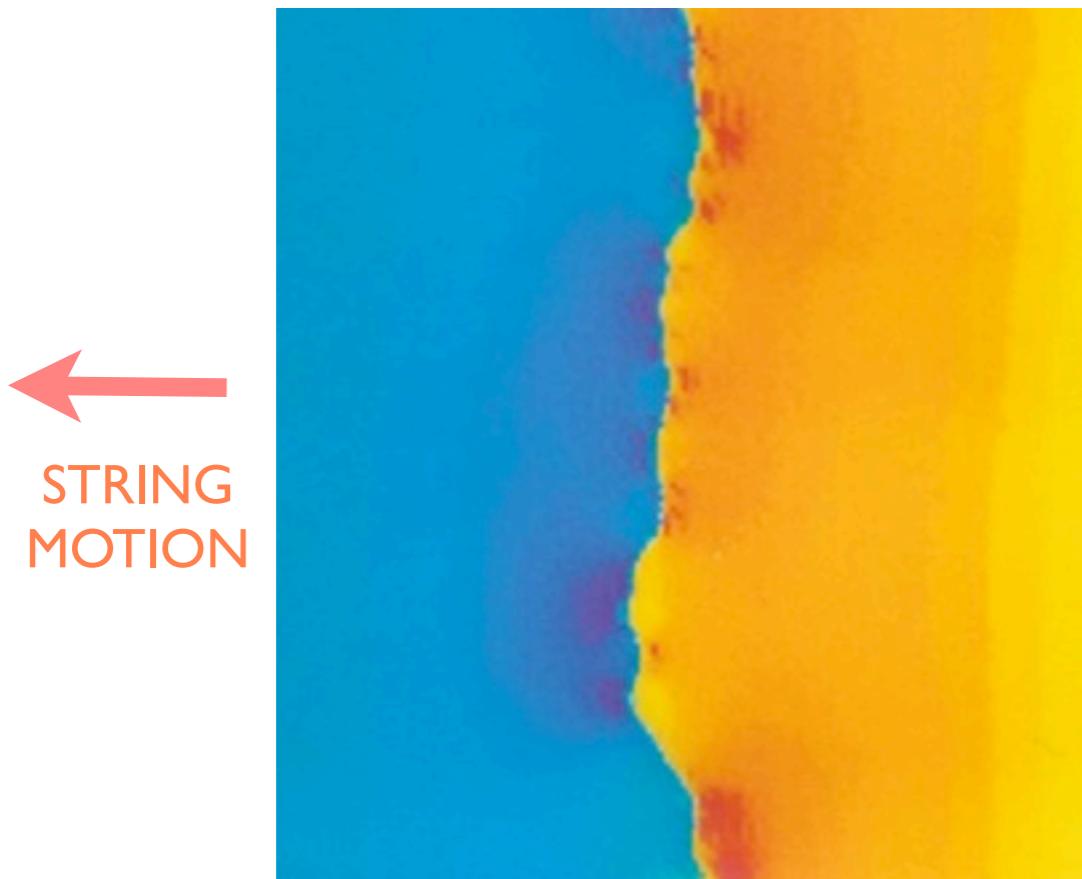


Cosmic (super)strings

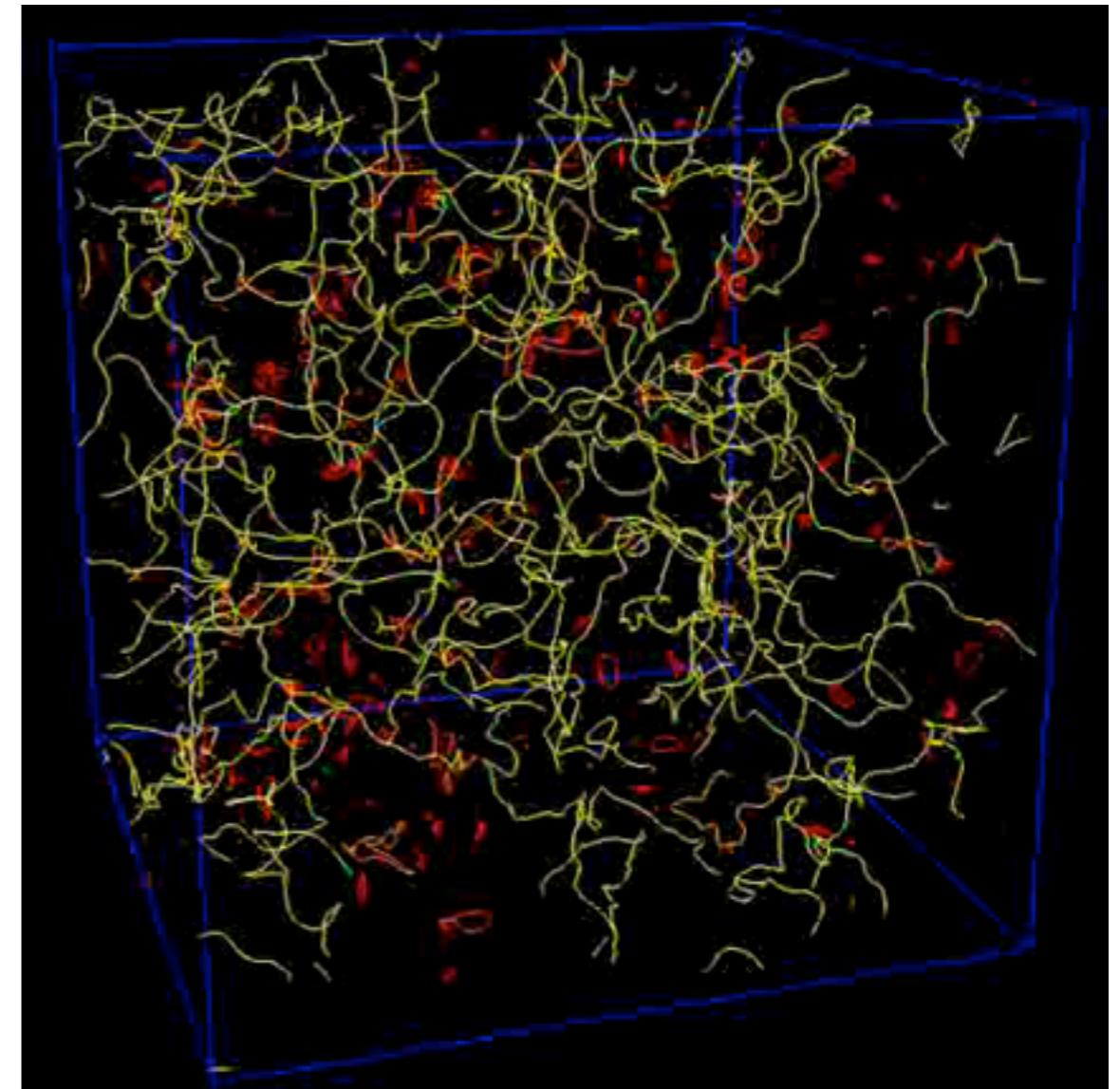
Brane inflation collisions (F- and D-strings) or SUSY GUT models

Scale-invariant evolution

Non-Gaussian CMB signature



Trispectrum key, not bispectrum



Analytic approximations for string bispectrum *Hindmarsh & Ringeval, 09*

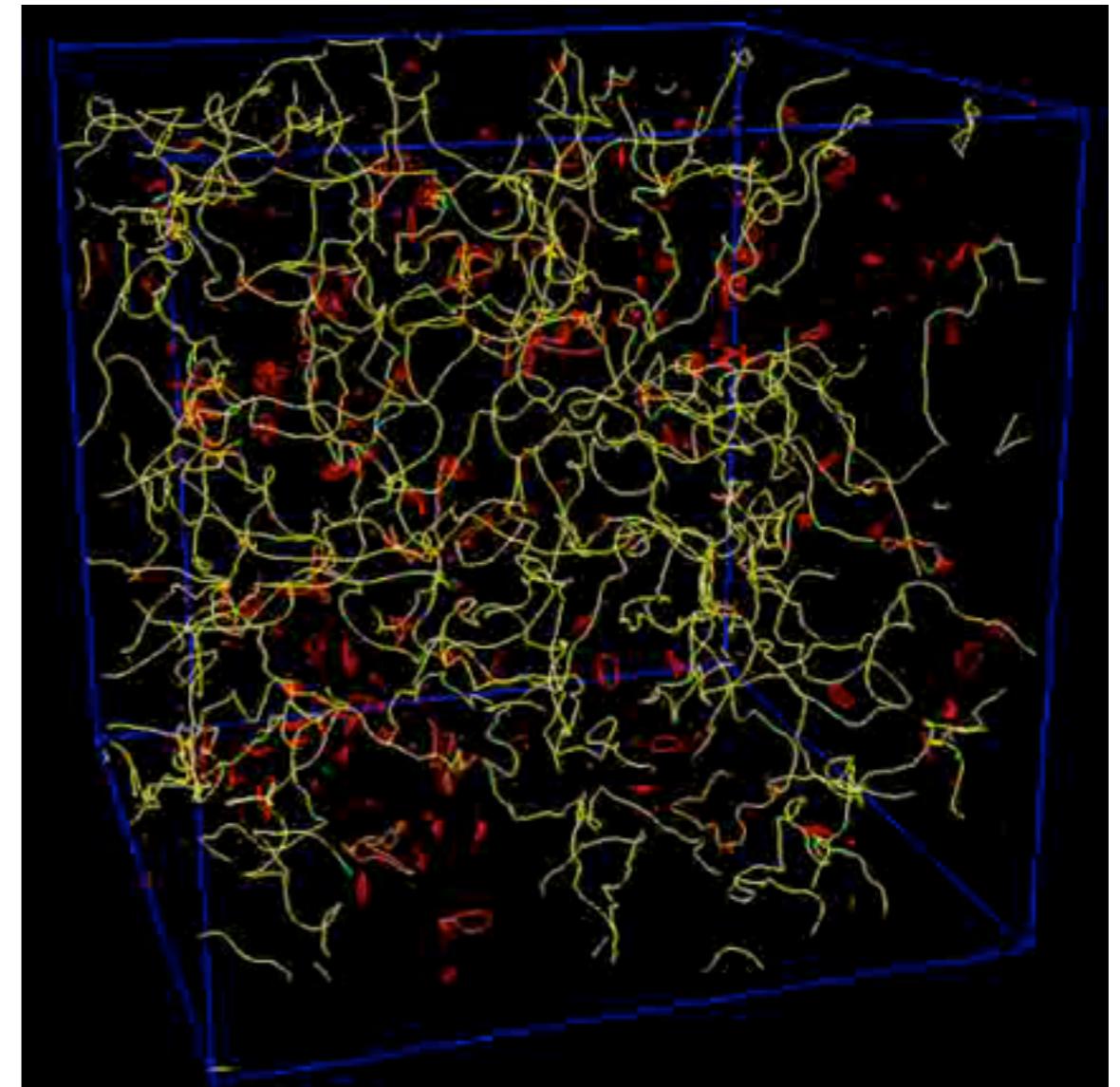
String polyspectra calculated for Planck - T competitive w. $P(k)$ *Regan & EPS, 09*

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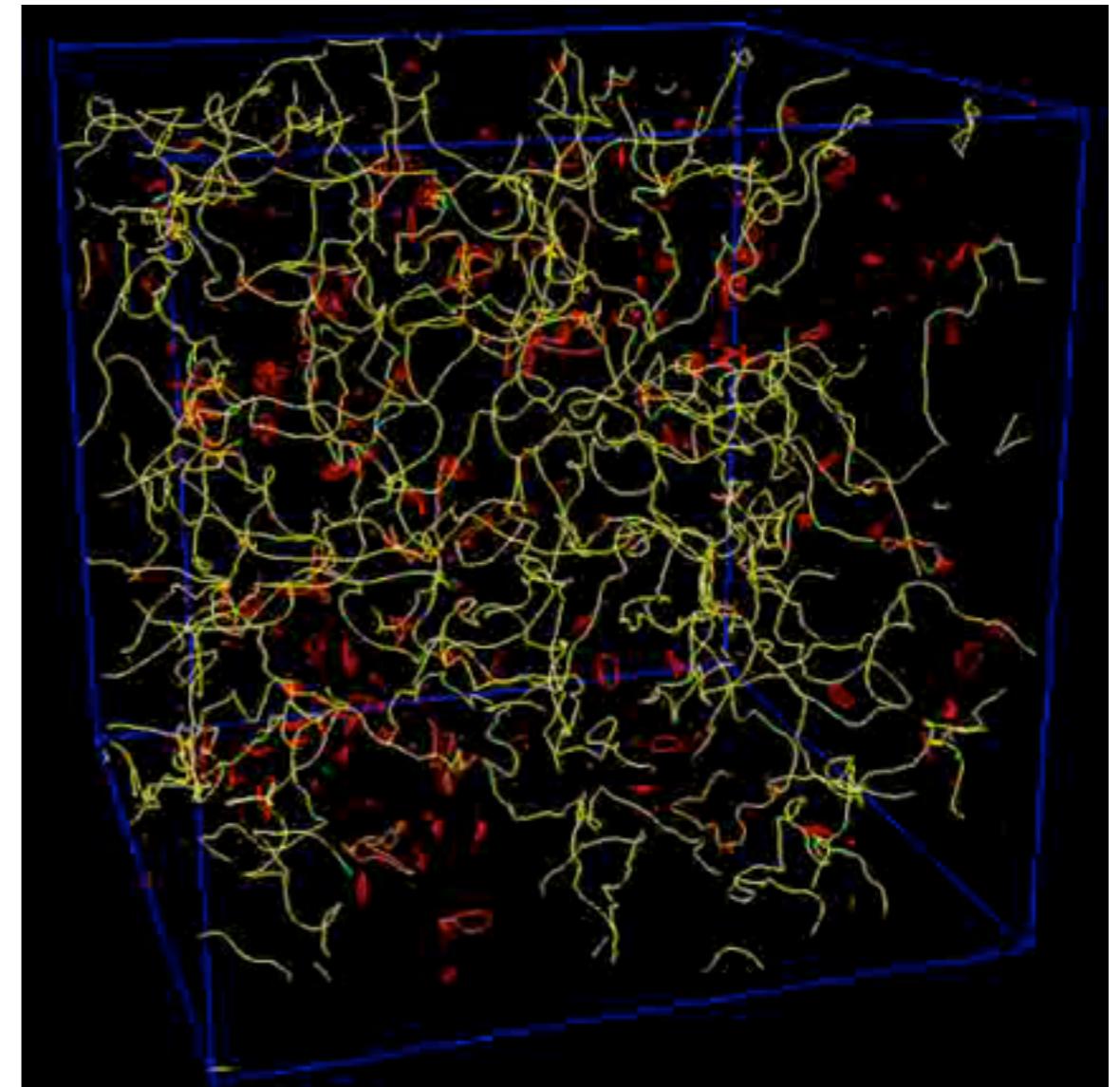
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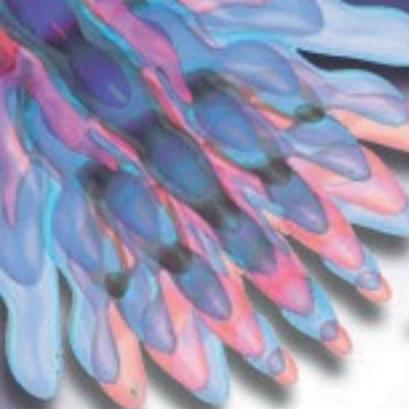
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Primordial & CMB Bispectra

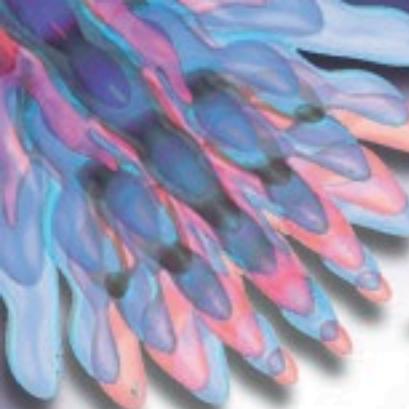
A primordial bispectrum $B(k_1, k_2, k_3) = S(k_1, k_2, k_3)/(k_1 k_2 k_3)^2$ induces

$$b_{l_1 l_2 l_3} = \left(\frac{2}{\pi}\right)^3 \int x^2 dx \int dk_1 dk_2 dk_3 S(k_1, k_2, k_3) \\ \times \Delta_{l_1}(k_1) \Delta_{l_2}(k_2) \Delta_{l_3}(k_3) j_{l_1}(k_1 x) j_{l_2}(k_2 x) j_{l_3}(k_3 x)$$

For the CMB, the bispectrum and trispectrum* are defined by

$$\langle a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3} \rangle = \left(\int d^2 \hat{n} Y_{l_1 m_1}(\hat{n}) Y_{l_2 m_2}(\hat{n}) Y_{l_3 m_3}(\hat{n}) \right) b_{l_1 l_2 l_3}$$
$$\langle a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3} a_{l_4 m_4} \rangle = \left(\int d^2 \hat{n} Y_{l_1 m_1}(\hat{n}) Y_{l_2 m_2}(\hat{n}) Y_{l_3 m_3}(\hat{n}) Y_{l_4 m_4}(\hat{n}) \right) t_{l_1 l_2 l_3 l_4}$$

* For simplicity we give formulae only for diagonal-free trispectra.



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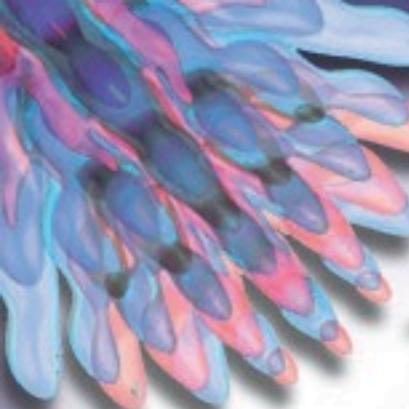
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Primordial shape ←

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$$\langle a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3} \rangle = \left(\int d^2 \hat{n} Y_{l_1 m_1}(\hat{n}) Y_{l_2 m_2}(\hat{n}) Y_{l_3 m_3}(\hat{n}) \right) b_{l_1 l_2 l_3}$$
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* For simplicity we give formulae only for diagonal-free trispectra.



Primordial & CMB Bispectra

A primordial bispectrum $B(k_1, k_2, k_3) = S(k_1, k_2, k_3)/(k_1 k_2 k_3)^2$ induces

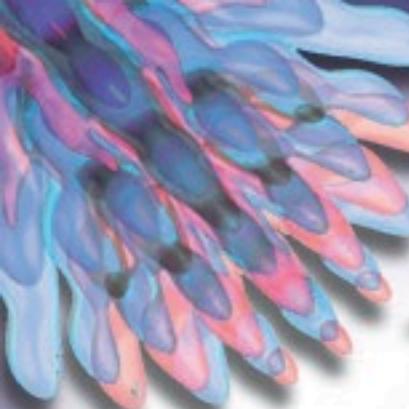
$$b_{l_1 l_2 l_3} = \left(\frac{2}{\pi}\right)^3 \int x^2 dx \int dk_1 dk_2 dk_3 S(k_1, k_2, k_3)$$
$$\times \Delta_{l_1}(k_1) \Delta_{l_2}(k_2) \Delta_{l_3}(k_3) j_{l_1}(k_1 x) j_{l_2}(k_2 x) j_{l_3}(k_3 x)$$

Primordial shape ←
Transfer functions →

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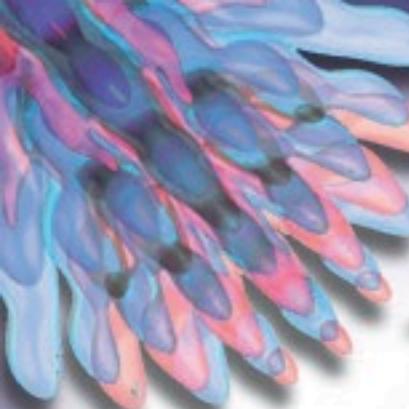
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Primordial shape ←
Transfer functions ↑ ↑ *Oscillatory*

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↑ $\mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3}$

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Exercise II: CMB Local non-Gaussianity

The reduced CMB bispectrum arises from the primordial bispectrum

$$b_{l_1 l_2 l_3} = \left(\frac{2}{\pi}\right)^3 \int dx dk_1 dk_2 dk_3 (x k_1 k_2 k_3)^2 B(k_1, k_2, k_3) \Delta_{l_1}(k_1) \Delta_{l_2}(k_2) \Delta_{l_3}(k_3) j_{l_1}(k_1 x) j_{l_2}(k_2 x) j_{l_3}(k_3 x)$$

In the large-angle approximation ($l \ll 200$), we can approximate the transfer functions by $\Delta_l(k) = \frac{1}{3} j_l(\Delta\tau k)$ with $\Delta\tau = \tau_0 - \tau_{\text{dec}}$.

In the local model with $P(k) \propto k^{-3}$, show that the 4D bispectrum integral above becomes separable yielding the large-angle analytic result

$$b_{l_1 l_2 l_3} \propto \frac{1}{27\pi^2} \left(\frac{1}{l_1(l_1+1)l_2(l_2+1)} + \frac{1}{l_2(l_2+1)l_3(l_3+1)} + \frac{1}{l_3(l_3+1)l_1(l_1+1)} \right).$$

Hint: Note the results

$$\int dk k^2 j_l(ak) j_l(bk) = \frac{\pi}{2a^2} \delta(a-b)$$

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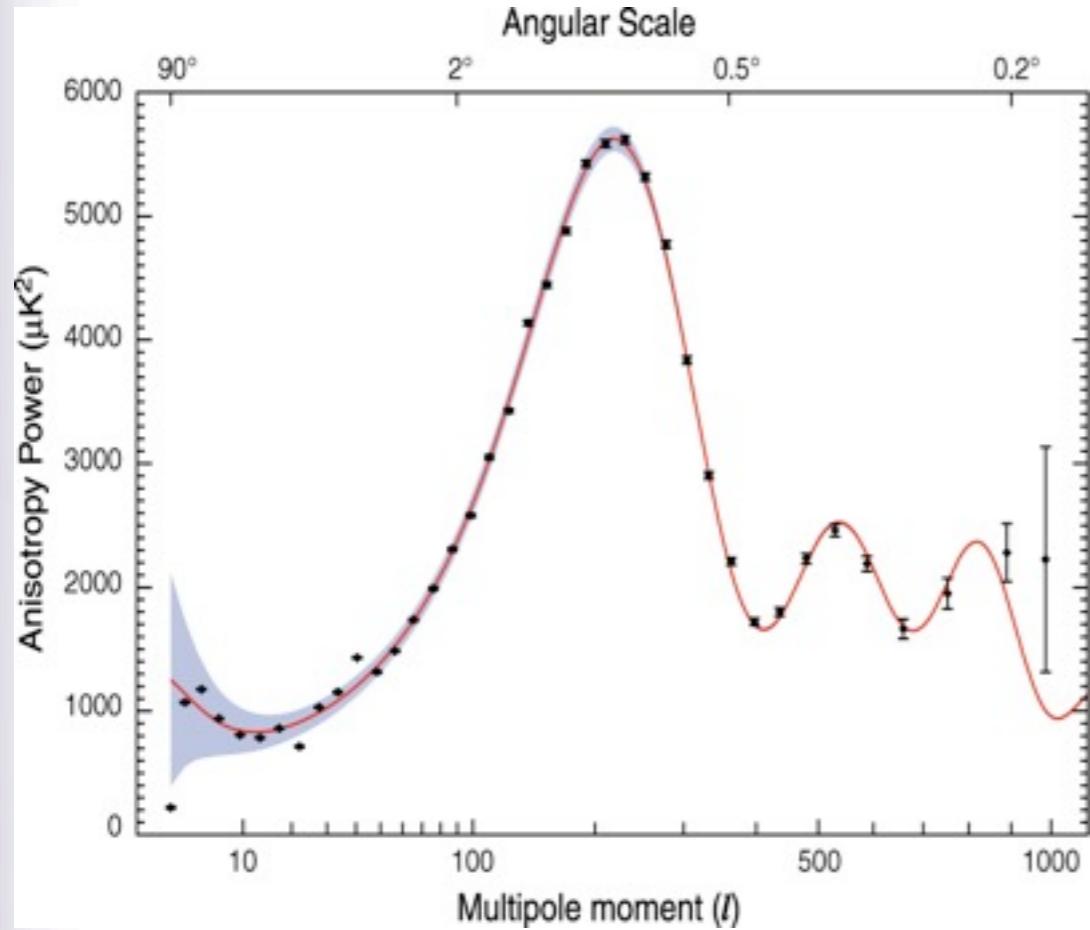
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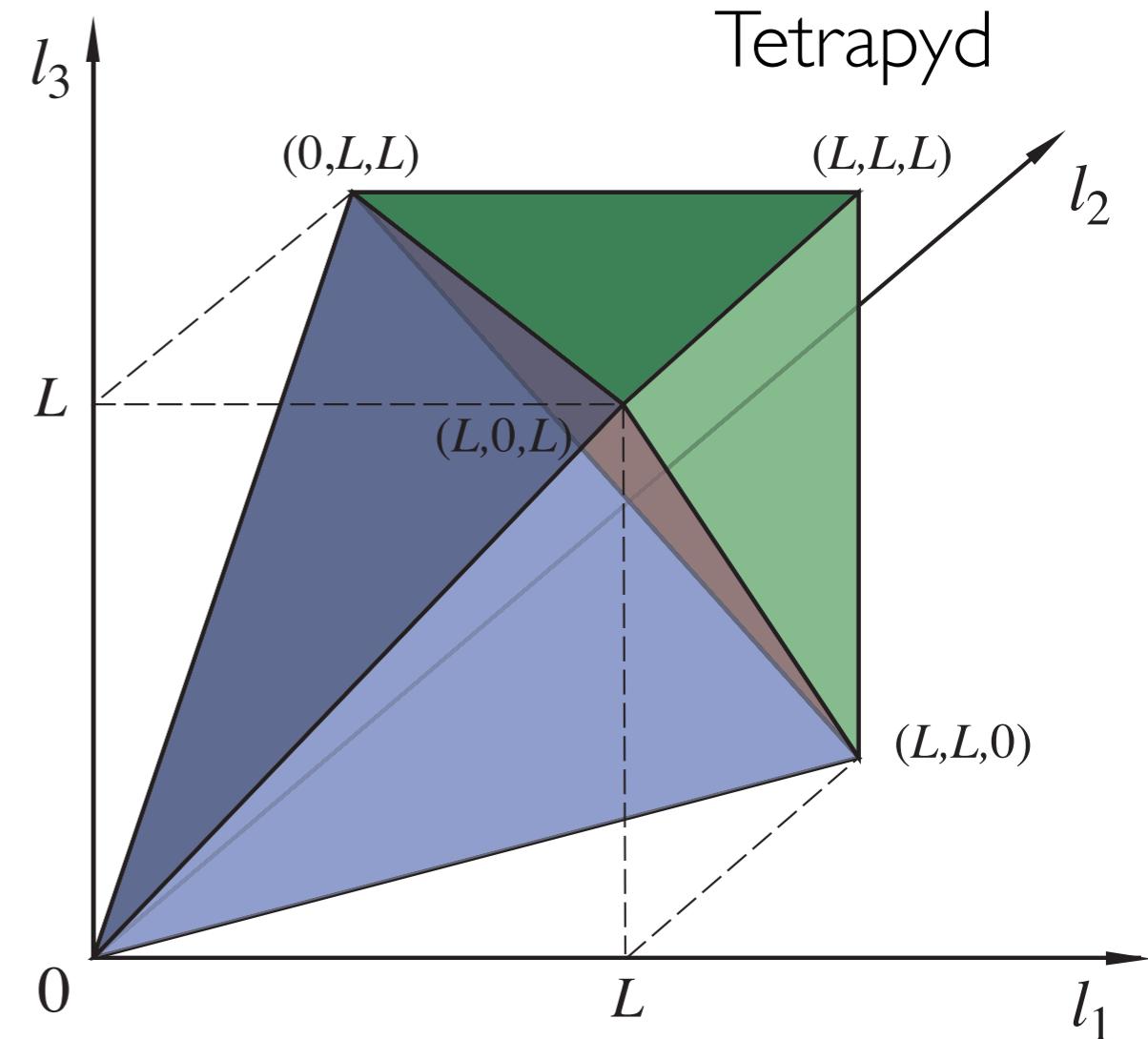
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CMB polyspectra

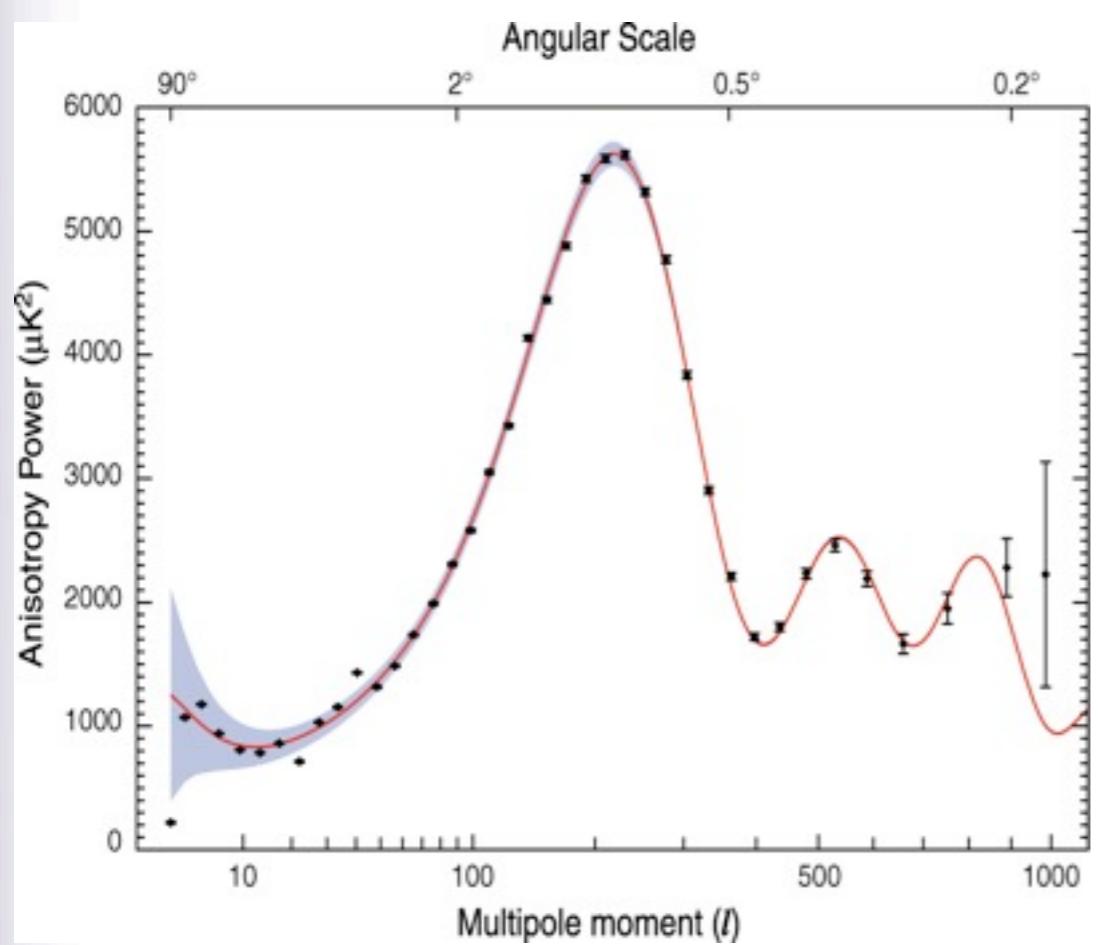


Power spectrum (2pt correlator)

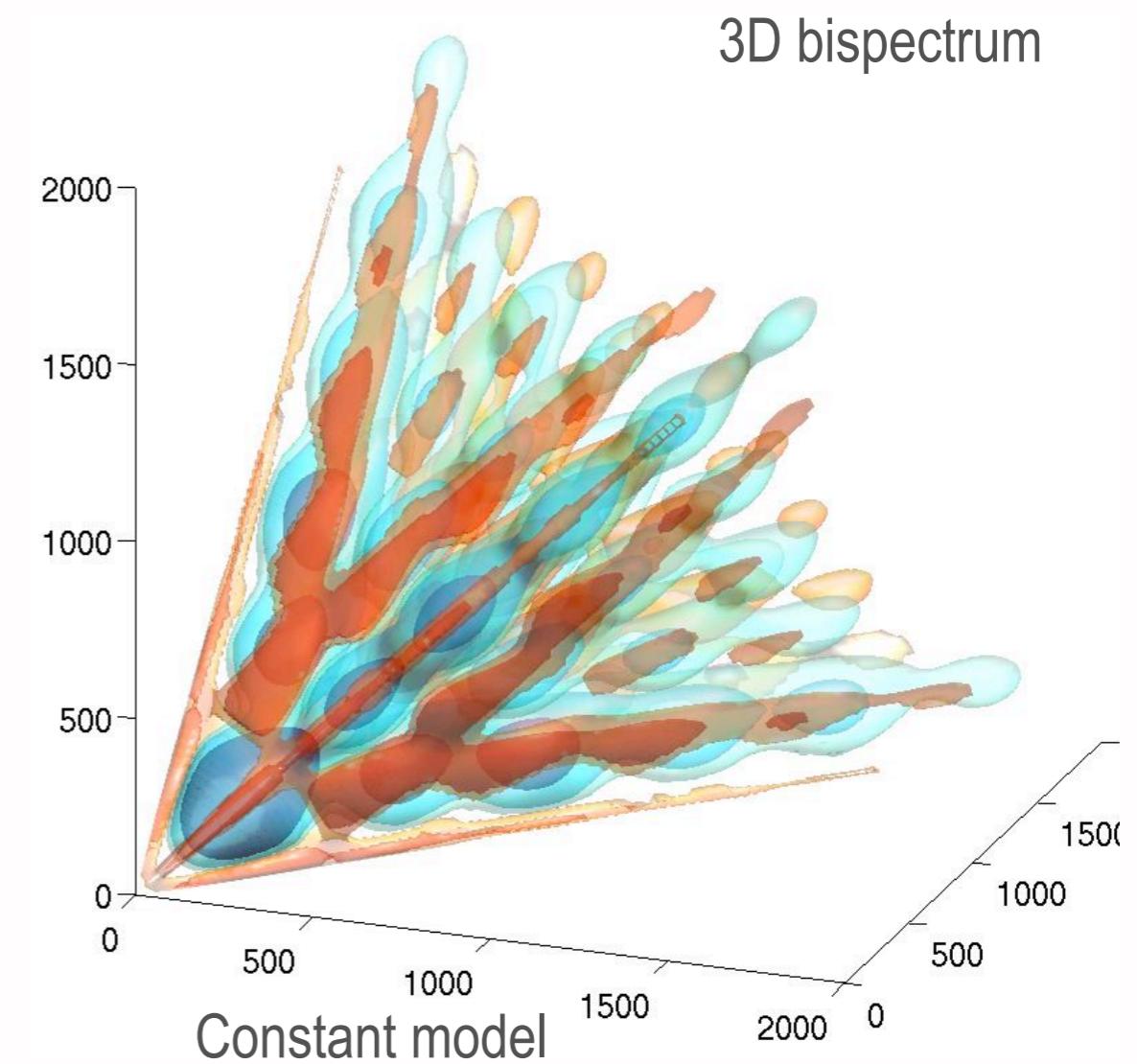


Bispectrum (3pt correlator)

CMB polyspectra

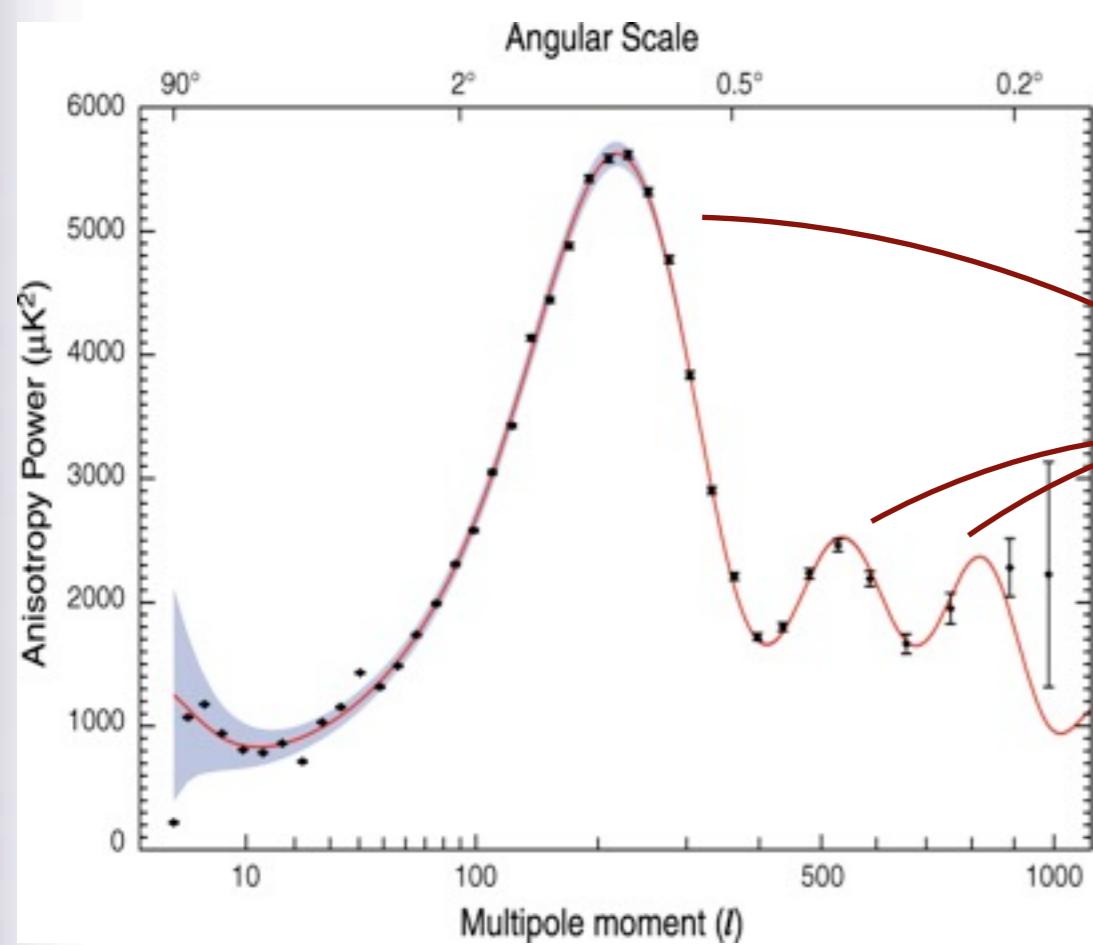


Power spectrum (2pt correlator)

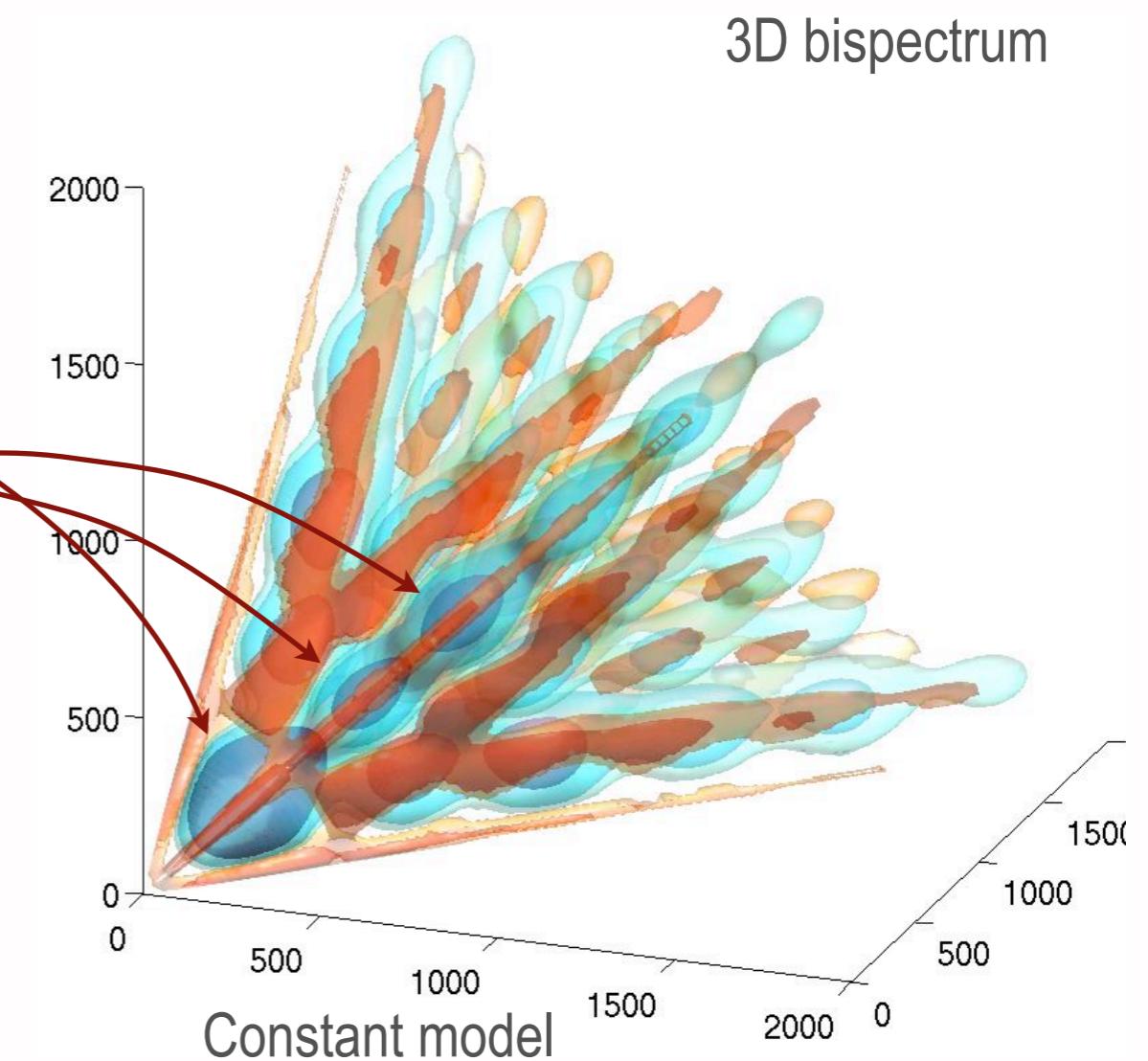


Bispectrum (3pt correlator)

CMB polyspectra

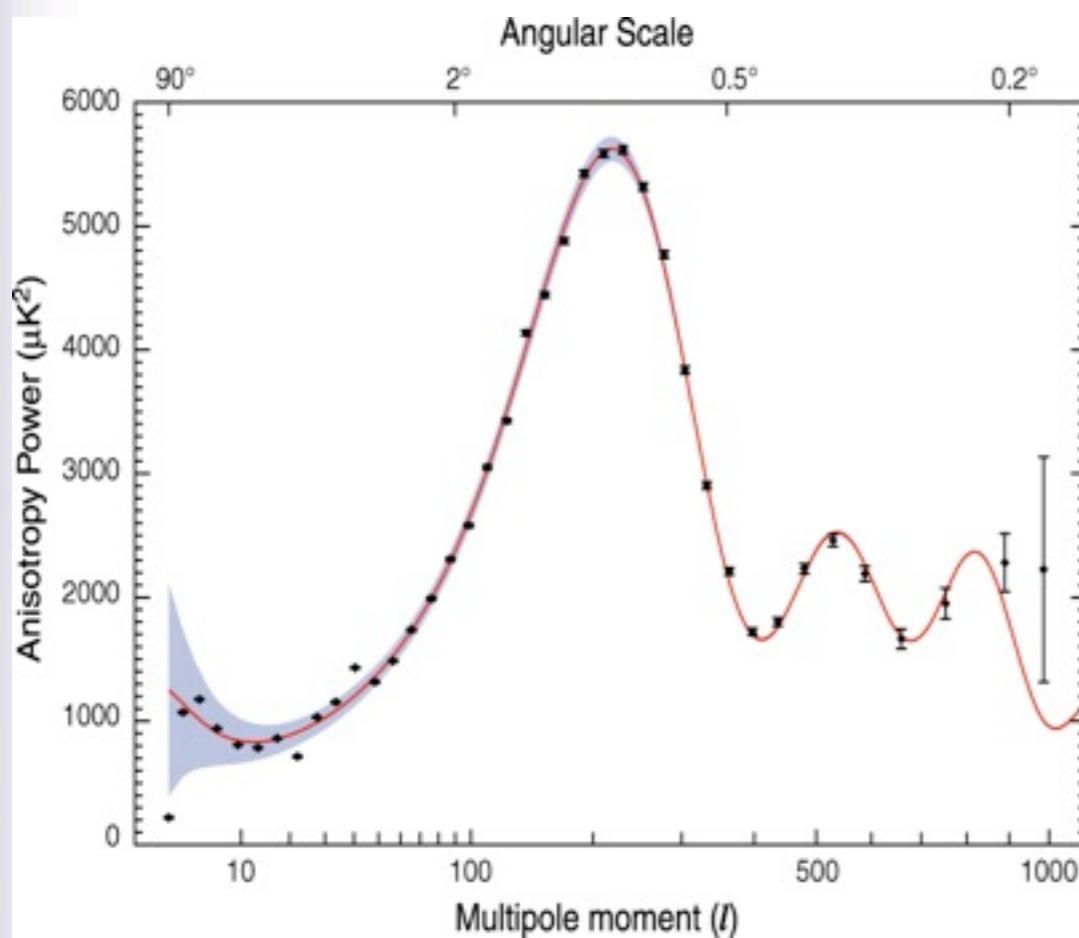


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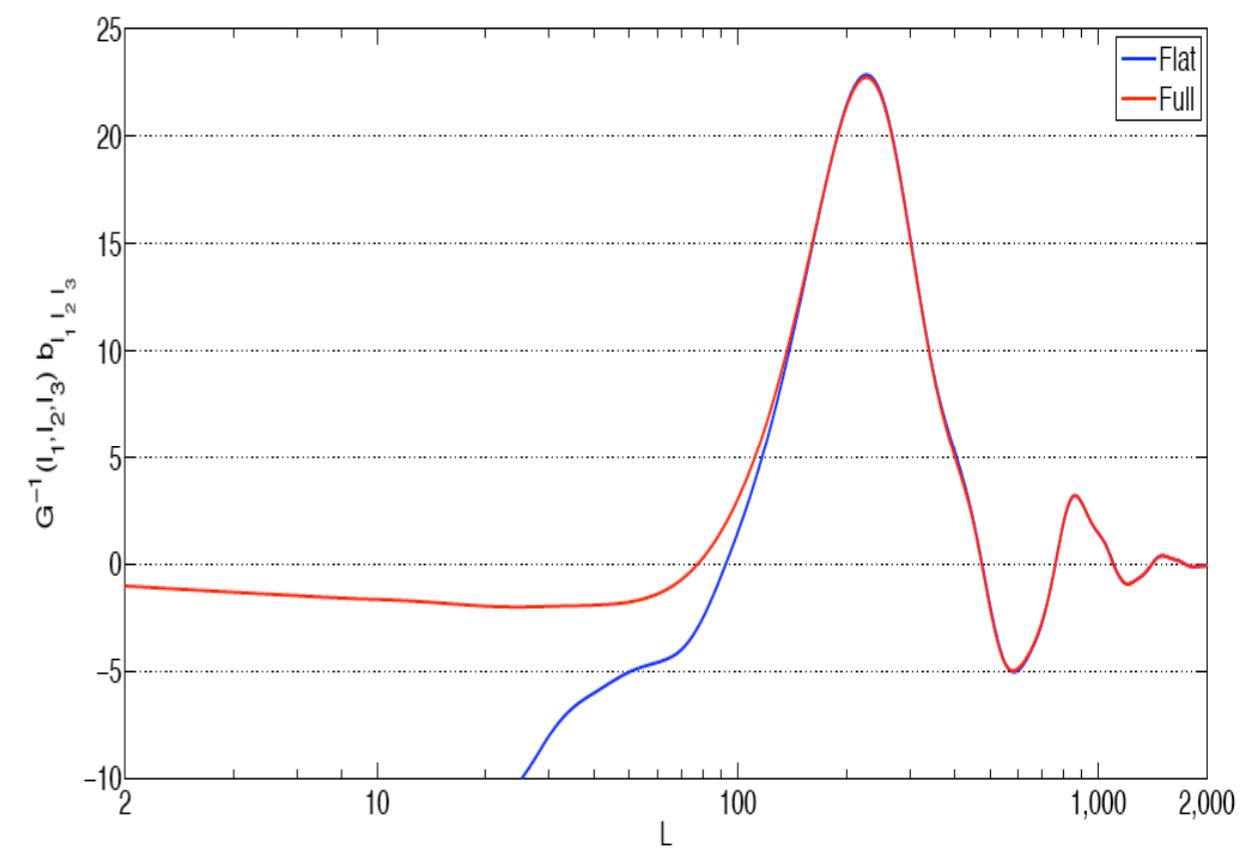


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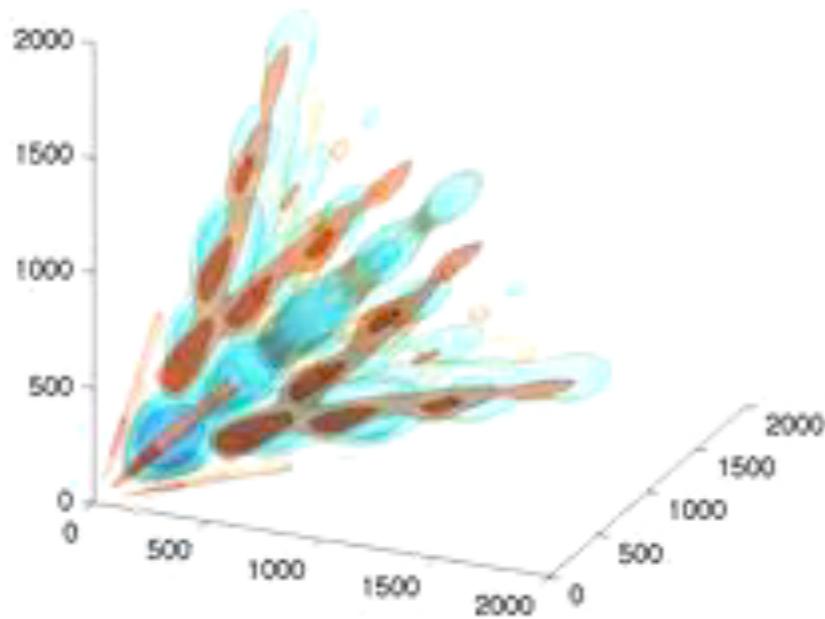
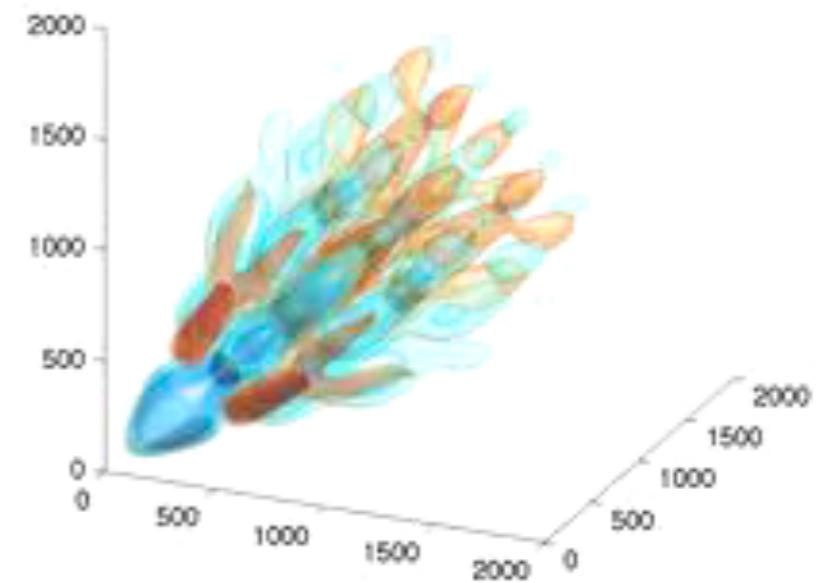
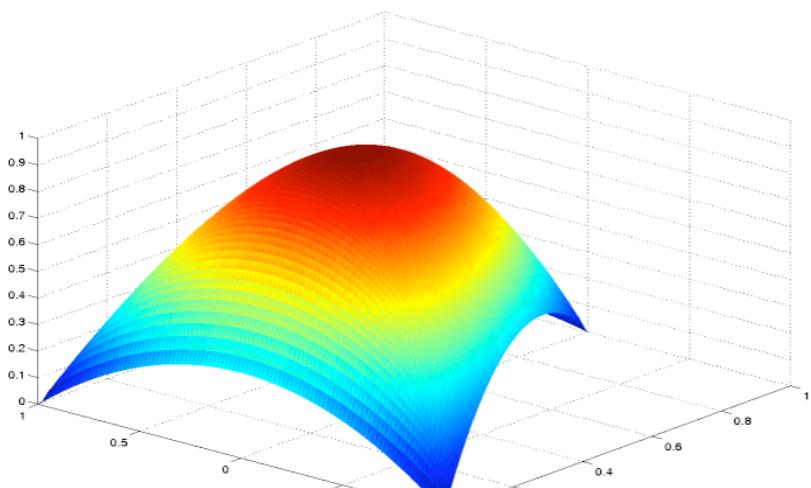


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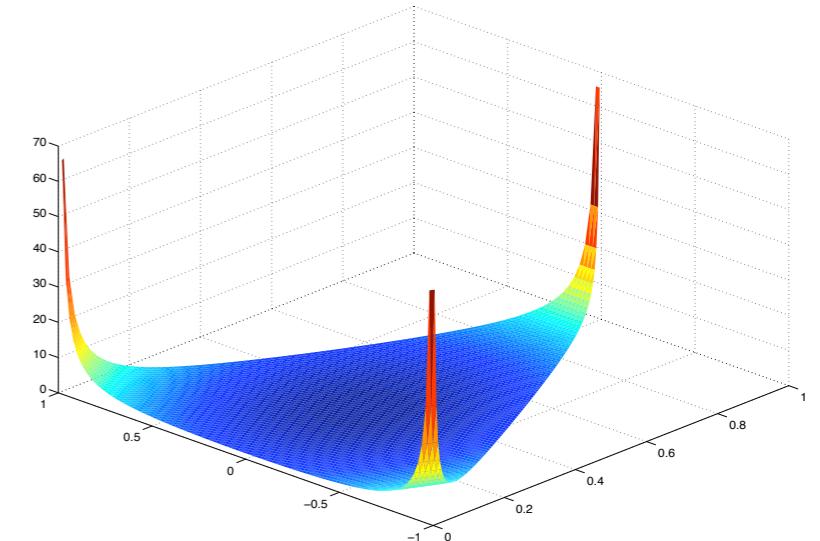


Equal- l B_{lll} Bispectrum

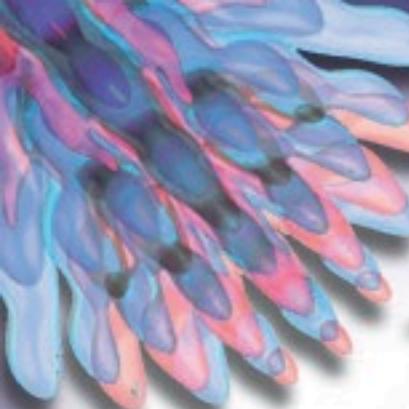
DBI Inflation >>>>>



<<< Multifield Inflation



Robust CMB $b_{\parallel\parallel}$ calculation – see arXiv:1008.1730



Bispectrum estimator

Purpose: Test a model with predicted theoretical bispectrum

$$b_{l_1 l_2 l_3}^{\text{th}} = \sum_{m_i} \mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3} \langle a_{l_1 m_1}^{\text{th}} a_{l_2 m_2}^{\text{th}} a_{l_3 m_3}^{\text{th}} \rangle$$

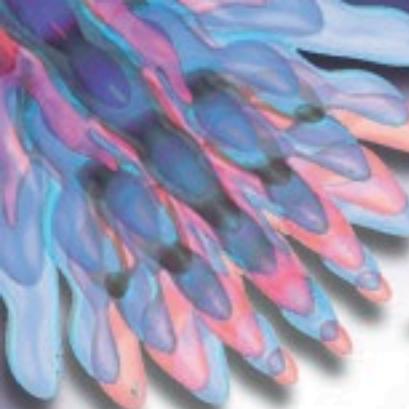
Estimator gives a least squares fit to the data

$$\begin{aligned} \mathcal{E} &= \frac{1}{N^2} \sum_{l_i, m_i} \langle a_{l_1 m_1}^{\text{th}} a_{l_2 m_2}^{\text{th}} a_{l_3 m_3}^{\text{th}} \rangle (C^{-1}a)_{l_1 m_1} (C^{-1}a)_{l_2 m_2} (C^{-1}a)_{l_3 m_3} \\ &= \frac{1}{N^2} \sum_{l_i m_i} \frac{\mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3} b_{l_1 l_2 l_3}^{\text{th}} a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3}}{C_{l_1} C_{l_2} C_{l_3}} \end{aligned}$$

Babich, 2005

with covariance matrix $C_{lm, l'm'} = \langle a_{lm} a_{l'm'} \rangle$

with inverse weighting $(C^{-1}a)_{lm} = C_{lm, l'm'}^{-1} a_{l'm'} \approx \frac{a_{lm}}{C_l}$ (ideal case)



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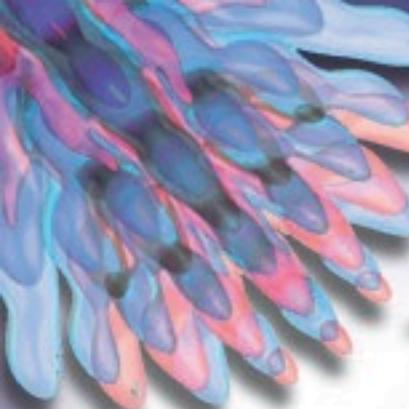
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Babich, 2005

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Model *Signal*
Noise

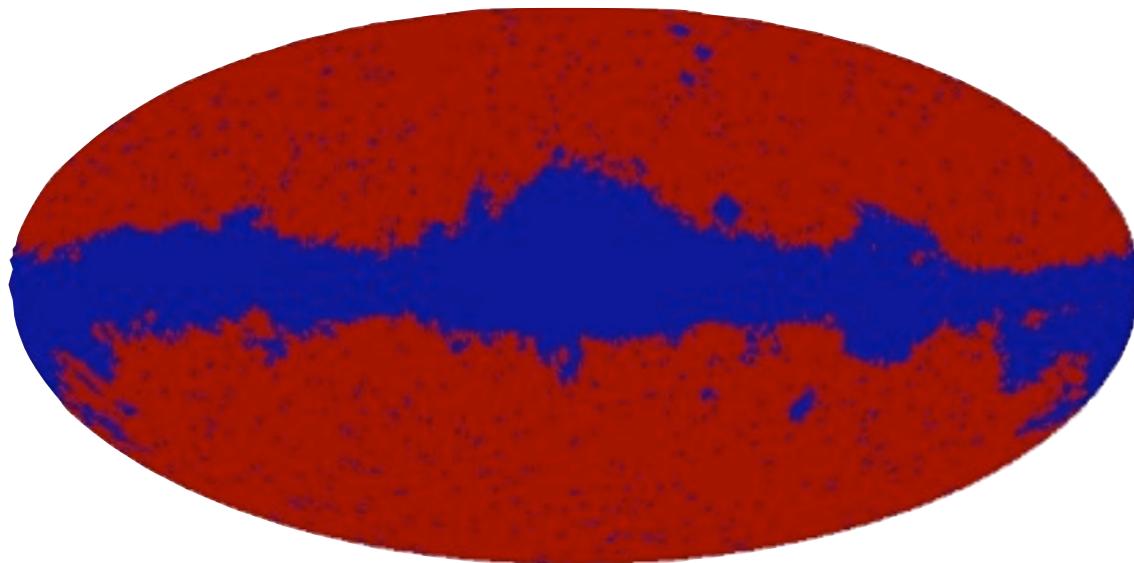
Babich, 2005

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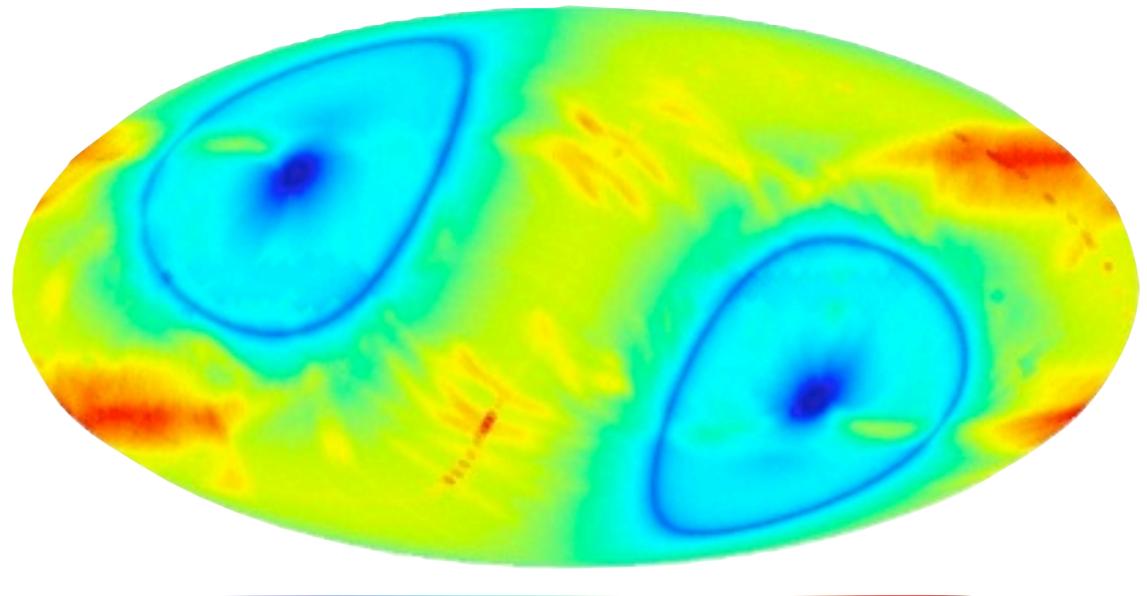
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Realistic CMB estimator

Need to incorporate effects of masks, noise, and beams



WMAP KQ75 Mask



WMAP inhomogeneous noise

Linear term to subtract out spurious contributions *Creminelli et al, 2006*

$$\mathcal{E} = \frac{1}{\tilde{N}^2} \sum_{l_i m_i} \frac{\mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3} \tilde{b}_{l_1 l_2 l_3}^{\text{th}}}{\tilde{C}_{l_1} \tilde{C}_{l_2} \tilde{C}_{l_3}} (a_{l_1 m_1} a_{l_2 m_2} - 6 C_{l_1 m_1, l_2 m_2}^{\text{sim}}) a_{l_3 m_3}$$

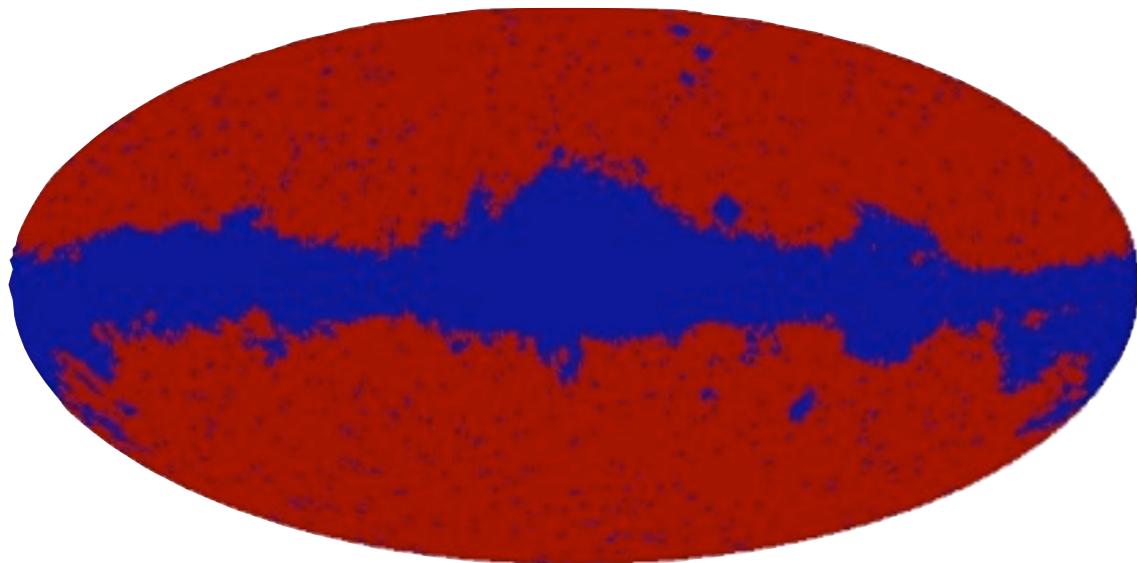
where $C_{l,m}^{\text{sim}}$ is an average over many realistic Gaussian realizations

Include effects of noise, beams and masking through

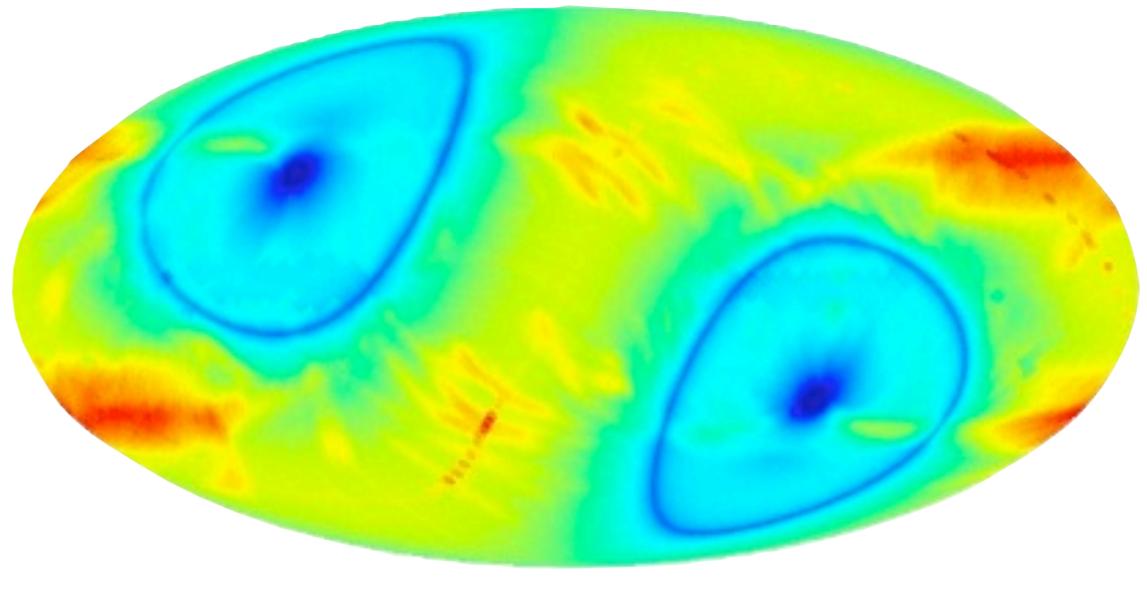
$$\begin{aligned} \tilde{C}_l &= b_l^2 C_l + N_l & \text{and} & \quad \tilde{b}_{l_1 l_2 l_3} = b_{l_1} b_{l_2} b_{l_3} b_{l_1 l_2 l_3} \\ b_{l_1 l_2 l_3}^{\text{mask}} &= f_{\text{sky}} b_{l_1 l_2 l_3} & \text{and} & \quad C_l^{\text{mask}} = f_{\text{sky}} C_l \end{aligned}$$

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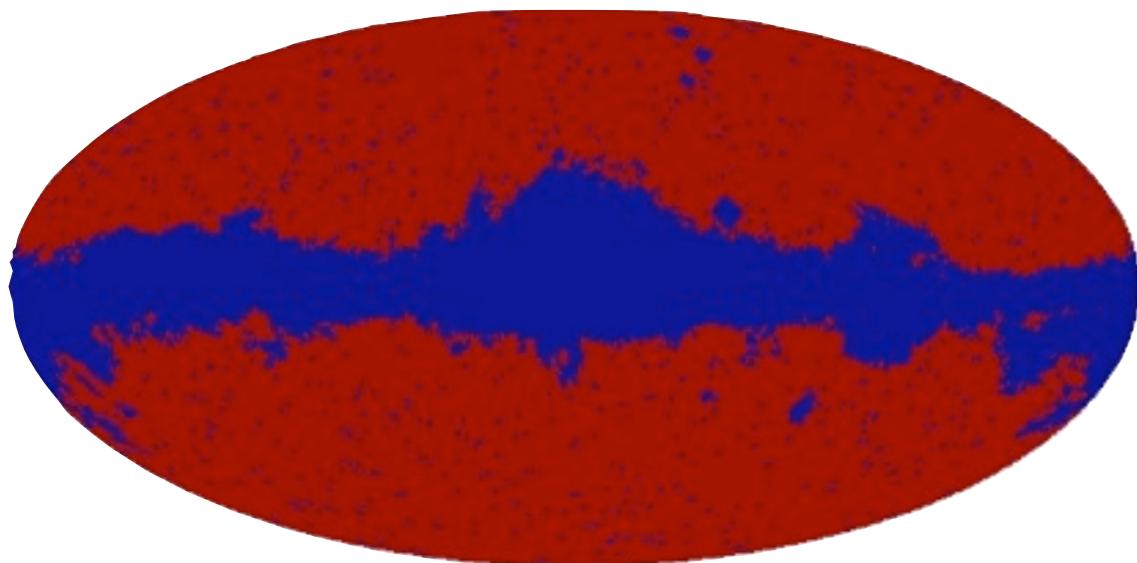
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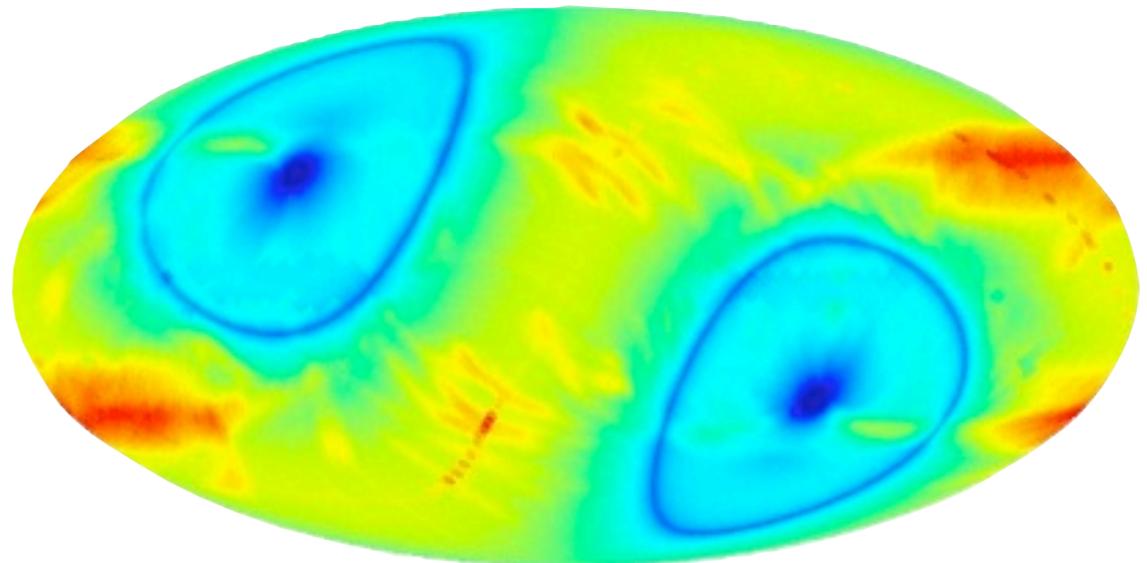
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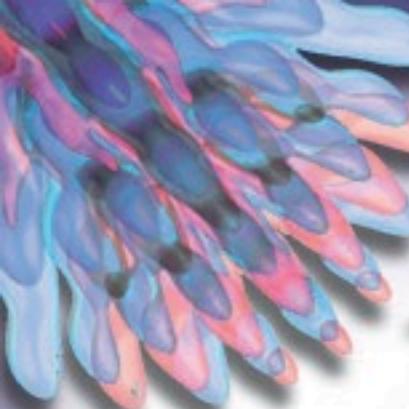
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where $C_{l,m}^{\text{sim}}$ is an average over many realistic Gaussian realizations

“*Linear term*”

Include effects of noise, beams and masking through

$$\begin{aligned} \tilde{C}_l &= b_l^2 C_l + N_l & \text{and} & \quad \tilde{b}_{l_1 l_2 l_3} = b_{l_1} b_{l_2} b_{l_3} b_{l_1 l_2 l_3} \\ b_{l_1 l_2 l_3}^{\text{mask}} &= f_{\text{sky}} b_{l_1 l_2 l_3} & \text{and} & \quad C_l^{\text{mask}} = f_{\text{sky}} C_l \end{aligned}$$



Bispectrum constraints

Computational cost of direct CMB estimation is Operations $\sim 10^3 \times L^5$

Separable primordial models: $B(k_1, k_2, k_3) = X(k_1)Y(k_2)Z(k_3) + \text{perms.}$
these require only Operations $\sim 10^3 \times L^2$ ($L=l_{\max}$)

TRACTABILITY = SEPARABILITY!

Example: Local model where $B(k_1, k_2, k_3) = P(k_1)P(k_2) + \text{perms.}$

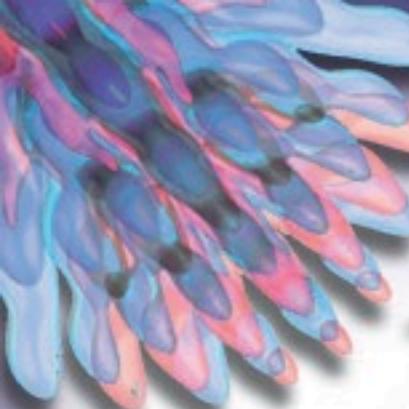
Recall the 4D bispectrum integral

$$b_{l_1 l_2 l_3} = \left(\frac{2}{\pi}\right)^3 \int x^2 dx \int dk_1 dk_2 dk_3 S(k_1, k_2, k_3) \\ \times \Delta_{l_1}(k_1) \Delta_{l_2}(k_2) \Delta_{l_3}(k_3) j_{l_1}(k_1 x) j_{l_2}(k_2 x) j_{l_3}(k_3 x)$$

which reduces to

$$b_{l_1 l_2 l_3} = \int x^2 dx A_{l_1}(x) A_{l_2}(x) B_{l_3}(x) + \text{perms.} \quad A_l(x) = \int k^2 dk P(k) \Delta_l(k) j_l(kx) \\ B_l(x) = \int k^2 dk \Delta_l(k) j_l(kx)$$

Limited constraints achieved - WMAP team only local, equilateral and orthogonal.
e.g. local model $-10 < f_{NL} < 74$ (Komatsu et al, 2010)



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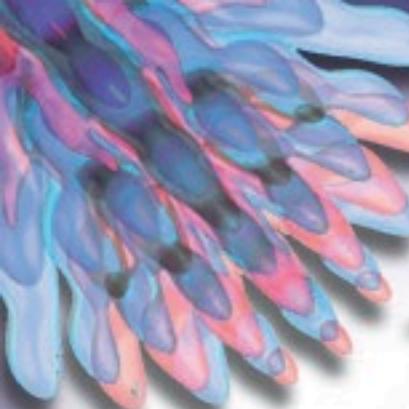
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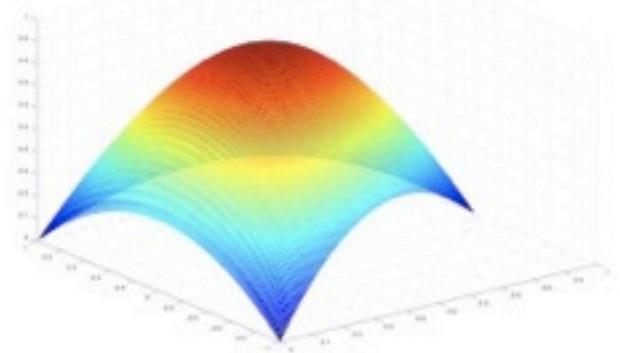
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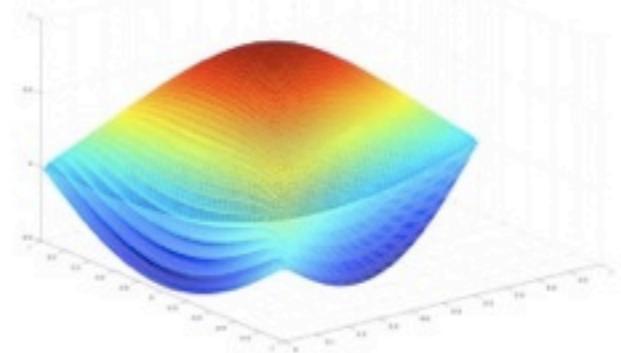
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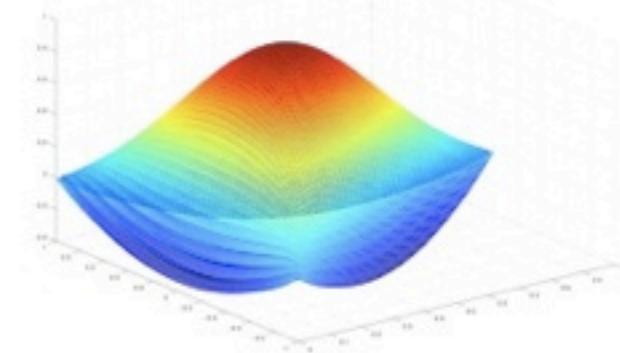
Three models down, ∞ to go



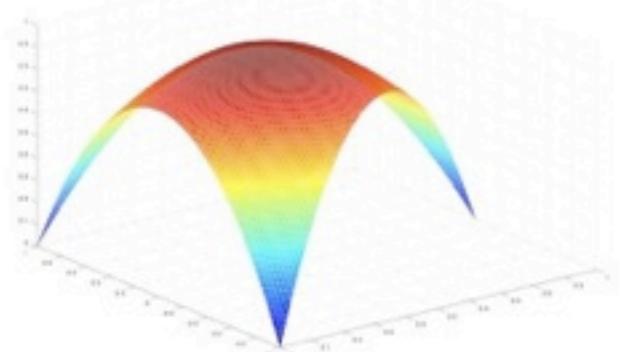
DBI



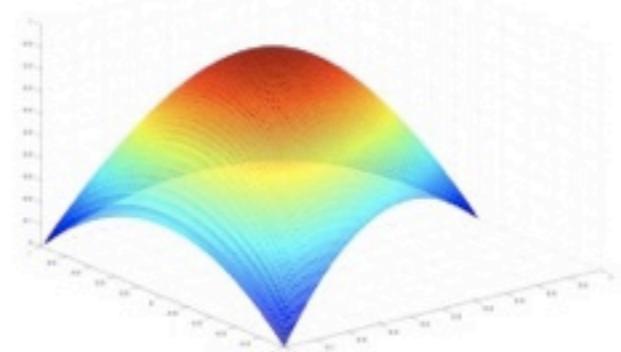
DBI with angular momentum



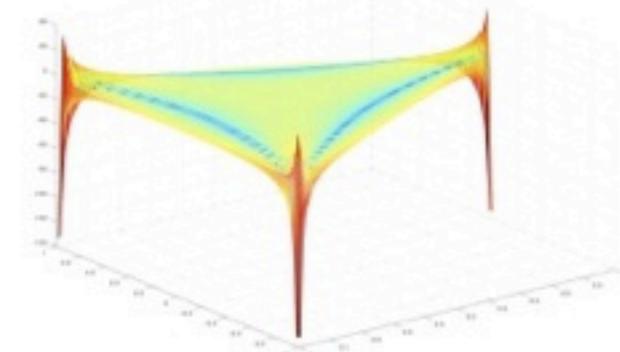
Ghost



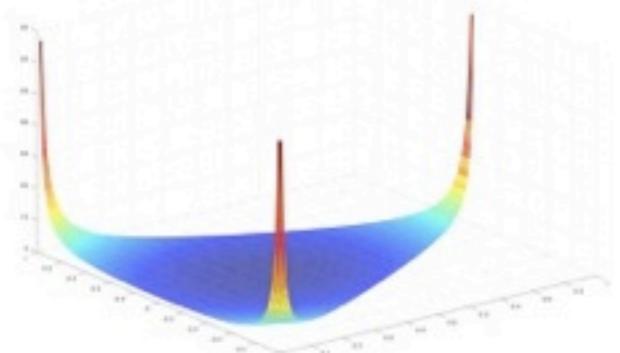
Single Field 1



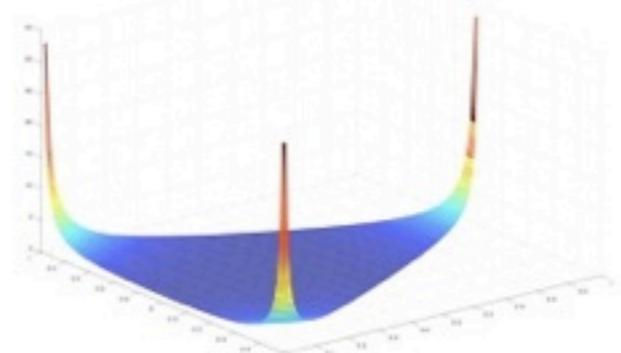
Single Field 2



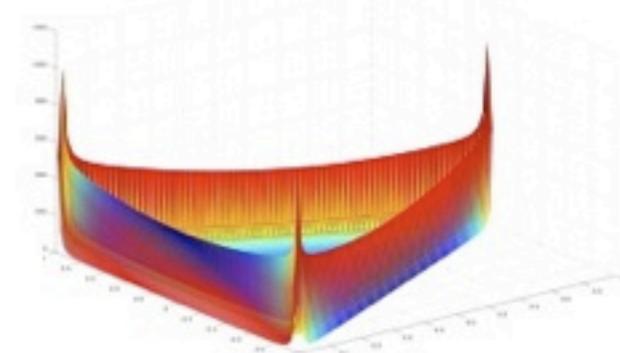
Warm



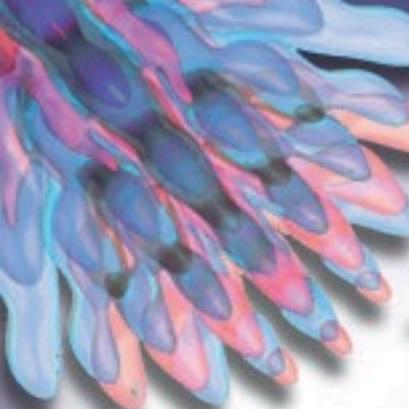
Non-Local



Exited Initial States 1



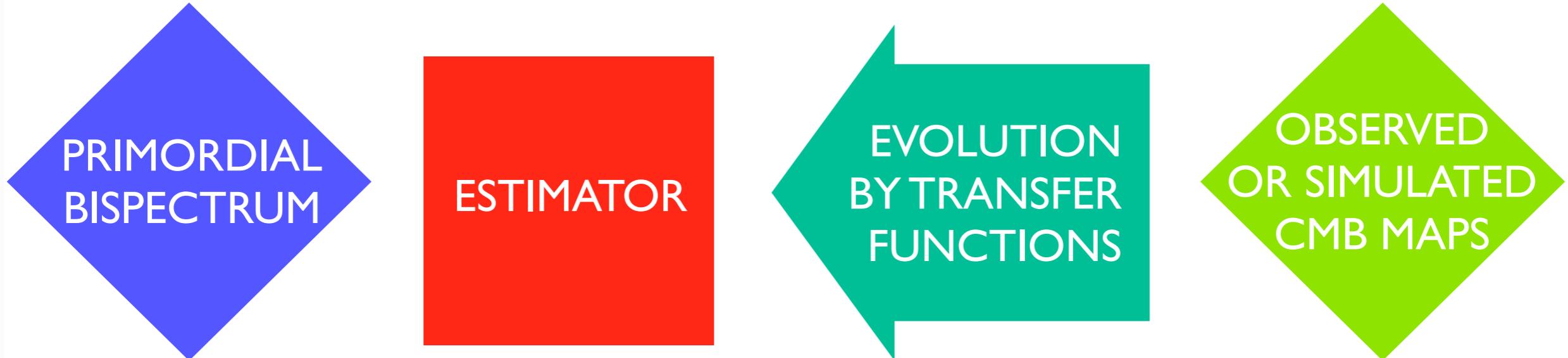
Exited Initial States 2



General Modal Estimator

[astro-ph/0612713](#), arXiv:1012.6039

Current estimators (e.g. KSW) extract primordial signals from the CMB



I. Late-time estimators filter for polyspectra in the CMB map

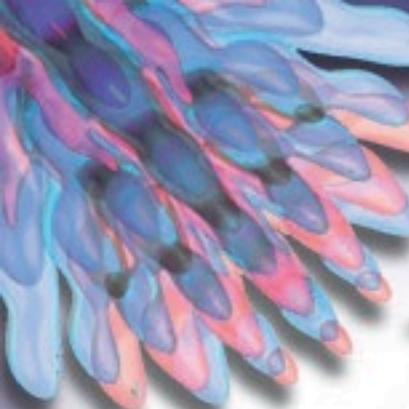
Efficient: Transfer functions evaluated only once for primordial theory

Flexible: Can seek late-time contributions (e.g. strings) / contaminants

2. Modal expansion allows efficient separable study of ***all bispectra***

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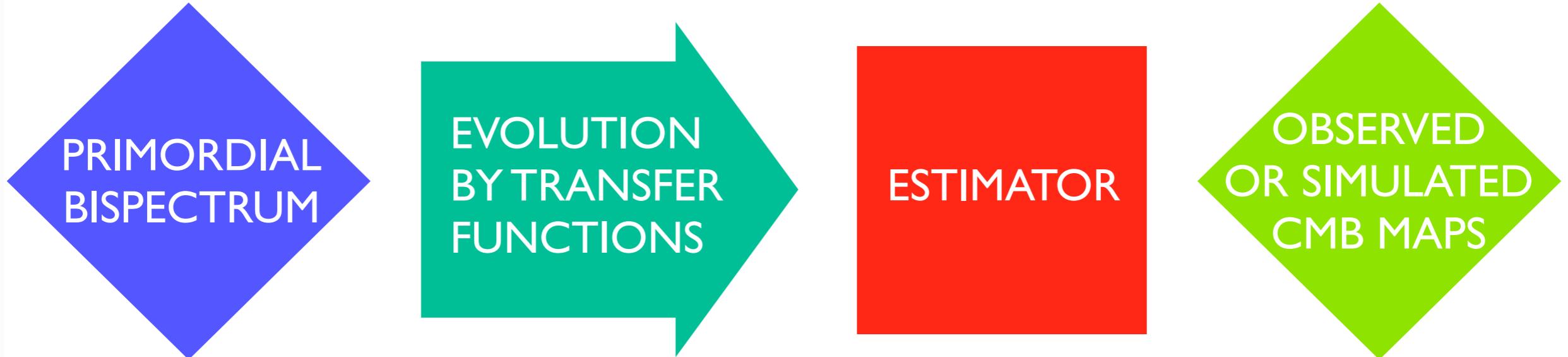
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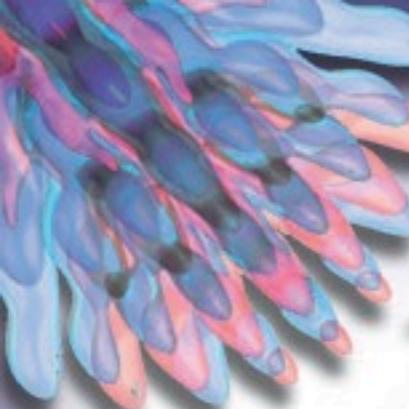
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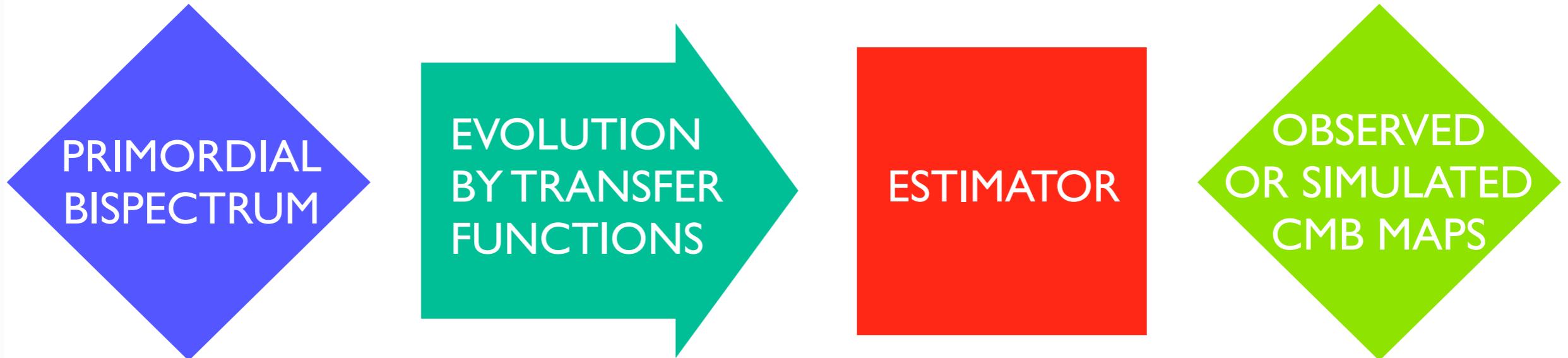
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TetraPyd - bispectrum domain

Allowed multipoles l_1, l_2, l_3 for the CMB bispectrum live in the domain

Resolution: $l_1, l_2, l_3 \leq l_{\max}, \quad l_1, l_2, l_3 \in \mathbb{N},$

Triangle condition: $l_1 \leq l_2 + l_3 \text{ for } l_1 \geq l_2, l_3, + \text{cyclic perms.}$

Parity condition: $l_1 + l_2 + l_3 = 2n, \quad n \in \mathbb{N}.$

Inner product:

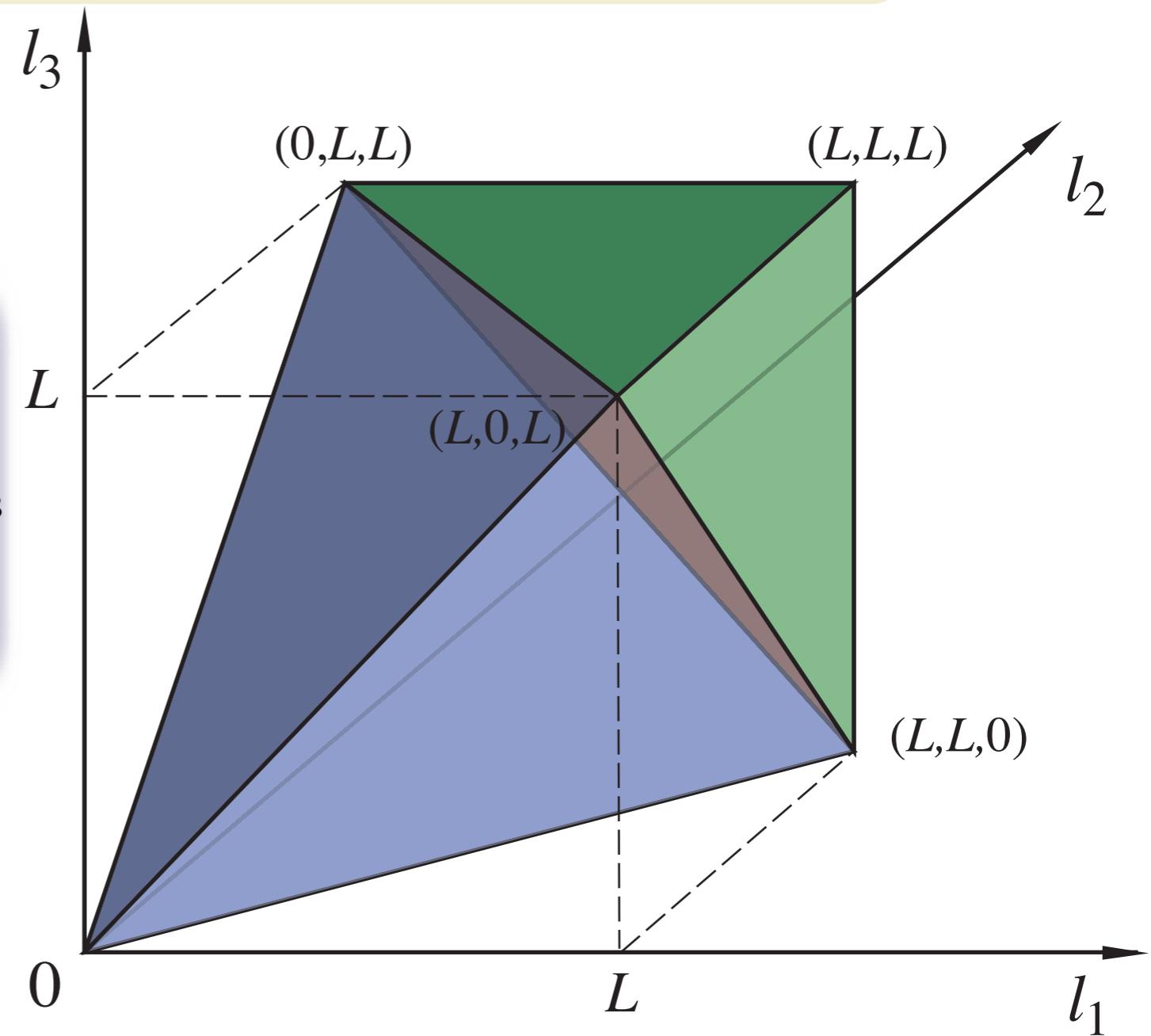
Defined by estimator sum

$$\langle b, b' \rangle \equiv \sum_{l_1, l_2, l_3 \in \mathcal{V}_T} w_{l_1 l_2 l_3} b_{l_1 l_2 l_3} b'_{l_1 l_2 l_3}$$

with weight $w_{l_1 l_2 l_3} = h_{l_1 l_2 l_3}^2$

an (uninteresting) geometric factor

$$h_{l_1 l_2 l_3} = \sqrt{\frac{(2l_1 + 1)(2l_2 + 1)(2l_3 + 1)}{4\pi}} \begin{pmatrix} l_1 & l_2 & l_3 \\ 0 & 0 & 0 \end{pmatrix}$$



Separable Q modes

Need complete set of separable eigenmodes spanning tetrapyd

Define **separable basis functions** \mathbf{Q}_n

$$\begin{aligned}\bar{\mathcal{Q}}_n(l_1, l_2, l_3) &= \frac{1}{6}[\bar{q}_p(l_1)\bar{q}_r(l_2)\bar{q}_s(l_3) + \bar{q}_r(l_1)\bar{q}_p(l_2)\bar{q}_s(l_3) + \text{cyclic perms in } prs] \\ &\equiv \bar{q}_{\{pqrq_s\}} \quad \text{with} \quad n \leftrightarrow \{prs\},\end{aligned}$$

where $q_p(l)$ can be Legendre polynomials, trig. functions, wavelets etc.

Note: $\langle \bar{\mathcal{Q}}_n, \bar{\mathcal{Q}}_p \rangle \equiv \gamma_{np} \neq \delta_{np}$ are not in general orthogonal

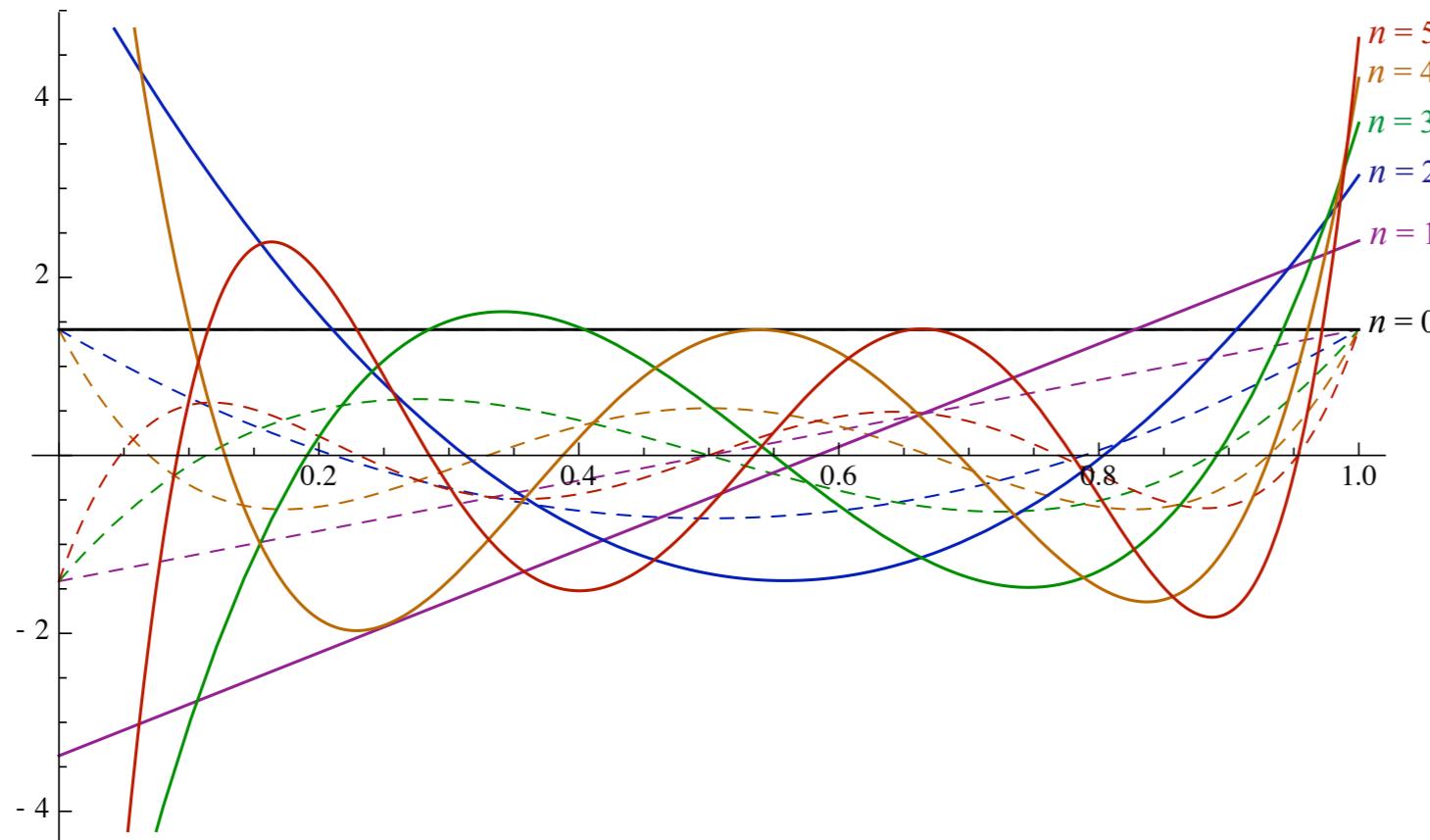
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Orthonormal R basis

Finally, we orthonormalise the Q to create the basis R

$$\langle \bar{\mathcal{R}}_n, \bar{\mathcal{R}}_p \rangle = \delta_{np} \quad \text{with} \quad \bar{\mathcal{R}}_n = \sum_{p=0}^n \lambda_{mp} Q_p$$

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This can be achieved step-by-step with Gram-Schmidt

Or (better) by Cholesky decomposition,
since λ_{nm} is lower triangular

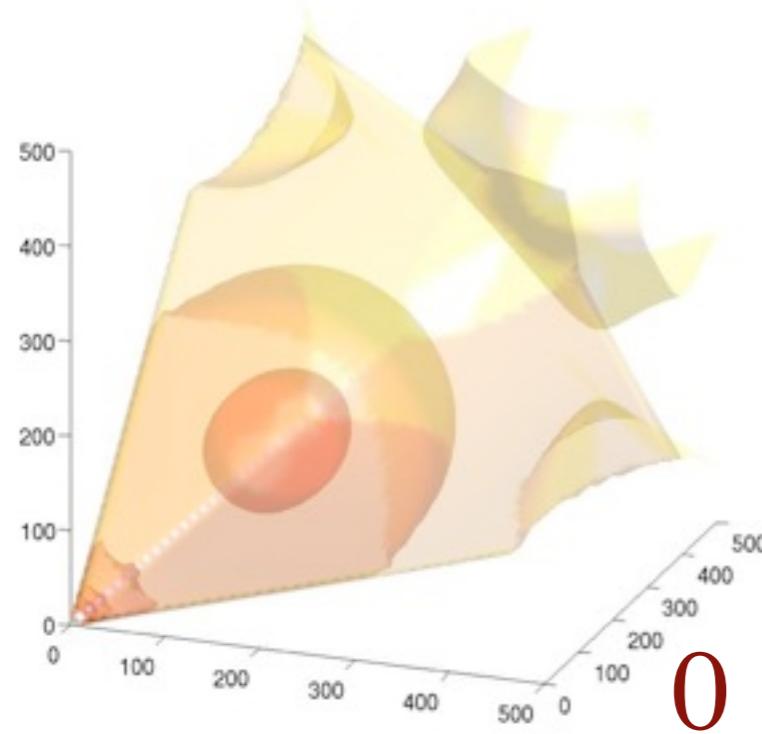
Here $\langle R_n R_m \rangle = \lambda_{nr} \lambda_{ms} \langle Q_r Q_s \rangle$

$$\langle Q_r Q_s \rangle = \gamma_{rs}$$

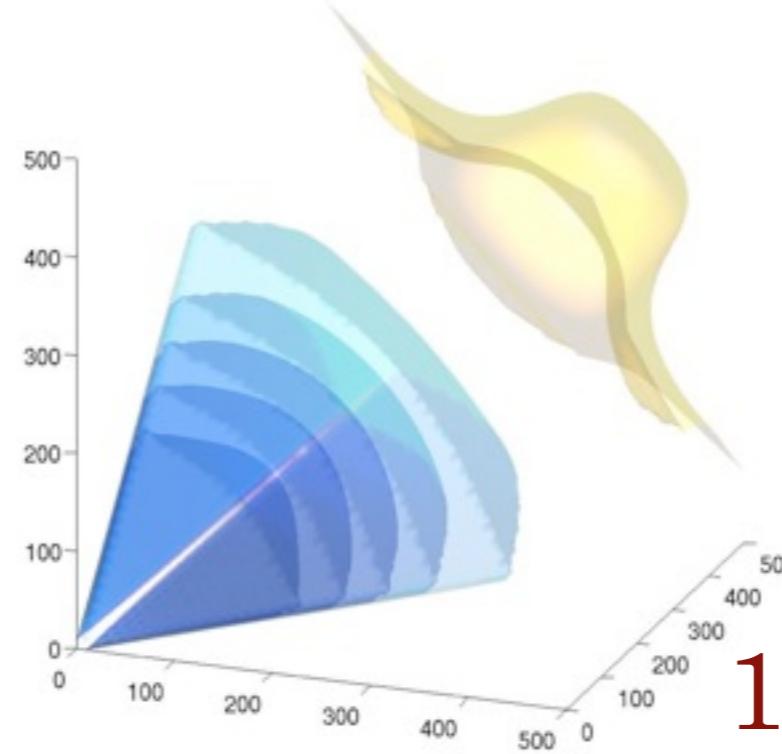
$$\text{so } I = \lambda \gamma \lambda^T \quad \text{and} \quad \gamma = \lambda^{-1} \lambda^{-1 T}$$

(See lecture 2.)

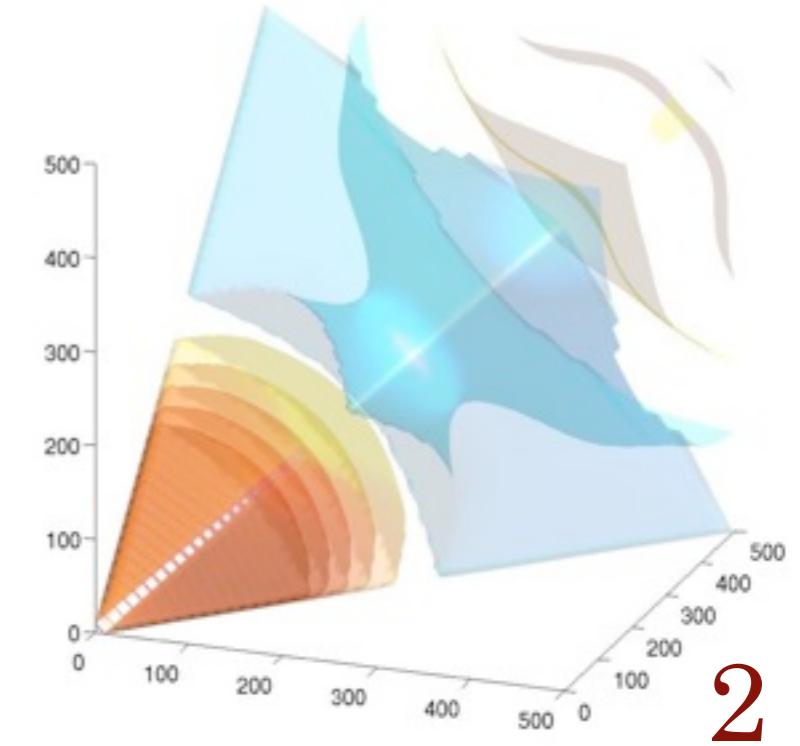
Orthogonal bispectrum modes



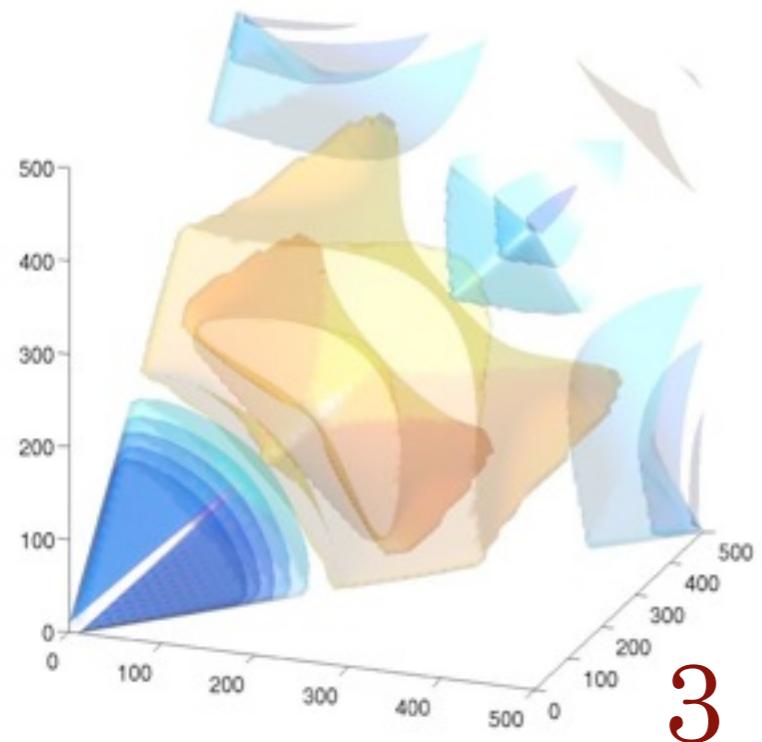
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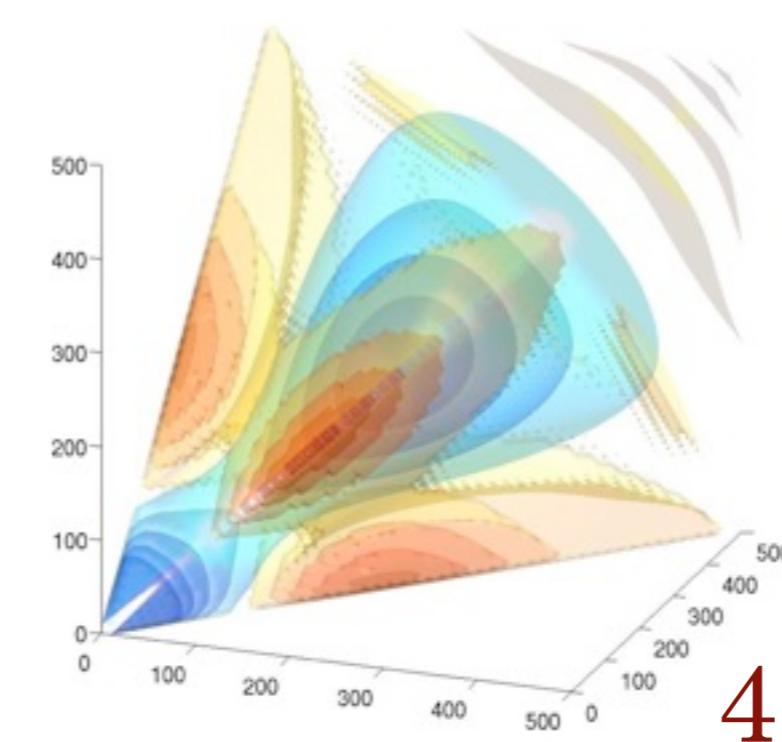
1



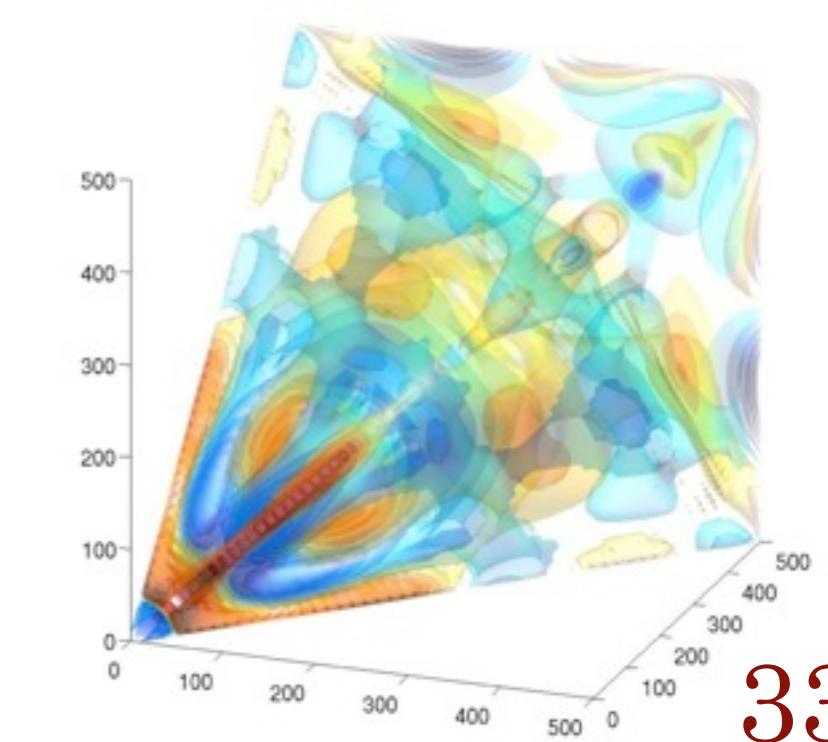
2



3



4



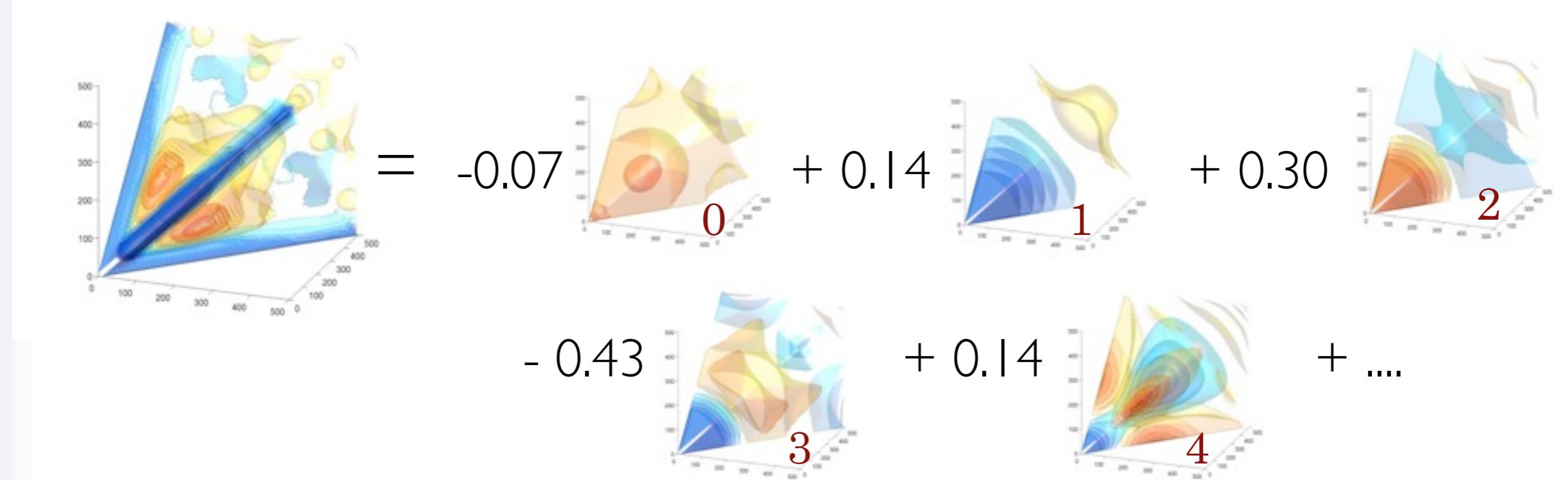
33

Arbitrary bispectrum expansion

Expand the bispectrum signal strength as

$$\frac{v_{l_1} v_{l_2} v_{l_3}}{\sqrt{C_{l_1} C_{l_2} C_{l_3}}} b_{l_1 l_2 l_3} = \sum_n \bar{\alpha}_n^{\mathcal{R}} \bar{\mathcal{R}}_n$$

E.g. Local f_{NL} Model expansion:

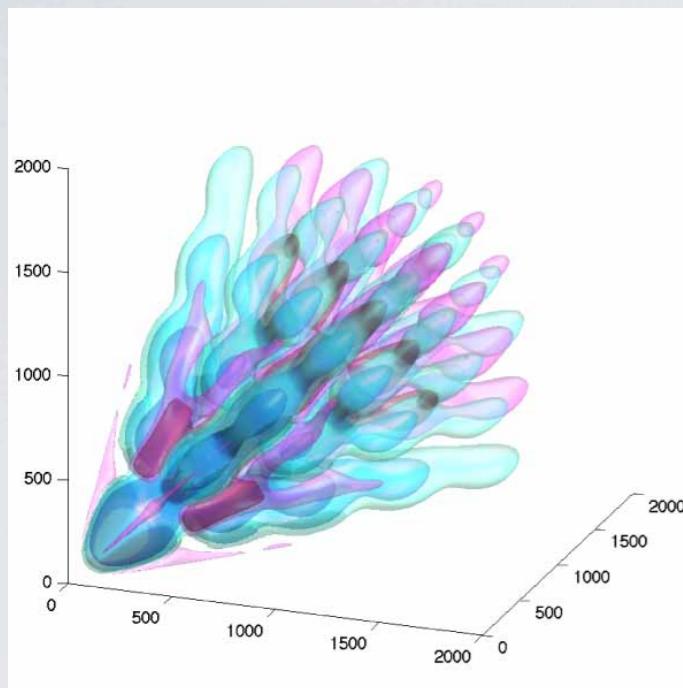


Almost all theoretical bispectra require $n_{\max} < 30$ at WMAP resolution!

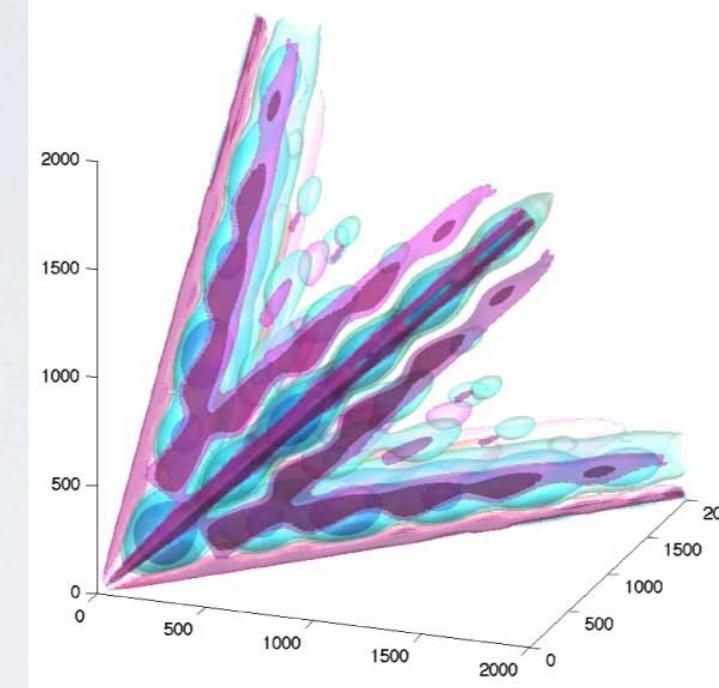
Applicable to inflation, defects, secondaries

Fergusson & EPS, arXiv:1008.1730

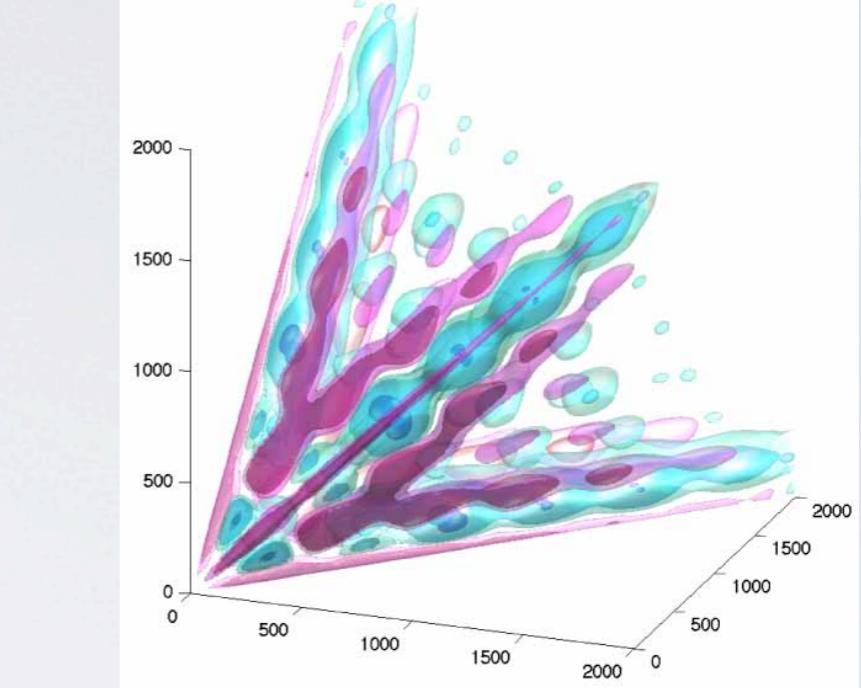
Planck resolution CMB bispectra (multipoles l_1, l_2, l_3)



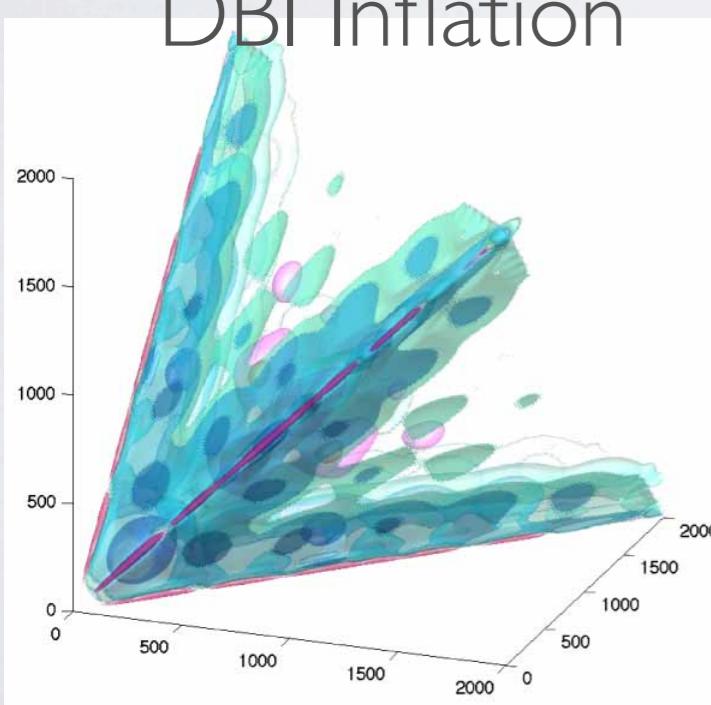
DBI Inflation



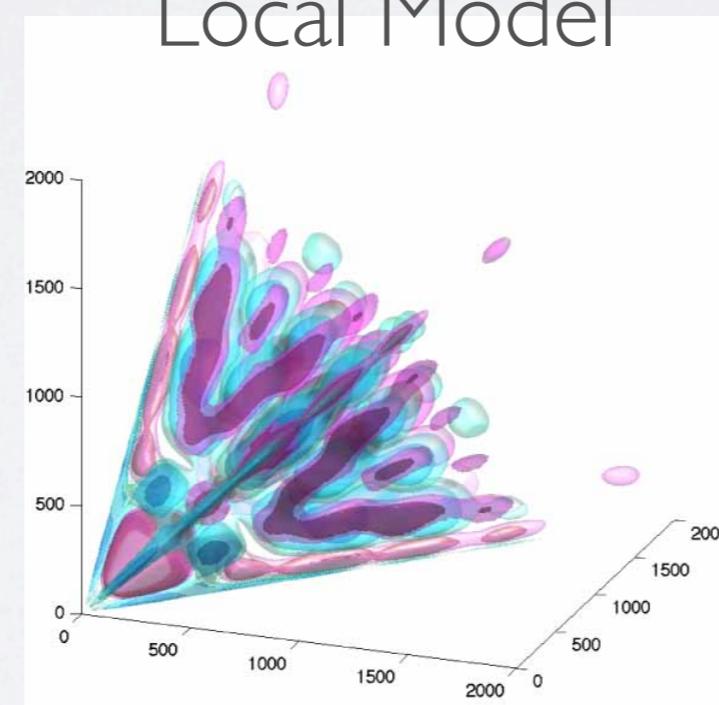
Local Model



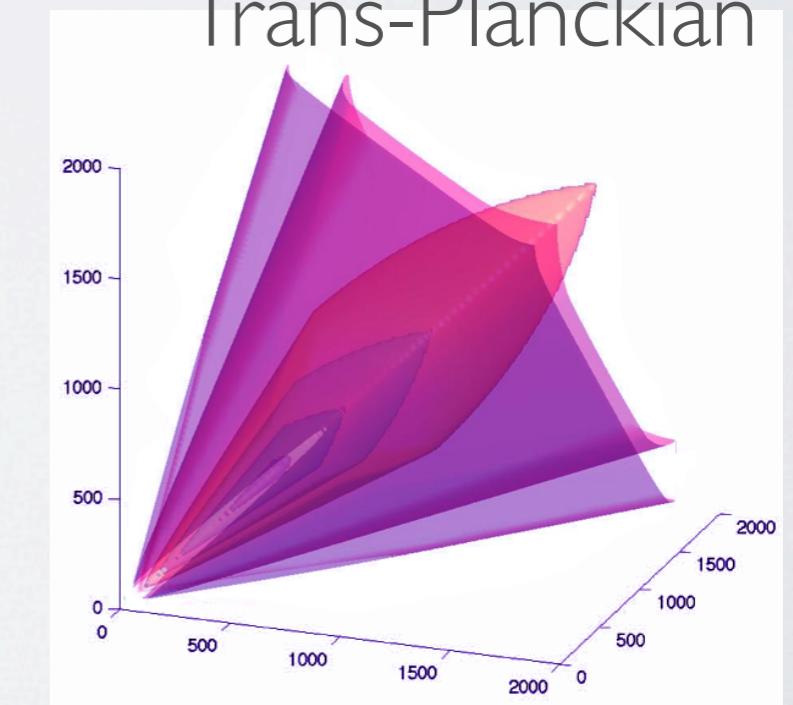
Trans-Planckian



Warm inflation

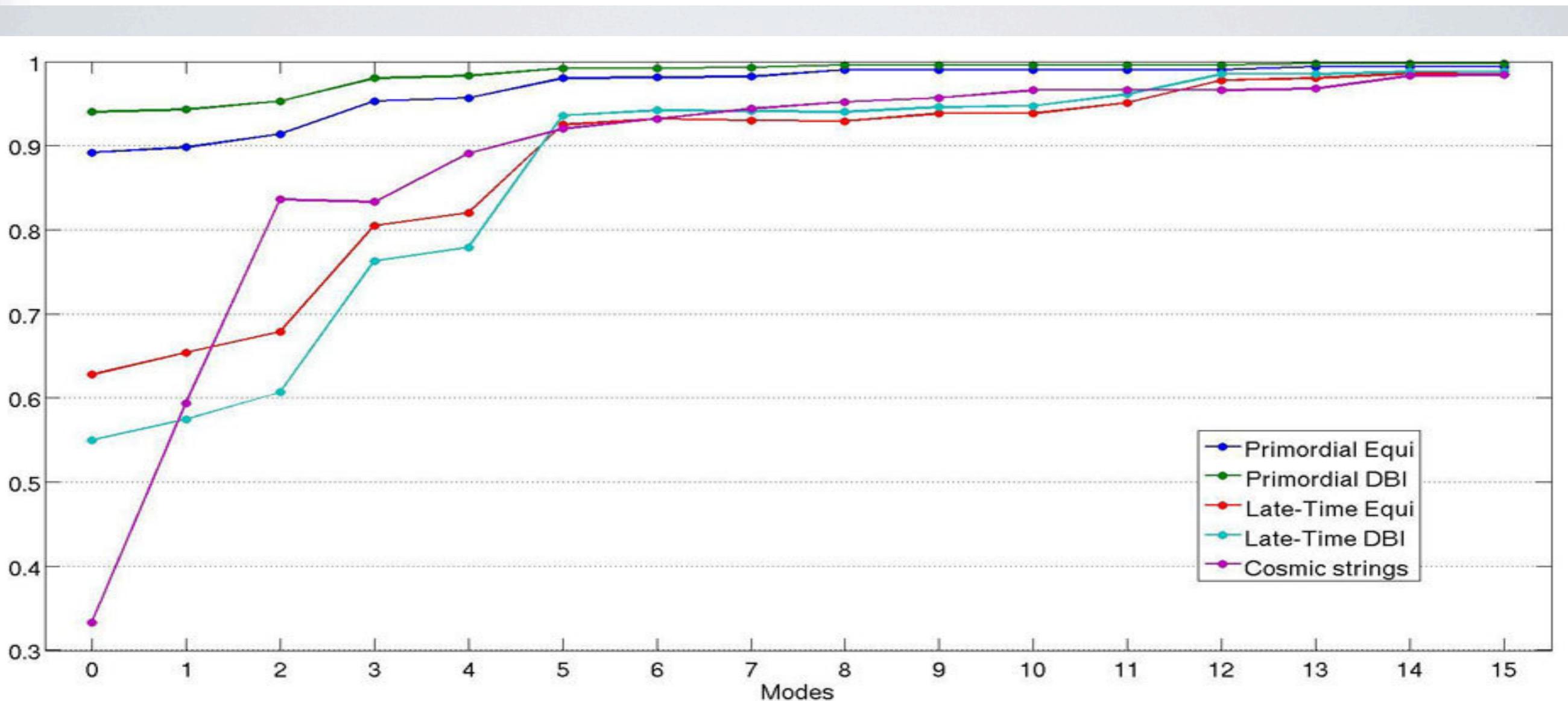


Feature Model



Cosmic strings

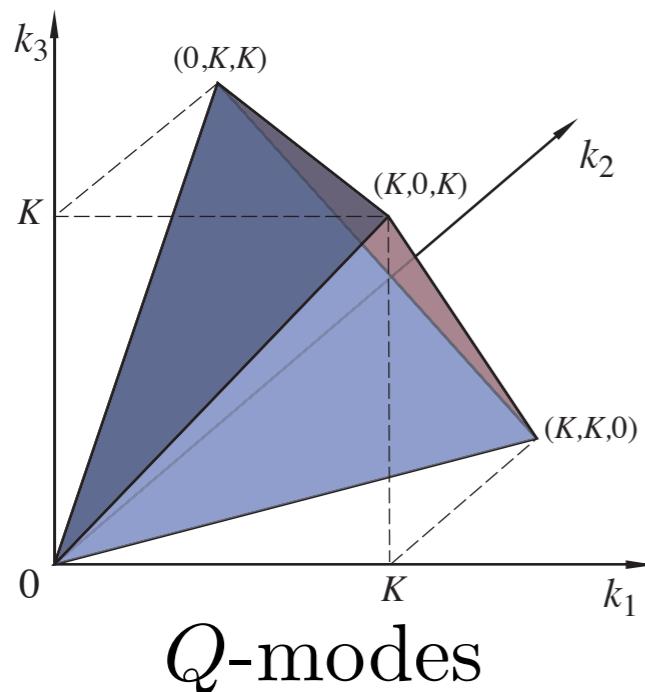
Rapid modal convergence



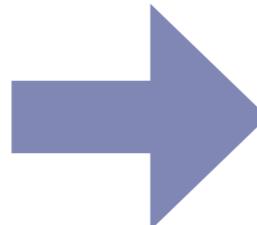
Correlation of the separable approximation to the original bispectra, both primordial and CMB

Primordial to CMB basis

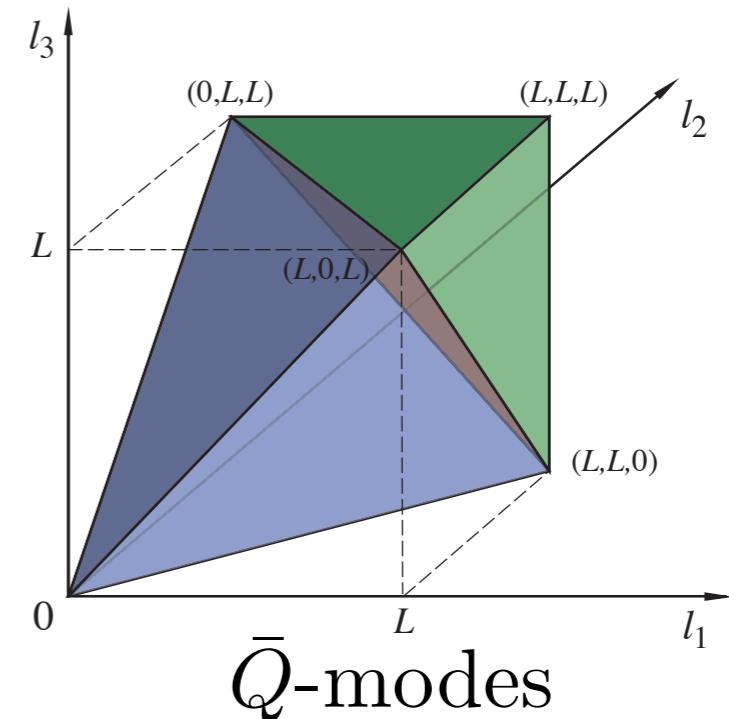
Primordial bispectrum (k -space)



Transfer
functions
 $\Delta_l(k)$



CMB bispectrum (l -space)



Use transfer functions once to project forward primordial modes $Q \rightarrow \tilde{Q}$, so we calculate

$$\Gamma_{nm} = \left\langle \bar{Q}^n \frac{vvv\tilde{Q}^m}{\sqrt{CCC}} \right\rangle$$

Then we can transform between the primordial and CMB expansions

$$\bar{\alpha}^Q = \bar{\gamma}^{-1} \Gamma \alpha^Q$$

Modal estimator

Substitute bispectrum mode expansion

$$\frac{v_{l_1} v_{l_2} v_{l_3}}{\sqrt{C_{l_1} C_{l_2} C_{l_3}}} b_{l_1 l_2 l_3} = \sum_n \bar{\alpha}_n^{\mathcal{Q}} \bar{\mathcal{Q}}_n(l_1, l_2, l_3),$$

into the estimator $\mathcal{E} = \frac{1}{N^2} \sum_{l_i m_i} \frac{\mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3} b_{l_1 l_2 l_3}^{\text{th}} a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3}}{C_{l_1} C_{l_2} C_{l_3}}$

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Bonus: Full CMB bispectrum reconstruction since

$$\langle \bar{\beta}_n^{\mathcal{R}} \rangle = \bar{\alpha}_n^{\mathcal{R}}$$

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Theory *Observed*

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Modal estimator

Substitute bispectrum mode expansion

$$\frac{v_{l_1} v_{l_2} v_{l_3}}{\sqrt{C_{l_1} C_{l_2} C_{l_3}}} b_{l_1 l_2 l_3} = \sum_n \bar{\alpha}_n^{\mathcal{Q}} \bar{\mathcal{Q}}_n(l_1, l_2, l_3),$$

into the estimator \mathcal{E} $= \frac{1}{N^2} \sum_{l_i m_i} \frac{\mathcal{G}_{m_1 m_2 m_3}^{l_1 l_2 l_3} b_{l_1 l_2 l_3}^{\text{th}} a_{l_1 m_1} a_{l_2 m_2} a_{l_3 m_3}}{C_{l_1} C_{l_2} C_{l_3}}$

$$\mathcal{E} = \sum_n \bar{\alpha}_n^{\mathcal{Q}} \bar{\beta}_n^{\mathcal{Q}}$$

Theory  *Observed*  *Mode filtered maps* 

where $\bar{\beta}_n^{\mathcal{Q}} = \int d^2 \hat{\mathbf{n}} \bar{M}_{\{p}(\hat{\mathbf{n}}) \bar{M}_r(\hat{\mathbf{n}}) \bar{M}_{s\}}(\hat{\mathbf{n}})$ with $\bar{M}_p(\hat{\mathbf{n}}) = \sum_{lm} \bar{q}_p(l) \frac{a_{lm}}{v_l \sqrt{C_l}} Y_{lm}(\hat{\mathbf{n}})$

Bonus: Full CMB bispectrum reconstruction since

$$\langle \bar{\beta}_n^{\mathcal{R}} \rangle = \bar{\alpha}_n^{\mathcal{R}}$$

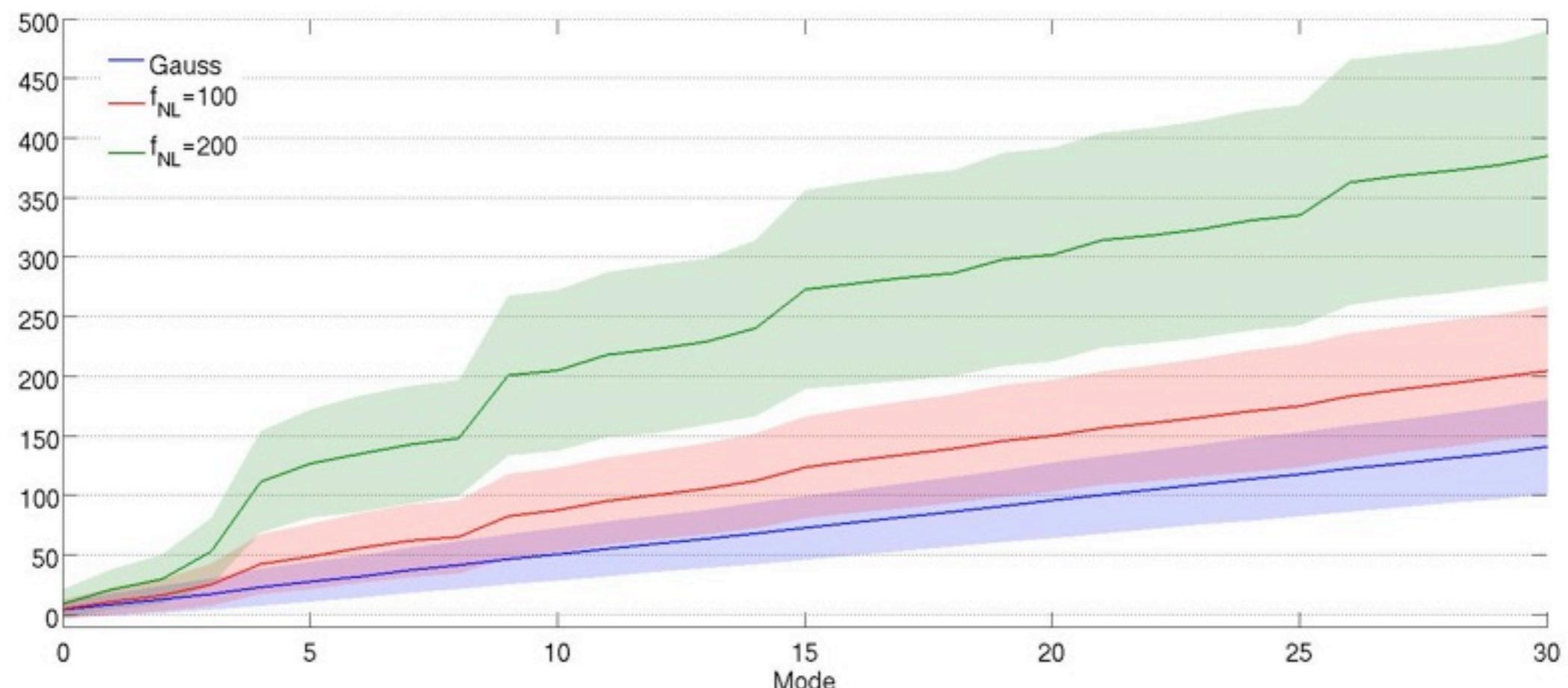
Total bispectrum measure

Total integrated bispectrum (with Parseval's theorem)

$$\bar{F}_{\text{NL}}^2 \equiv \frac{1}{N_{\text{loc}}^2} \sum_{l_i} \frac{h_{l_1 l_2 l_3}^2 b_{l_1 l_2 l_3}^2}{C_{l_1} C_{l_2} C_{l_3}} = \frac{1}{N_{\text{loc}}^2} \sum_n \bar{\alpha}_n^{\mathcal{R}2}$$

Remove the Gaussian mean

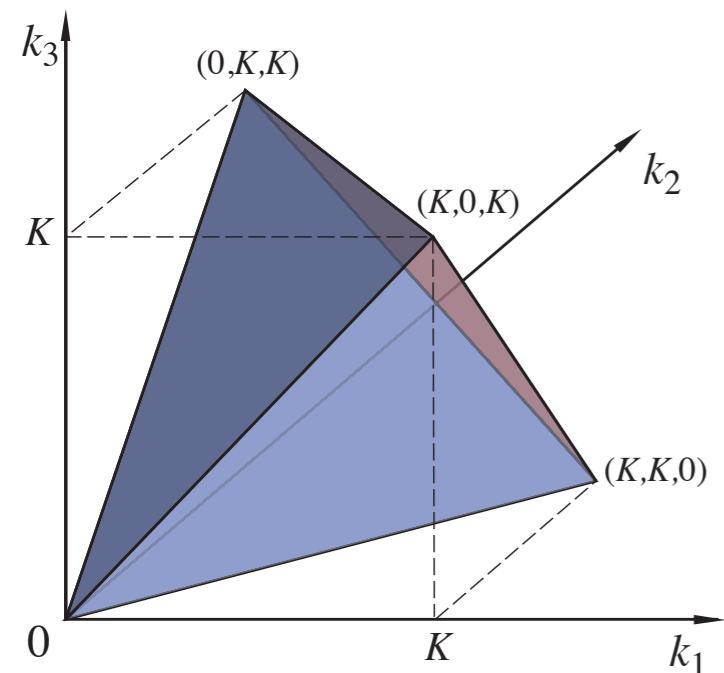
$$\bar{F}_{\text{NL}}^2 = \frac{1}{N_{\text{loc}}^2} \left[\sum_n \bar{\beta}_n^{\mathcal{R}2} - n_{\max} \sigma^2 \right]$$



Modal Polyspectra Estimation

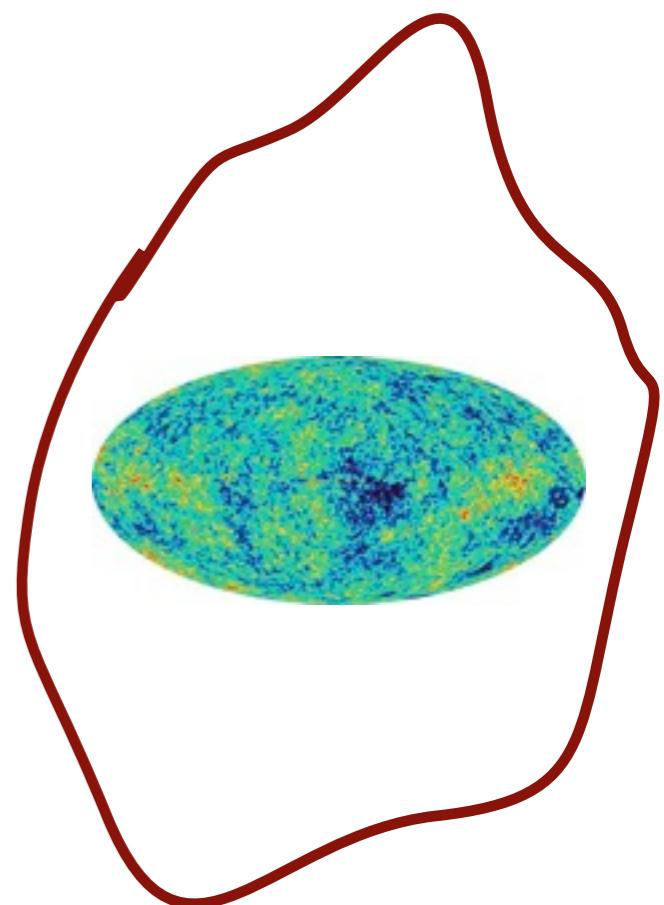
THEORY

Space of (primordial)
isotropic polyspectra
(k -space)



OBSERVATION

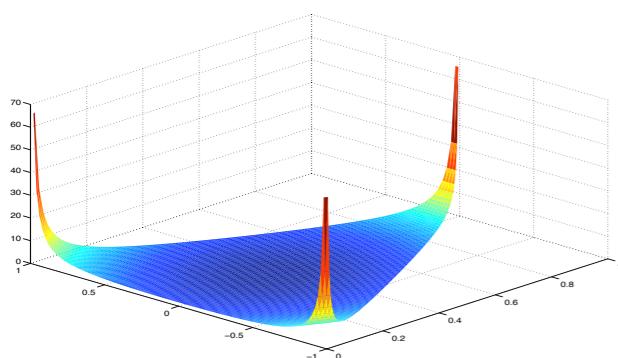
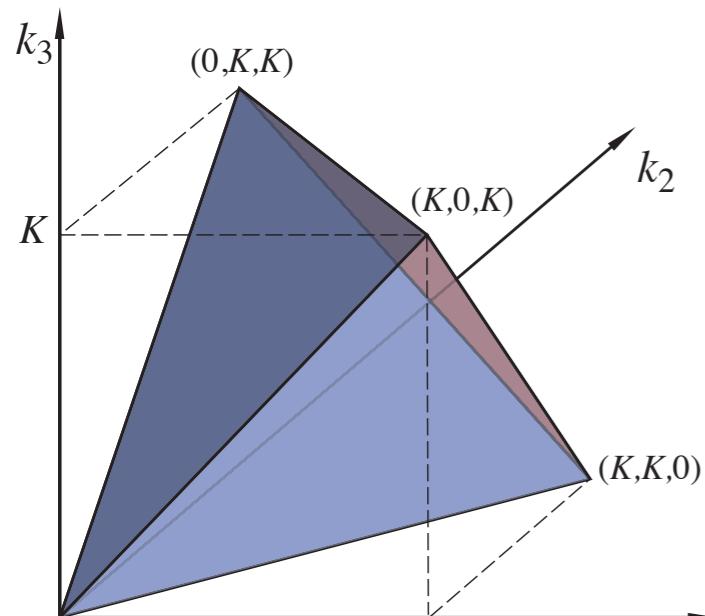
Space V of possible
CMB polyspectra



Modal Polyspectra Estimation

THEORY

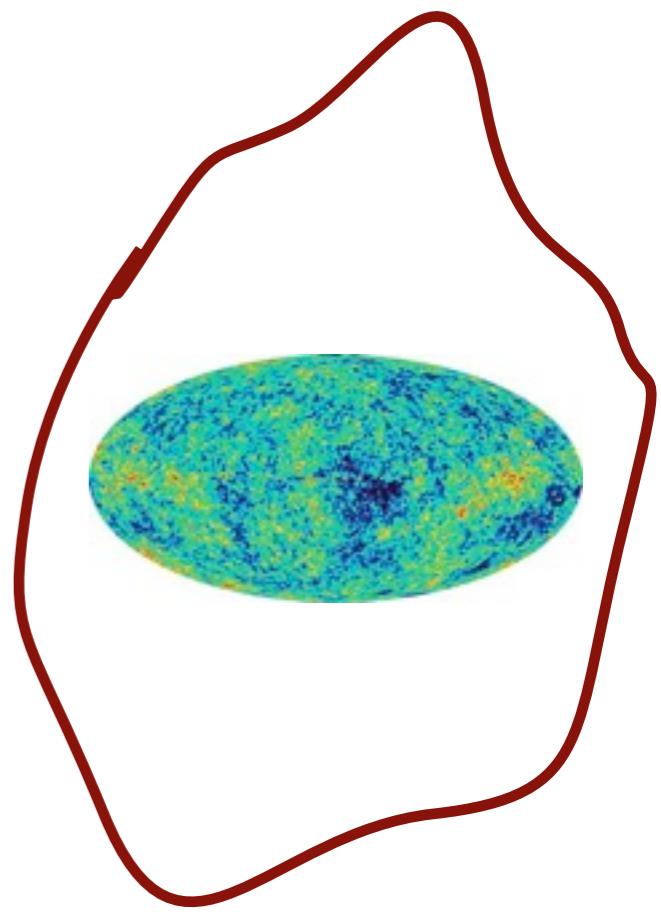
Space of (primordial)
isotropic polyspectra
(k -space)



Expand any model with
primordial modes α_n

OBSERVATION

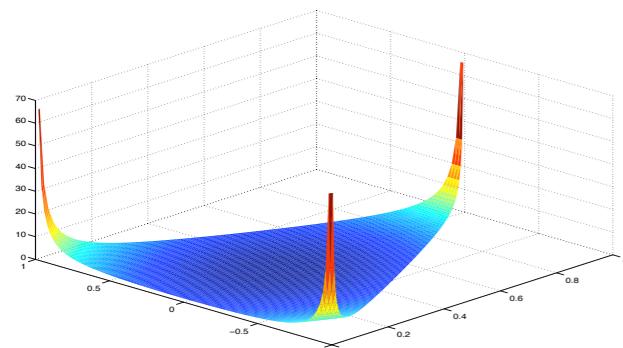
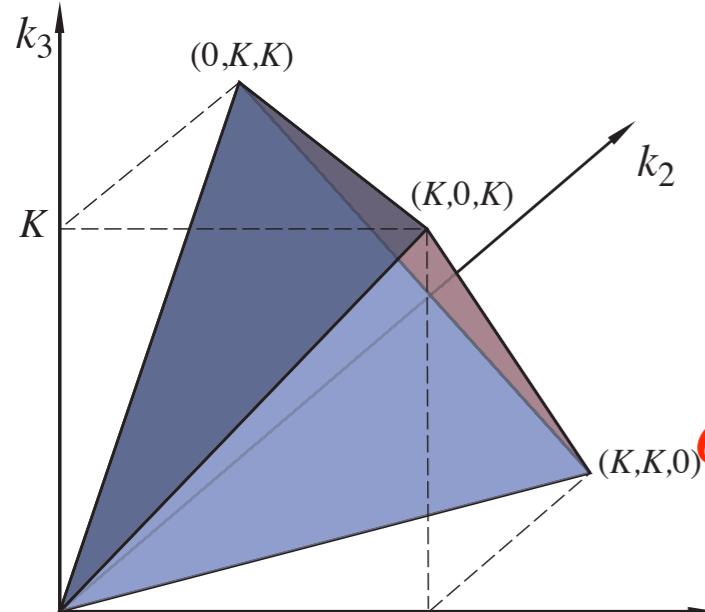
Space V of possible
CMB polyspectra



Modal Polyspectra Estimation

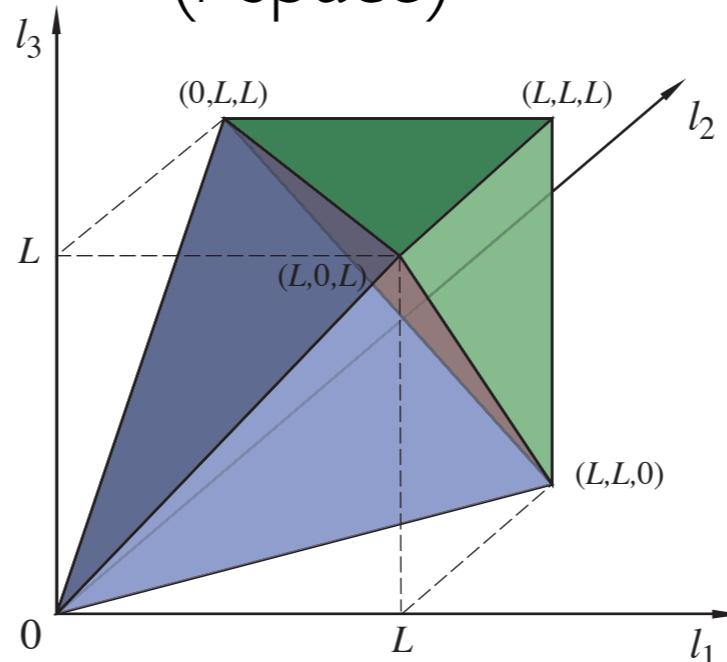
THEORY

Space of (primordial)
isotropic polyspectra
(k -space)



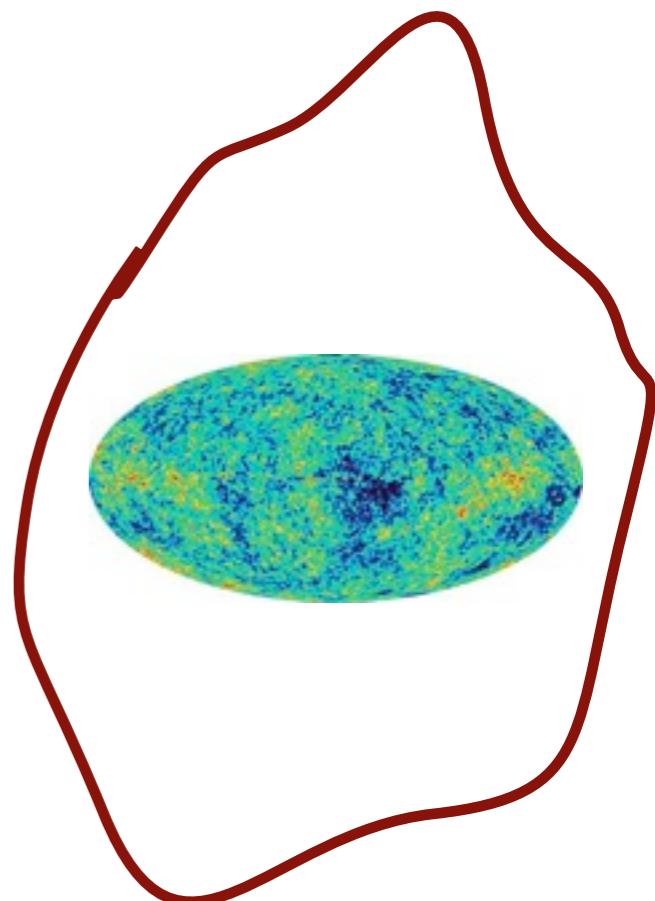
Expand any model with
primordial modes α_n

Projected space V_P
of CMB polyspectra
(l -space)



OBSERVATION

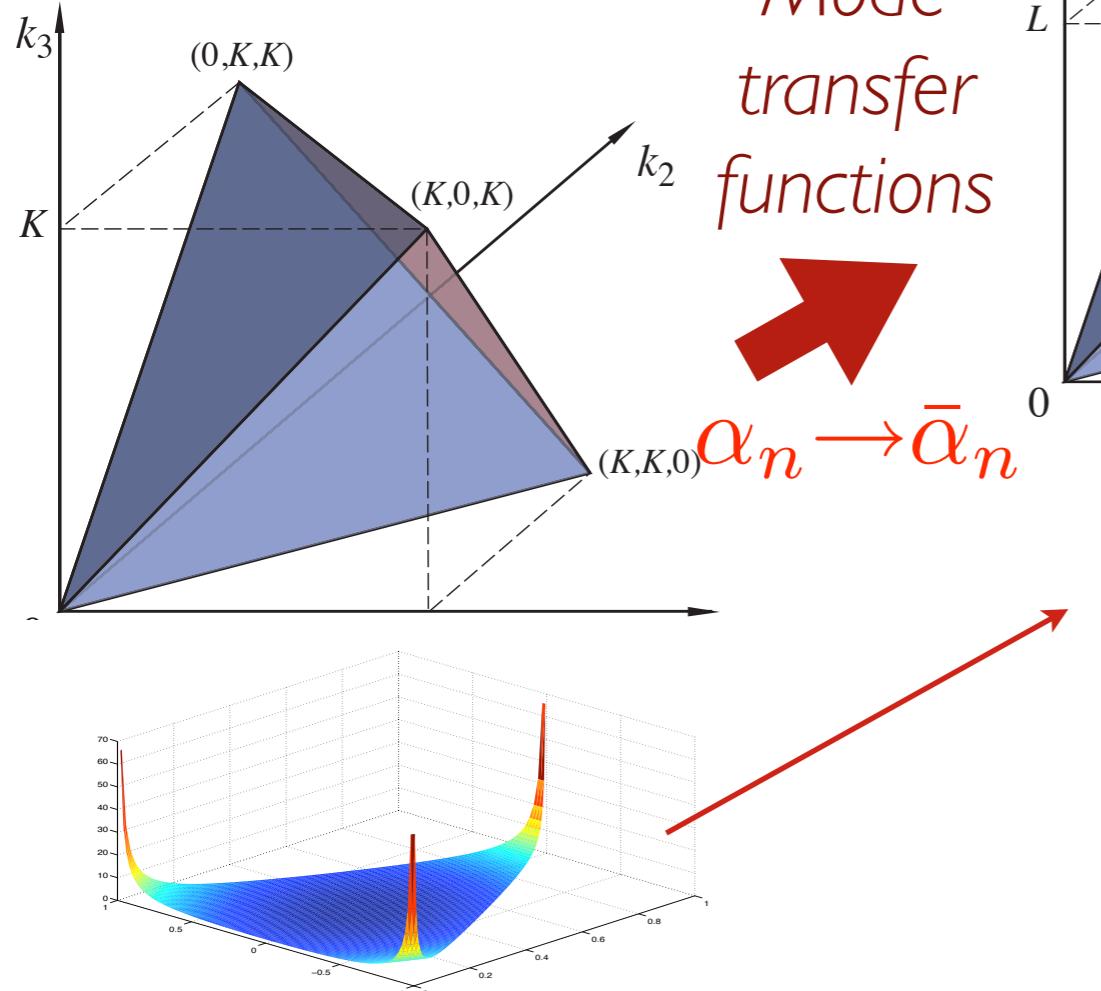
Space V of possible
CMB polyspectra



Modal Polyspectra Estimation

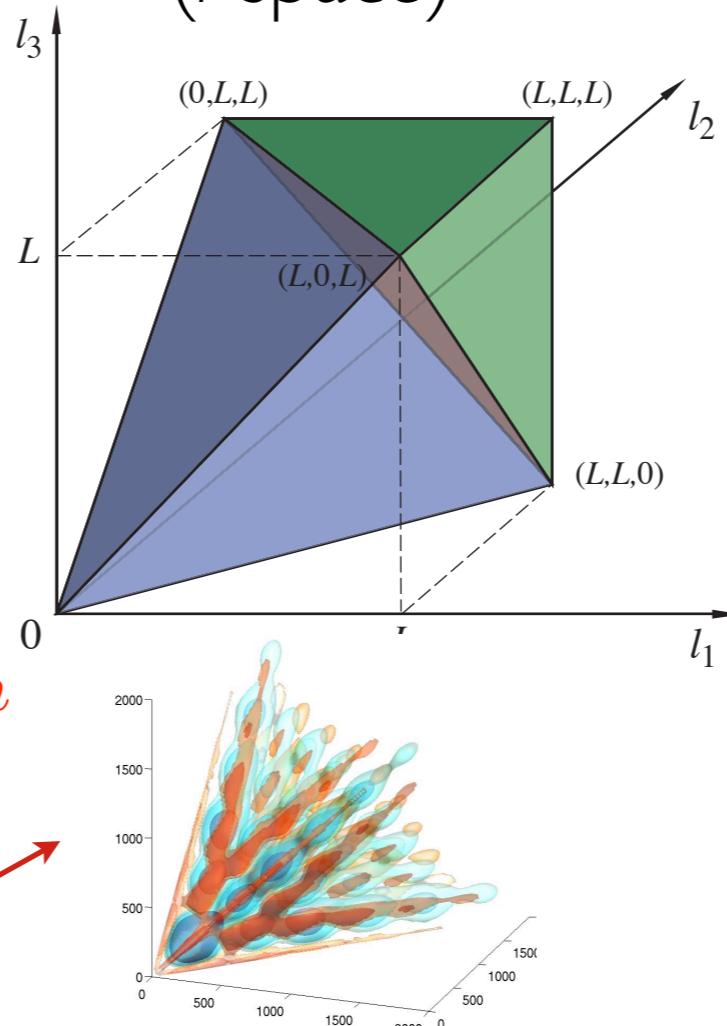
THEORY

Space of (primordial)
isotropic polyspectra
(k -space)



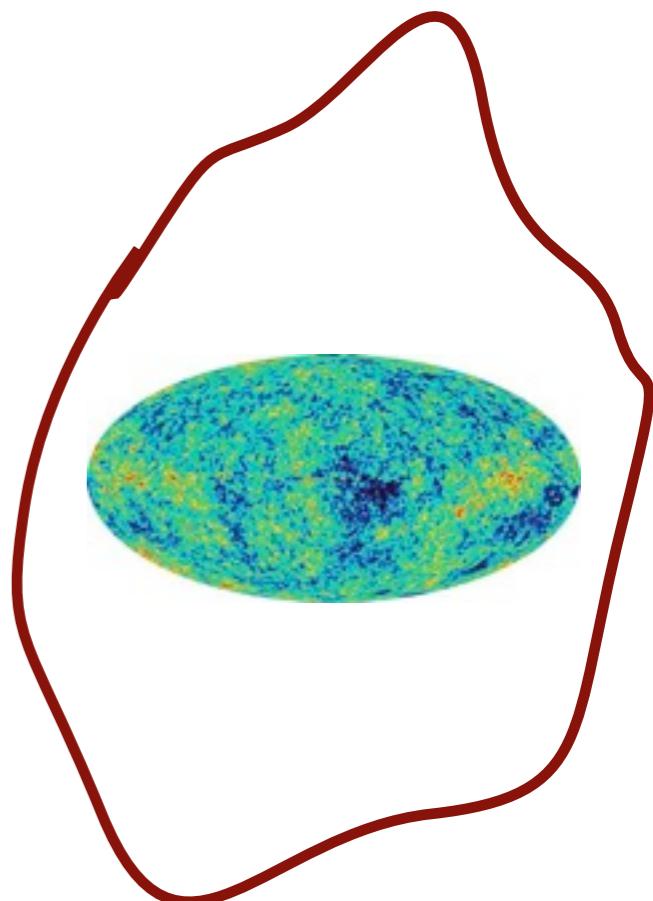
Expand any model with
primordial modes α_n

Projected space V_P
of CMB polyspectra
(l -space)



OBSERVATION

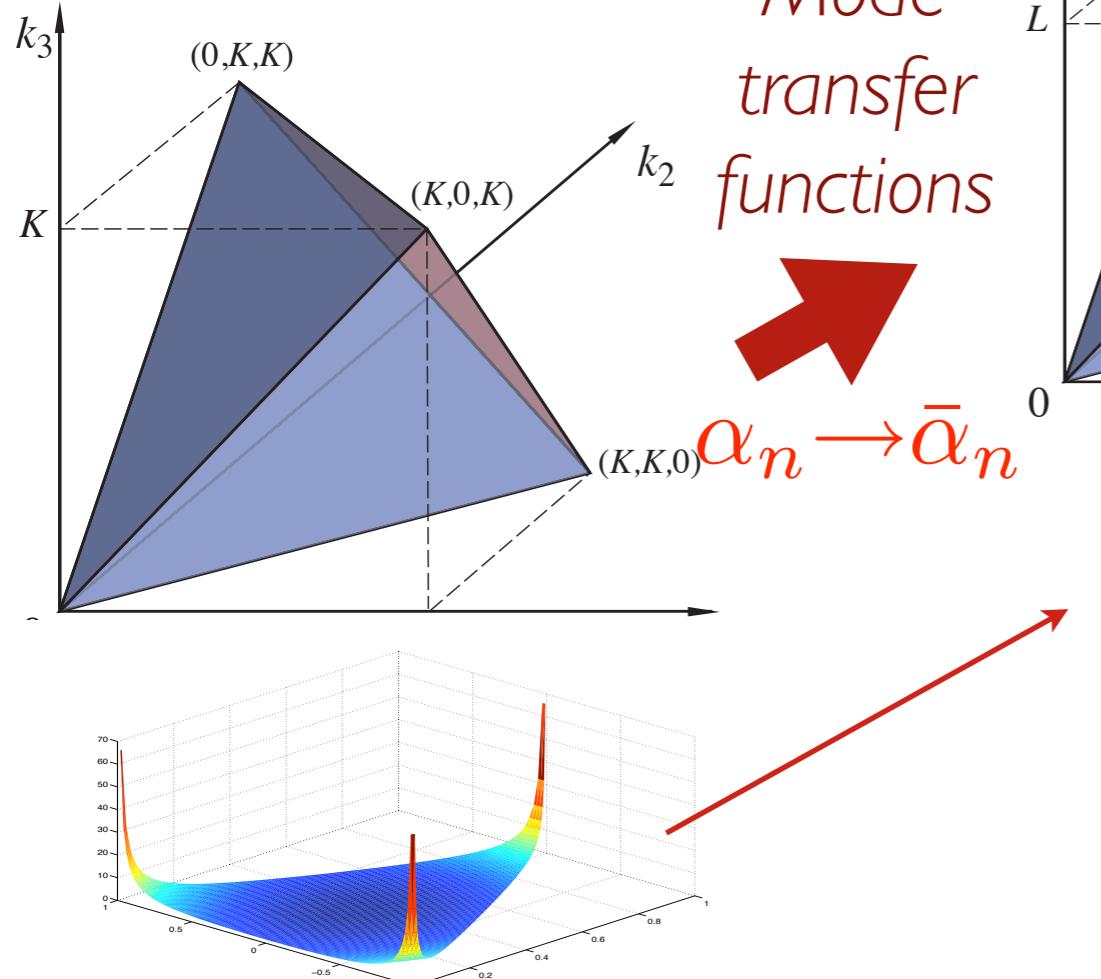
Space V of possible
CMB polyspectra



Modal Polyspectra Estimation

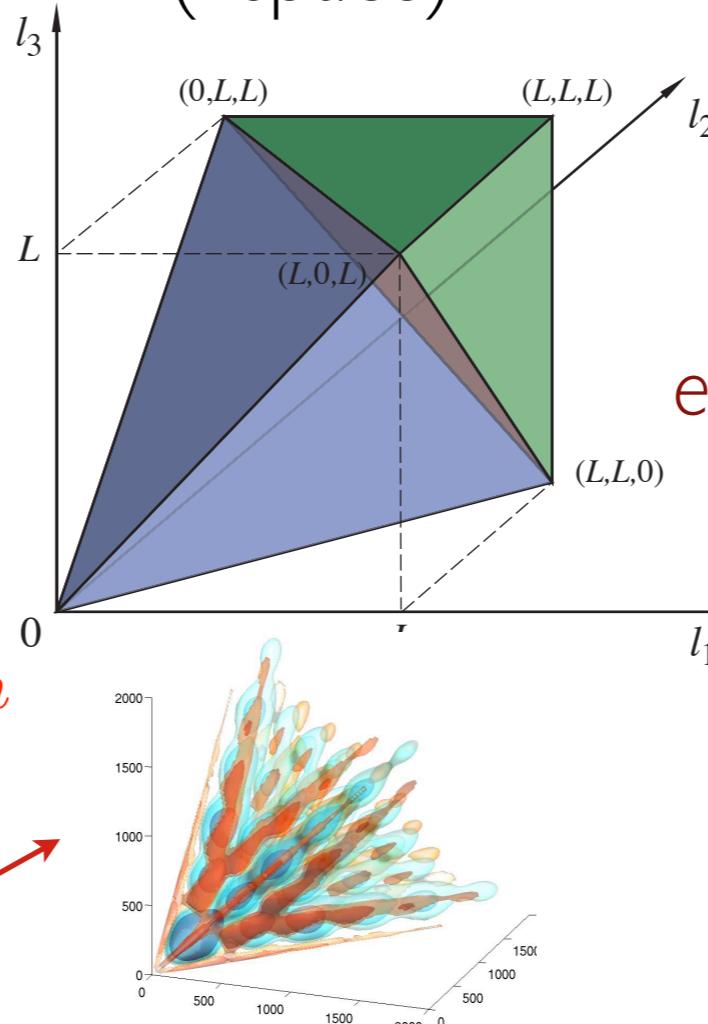
THEORY

Space of (primordial) isotropic polyspectra (k-space)



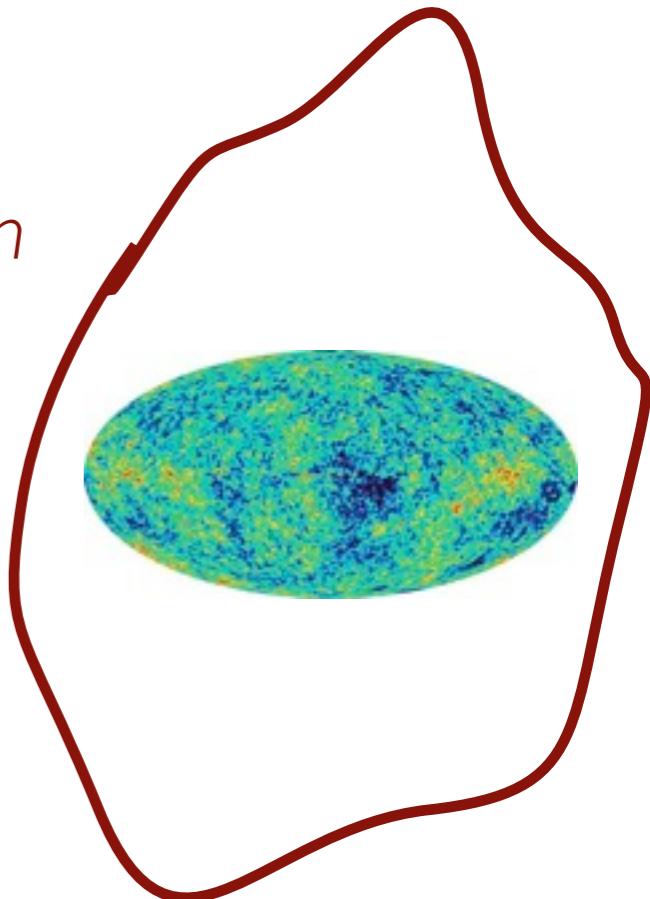
Expand any model with primordial modes α_n

Projected space V_P of CMB polyspectra (l-space)



Map extraction
 $\bar{\beta}_n$

OBSERVATION
Space V of possible CMB polyspectra

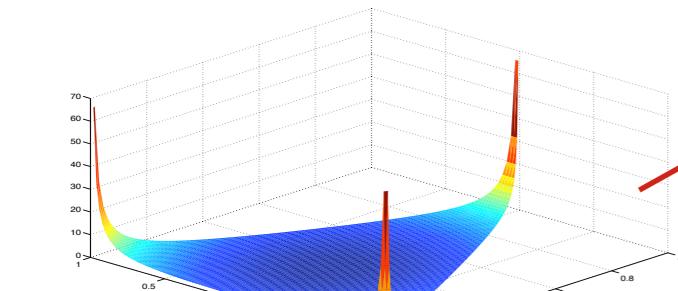
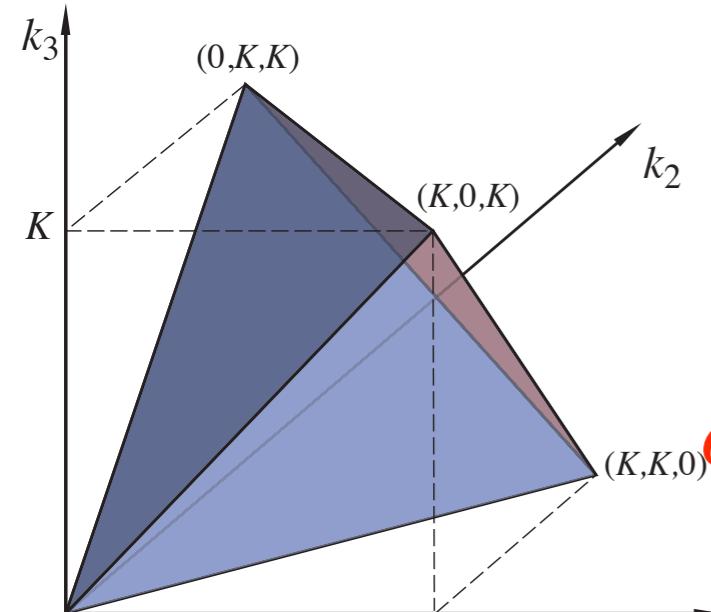


Filter with sufficient separable eigenmodes

Modal Polyspectra Estimation

THEORY

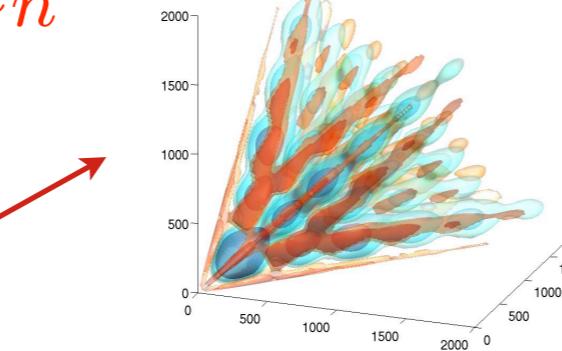
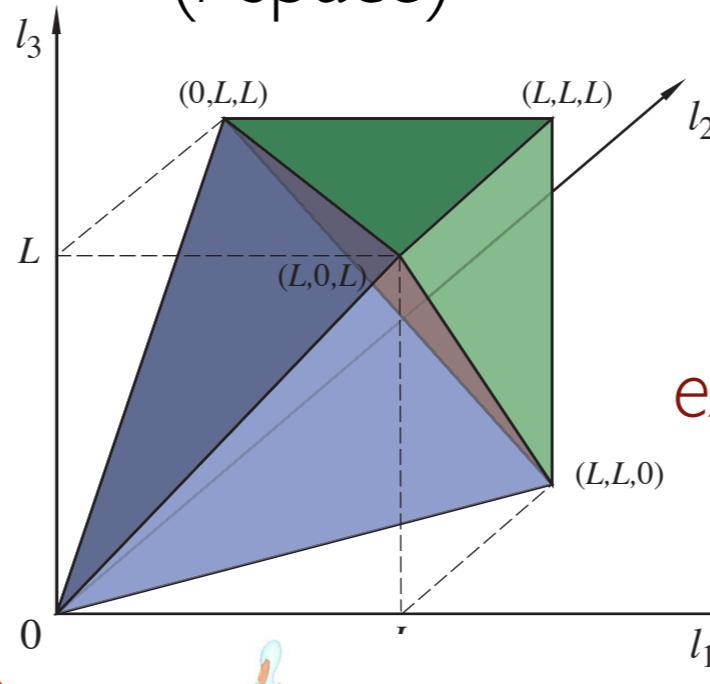
Space of (primordial) isotropic polyspectra (k-space)



Expand any model with primordial modes α_n

Mode transfer functions
 $\alpha_n \rightarrow \bar{\alpha}_n$

Projected space V_P of CMB polyspectra (l-space)

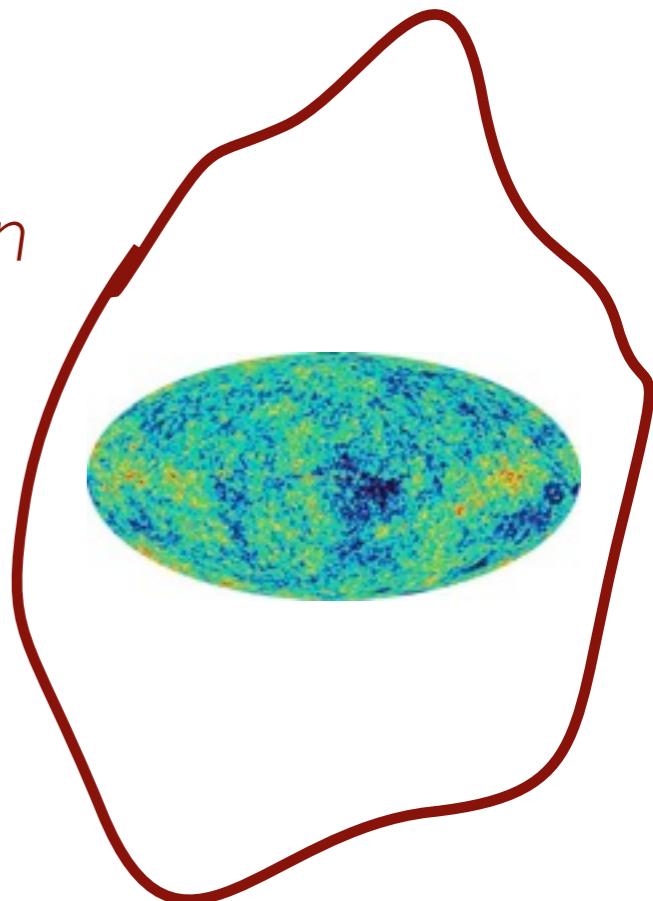


Modal estimator

$$\mathcal{E} = \frac{\sum_n \bar{\alpha}_n^R \bar{\beta}_n^R}{\sum_n (\bar{\alpha}_n^R)^2}$$

Map extraction
 $\bar{\beta}_n$

OBSERVATION
 Space V of possible CMB polyspectra

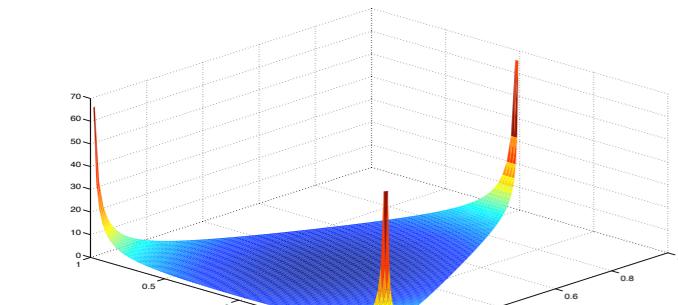
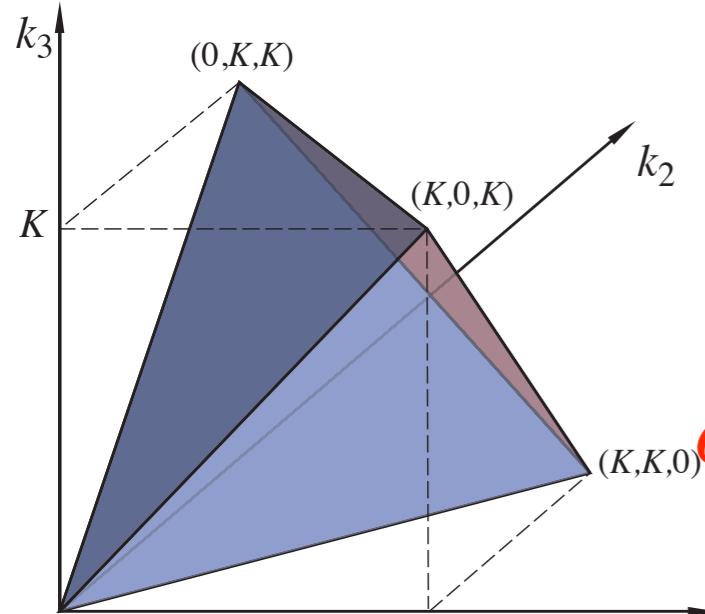


Filter with sufficient separable eigenmodes

Modal Polyspectra Estimation

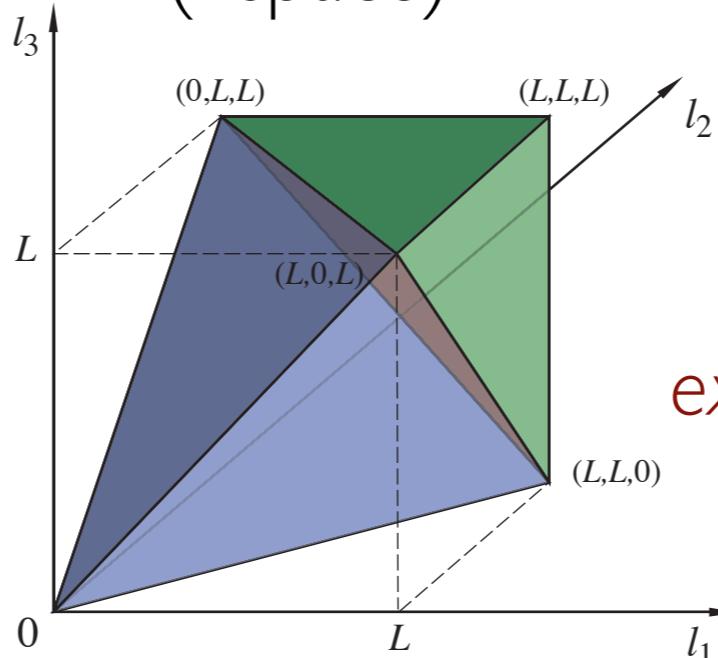
THEORY

Space of (primordial) isotropic polyspectra (k-space)



Expand any model with primordial modes α_n

Projected space V_P of CMB polyspectra (l-space)

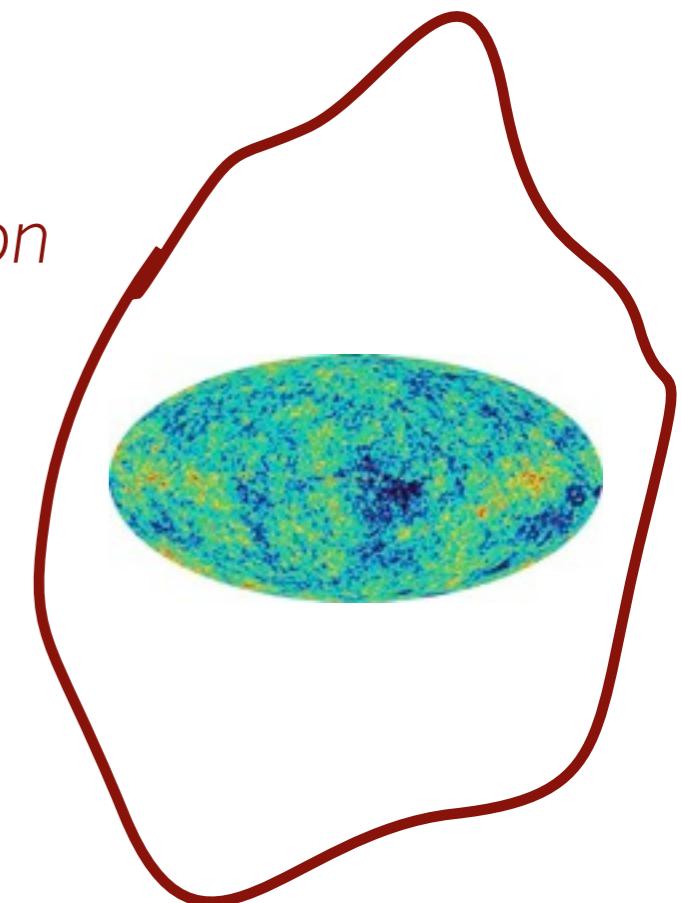


Optimal modal estimator with weight $\zeta = \mathcal{R}\mathcal{C}\mathcal{R}^T = \langle \beta\beta^T \rangle$
see Ferguson & EPS arXiv:1105.2791

$$\mathcal{E} = \frac{\alpha^T \zeta^{-1} \beta}{\alpha^T \zeta^{-1} \alpha}$$

OBSERVATION

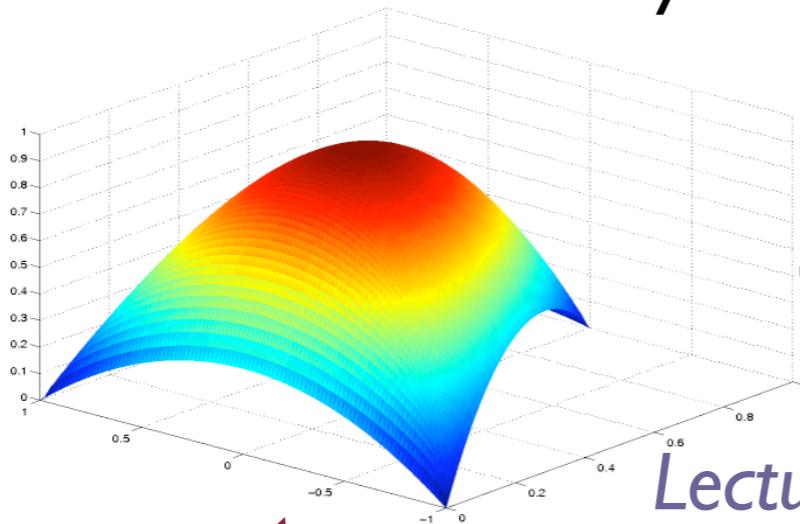
Space V of possible CMB polyspectra



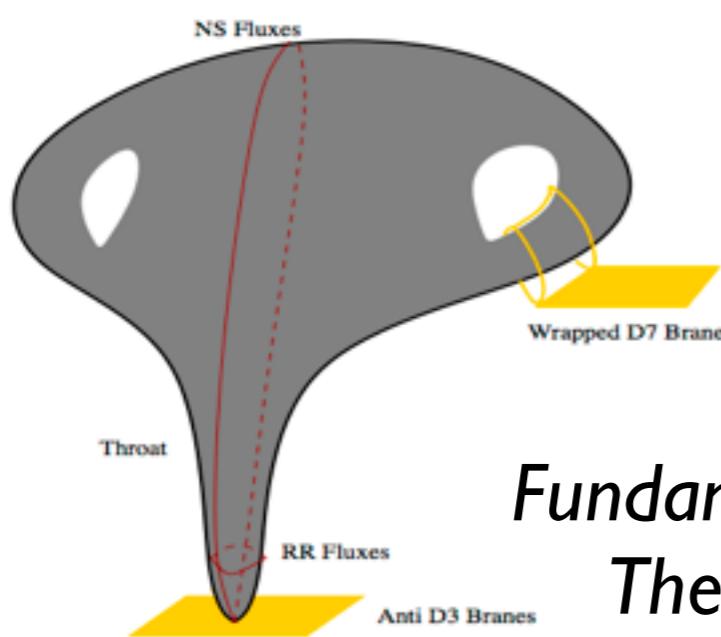
Filter with sufficient separable eigenmodes

Motivation

Primordial non-Gaussianity

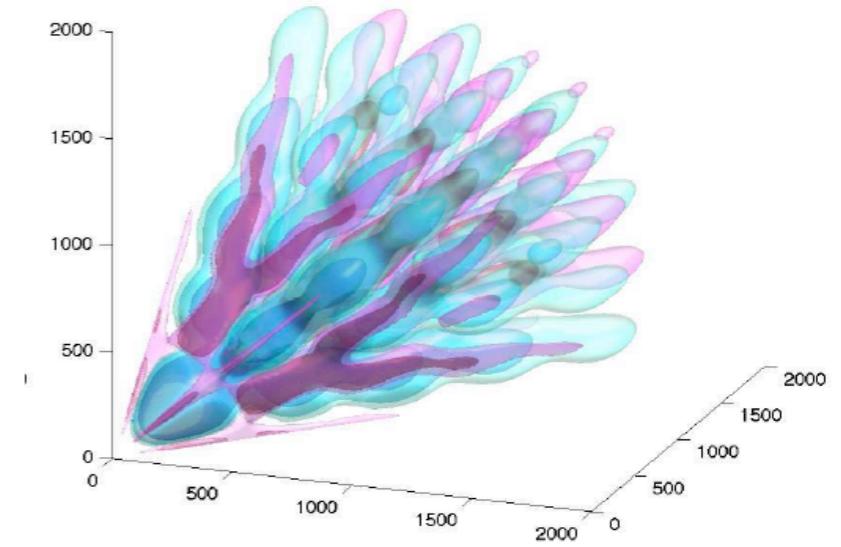


Lecture 1

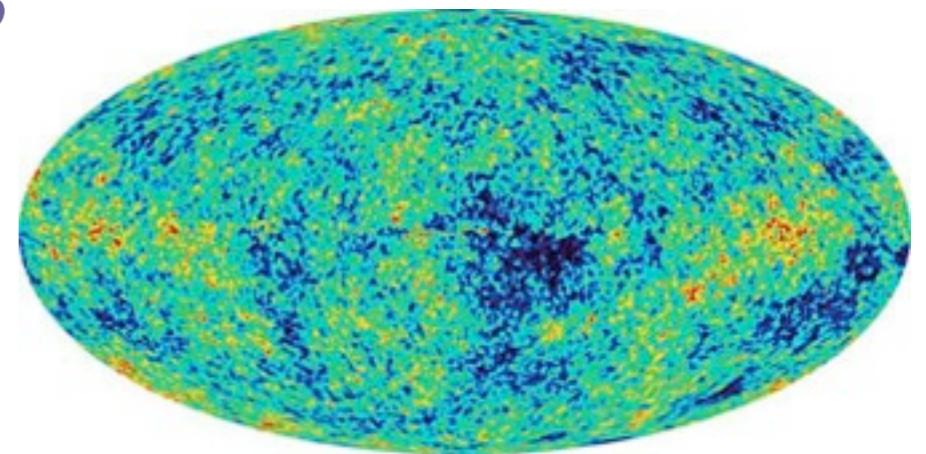


Fundamental
Theory

CMB (or LSS) fingerprint



Lecture 2



Lecture 3

Observational Data