

Primordial black holes and stochastic inflation

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in collaboration with D. Figueroa, S. Raatikainen, S. Räsänen

Why primordial black holes (PBHs)?

Black holes formed in early Universe

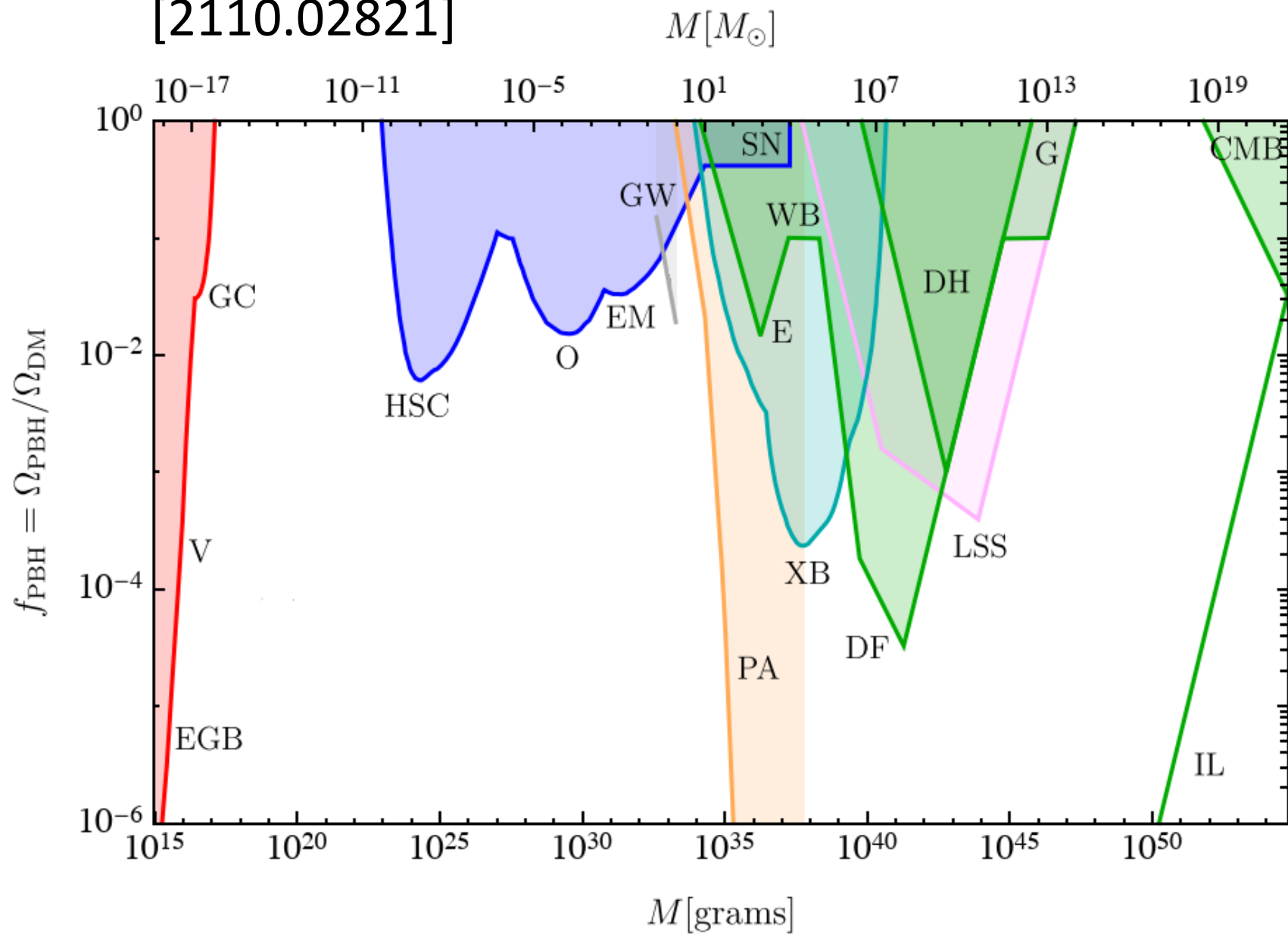
- Carry information of conditions there (small-scale perturbations)
- Any mass (Hawking evaporation?)

Applications in cosmology

- Dark matter candidate
- Seeds of supermassive black holes



[2110.02821]



Renewed interest since GW detection

Gravitational wave signals:

- Black hole mergers
- Stochastic scalar-induced gravitational waves

Matching stochastic GW signal to PBH statistics?

- GWs: sourced by typical scalar perturbations
- PBHs: sourced by extreme scalar perturbations

Origins of primordial black holes

Cosmological phase transitions

Cosmic strings

Primordial perturbations: cosmic inflation

Origins of primordial black holes

Cosmological phase transitions

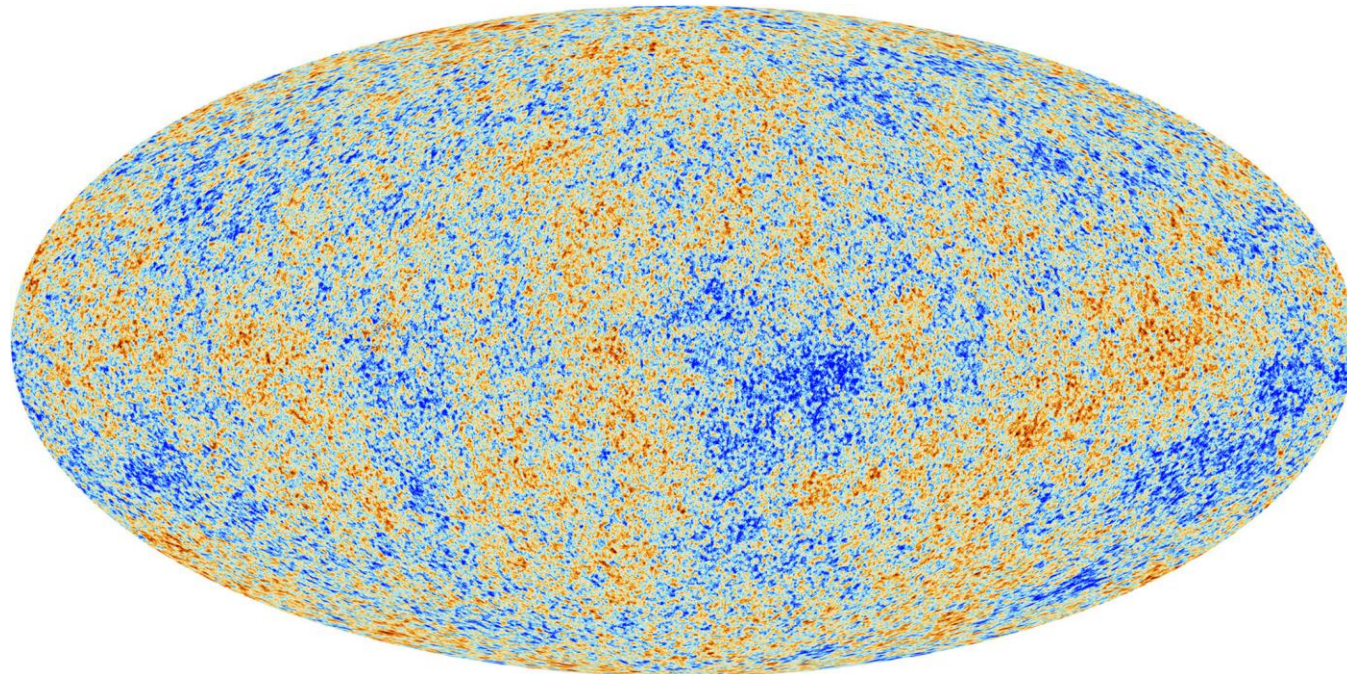
Cosmic strings

Primordial perturbations: cosmic inflation

Black holes from primordial perturbations

Cosmic inflation: quantum fluctuations

Later: strongest collapse into black holes



I. (Semi-)inflection point inflation

II. Stochastic inflation

III. Black hole statistics

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Single-field inflation is simple

Action:

$$S = \int d^4x \sqrt{-g} \left[\frac{1}{2} R - \frac{1}{2} \partial^\mu \varphi \partial_\mu \varphi - V(\varphi) \right]$$

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Background equations of motion:

$$\ddot{\varphi} + 3H\dot{\varphi} + V'(\varphi) = 0, \quad 3H^2 = \frac{1}{2}\dot{\varphi}^2 + V(\varphi)$$

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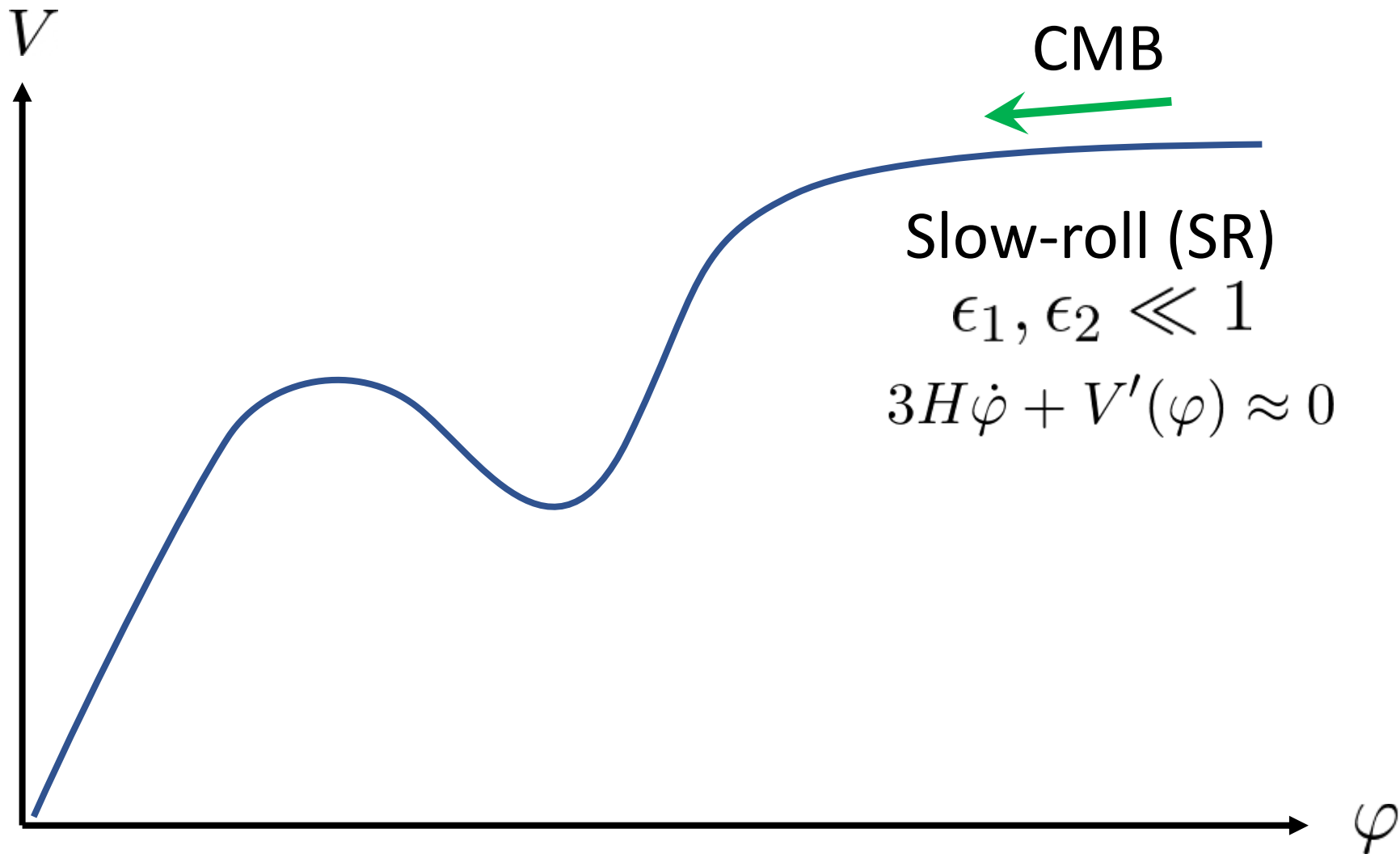
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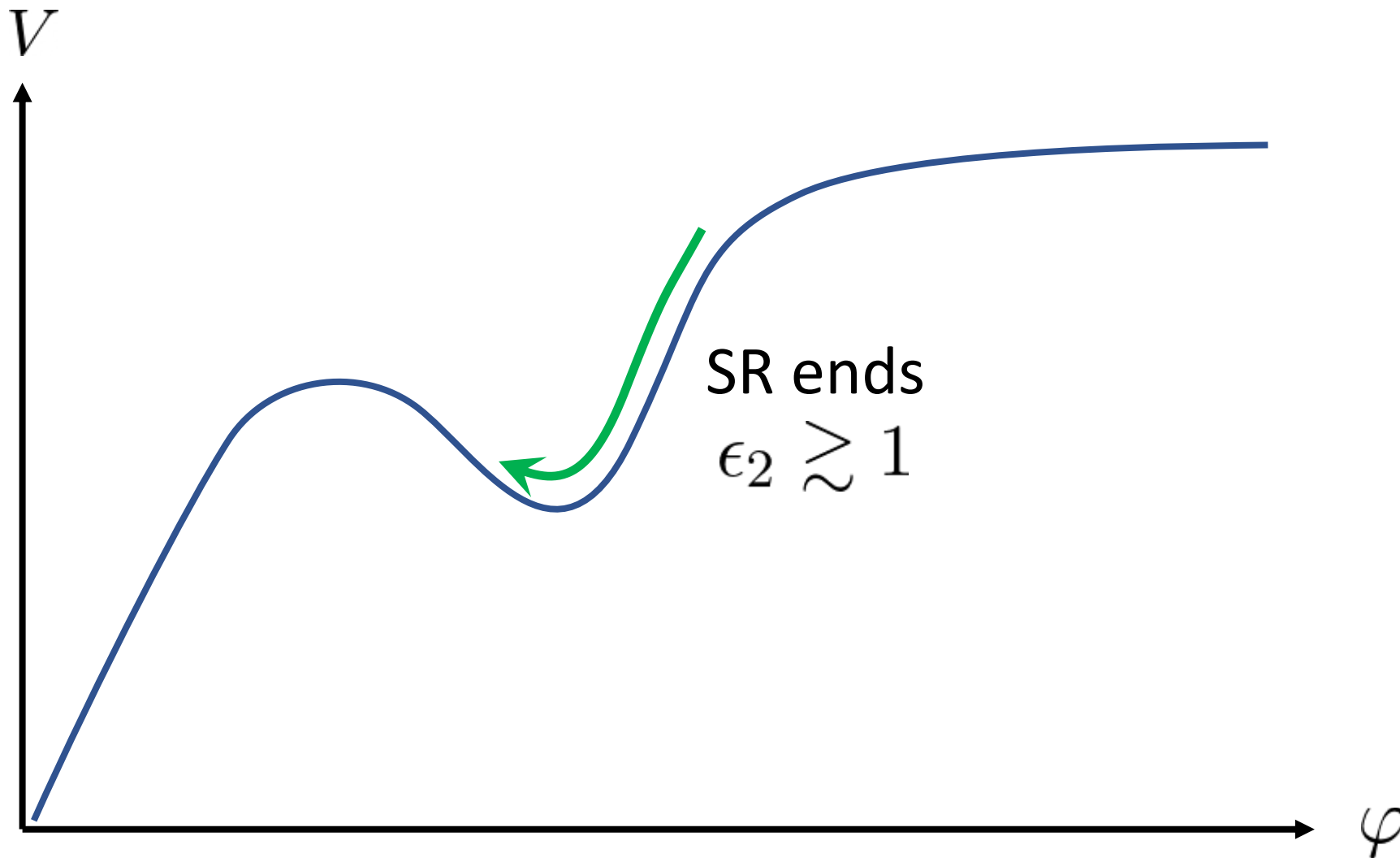
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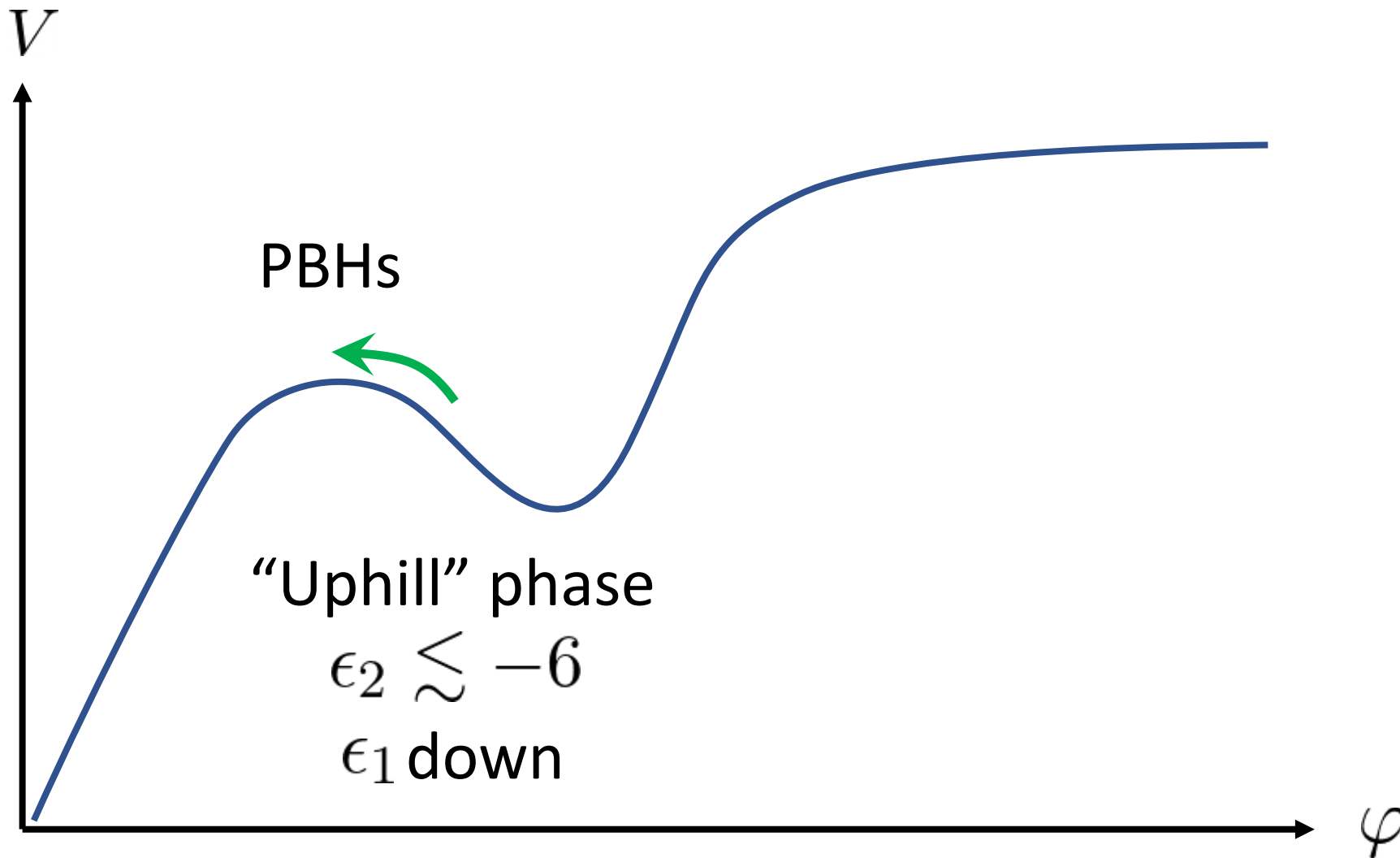
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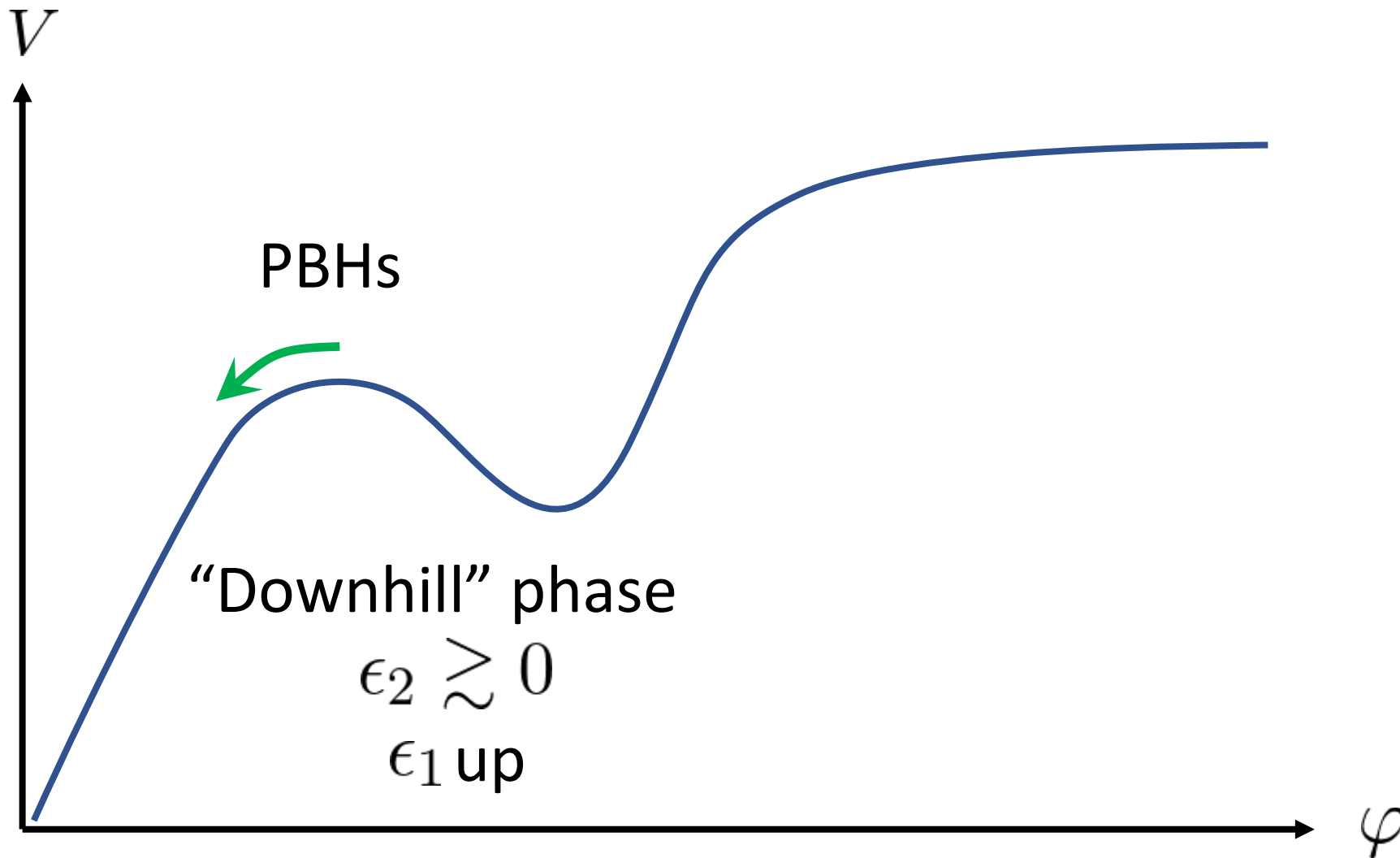
Slow-roll parameters:

$$\epsilon_1 \equiv -\partial_N \ln H, \quad \epsilon_2 \equiv \partial_N \ln \epsilon_1$$









Linear perturbations grow near feature

Comoving curvature perturbation $\mathcal{R} = \frac{\delta\varphi}{\sqrt{2\epsilon_1}}$

$$\ddot{\mathcal{R}}_k + H(3 + \epsilon_2)\dot{\mathcal{R}}_k + \frac{k^2}{a^2}\mathcal{R}_k = 0$$

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Vacuum initial conditions:

$$\mathcal{R}_k = \frac{1}{2a\sqrt{k\epsilon_1}}e^{ik/(aH)}$$

Late times:

$$\mathcal{R}_k \rightarrow \text{const. if } \epsilon_2 > -3$$

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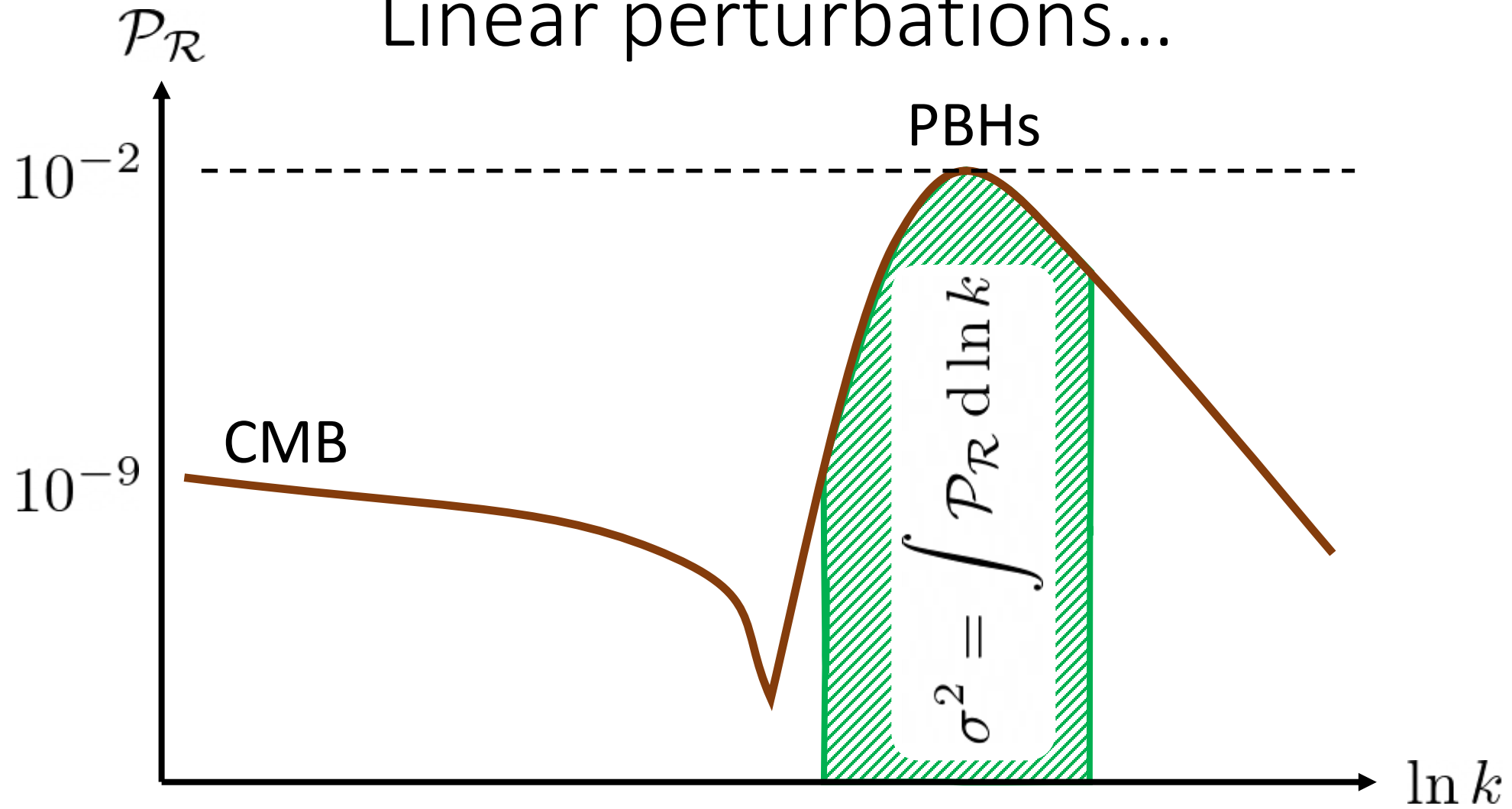
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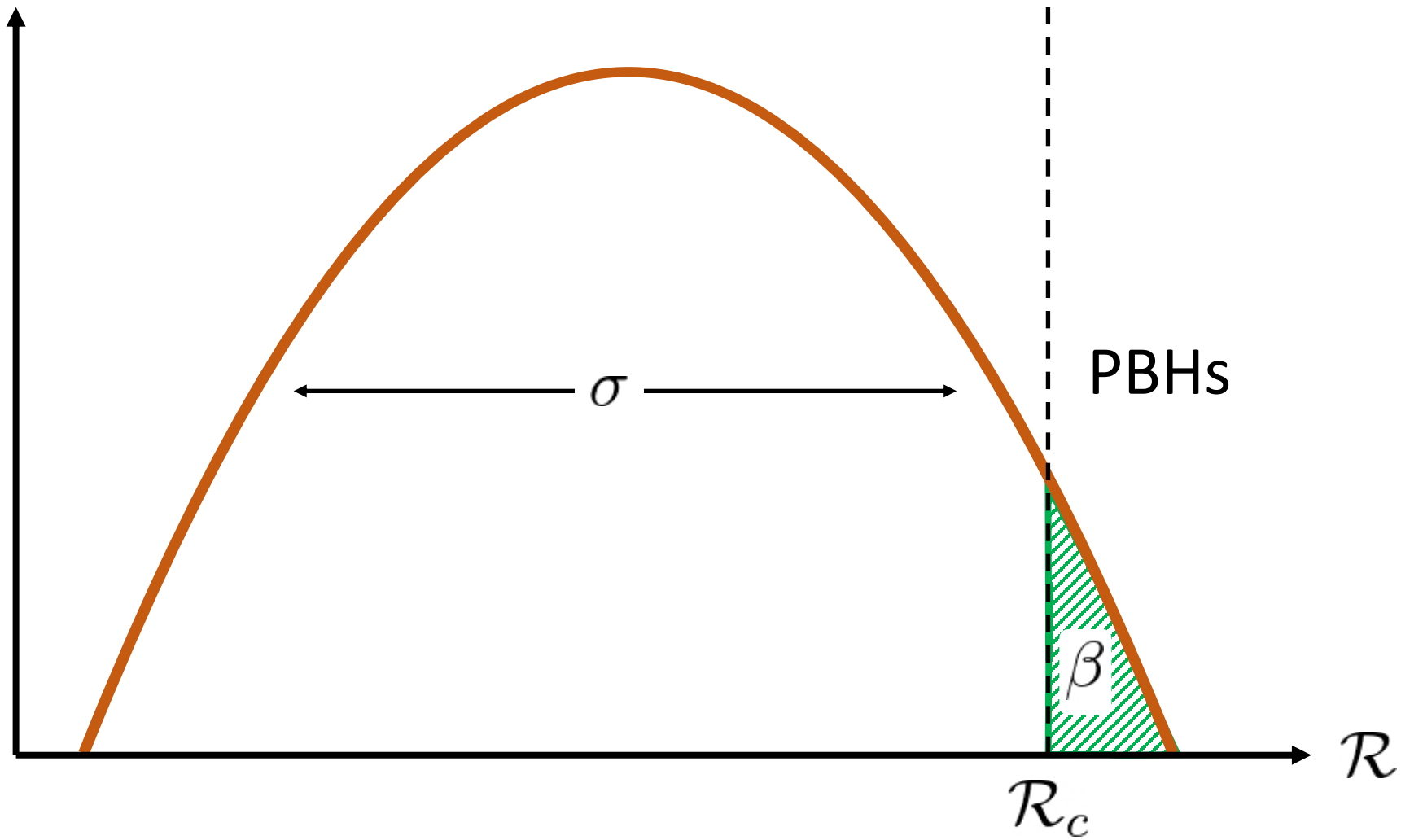
$$\mathcal{R}_k \rightarrow \text{const. if } \epsilon_2 > -3$$

Define power spectrum: $\mathcal{P}_{\mathcal{R}}(k) \equiv \frac{k^3}{2\pi^2}|\mathcal{R}_k|^2$

Linear perturbations...



$\log p(\mathcal{R})$...Gaussian distribution



Why this picture is wrong

\mathcal{R} is not the correct statistic for PBH formation

Perturbations in the tail are not Gaussian

I. (Semi-)inflection point inflation

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Approximations in two regimes

Sub-Hubble scales:

Linear perturbation theory good; neglect mode couplings

$$\delta\ddot{\varphi}_k + 3H\delta\dot{\varphi}_k + H^2 \left(\frac{k^2}{a^2 H^2} - \frac{3}{2}\epsilon_2 + \frac{1}{2}\epsilon_1\epsilon_2 - \frac{1}{4}\epsilon_2^2 - \frac{1}{2}\epsilon_2\epsilon_3 \right) \delta\varphi_k = 0$$

Super-Hubble scales:

Local FLRW equations good; neglect gradient terms

$$\ddot{\varphi} + 3H\dot{\varphi} + V'(\varphi) = 0$$

Approximations in two regimes

Inflaton field: $\varphi = \phi + \delta\phi$

Coarse-grained:
FLRW

Short-wavelength:
linear perturbation theory

$$\phi \equiv \int_{k < k_\sigma} \frac{d^3k}{(2\pi)^{2/3}} \varphi_k(N) e^{-i\vec{k}\cdot\vec{x}} \quad \delta\phi \equiv \int_{k > k_\sigma} \frac{d^3k}{(2\pi)^{2/3}} \varphi_k(N) e^{-i\vec{k}\cdot\vec{x}}$$

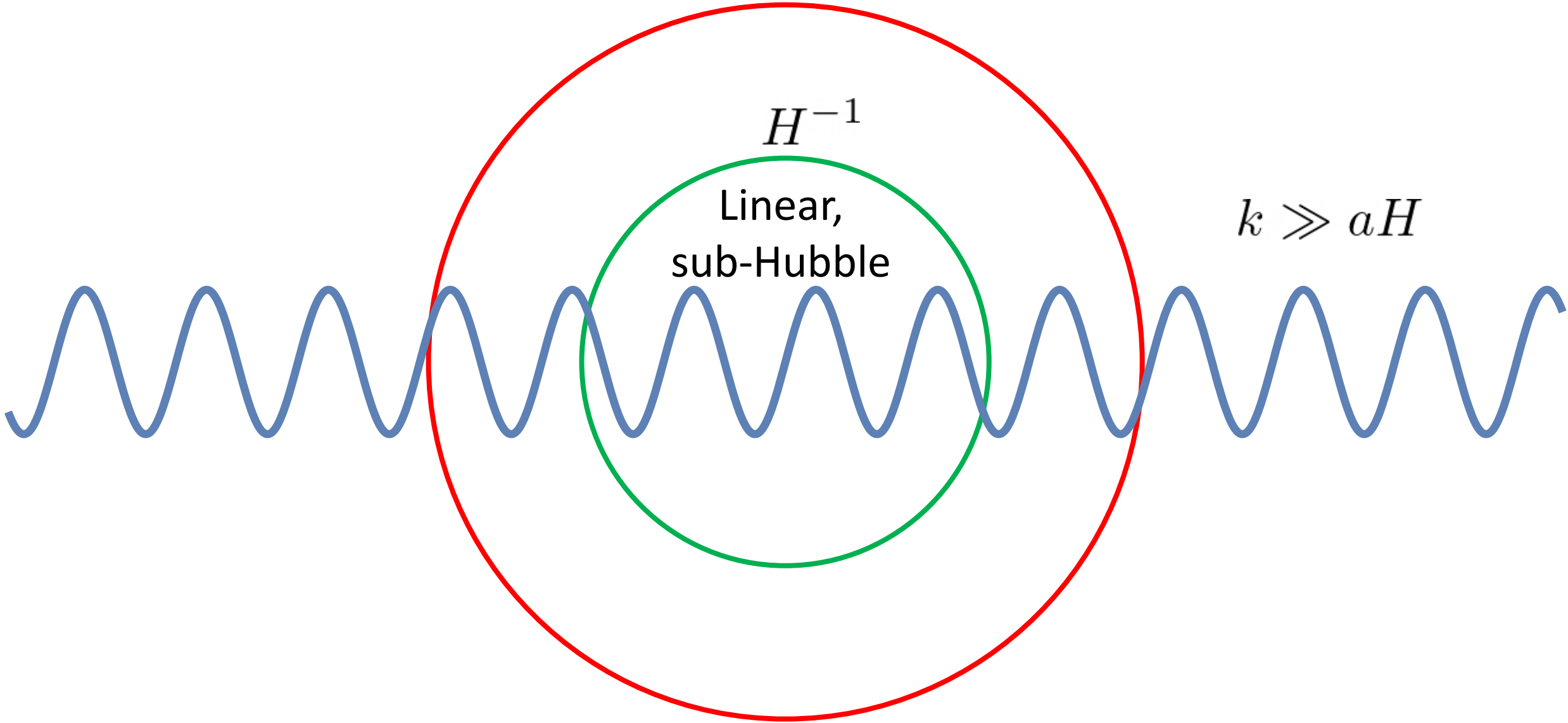
Patched together at the coarse-graining scale $k = k_\sigma \equiv \sigma aH$

$$(\sigma H)^{-1}$$

$$H^{-1}$$

Linear,
sub-Hubble

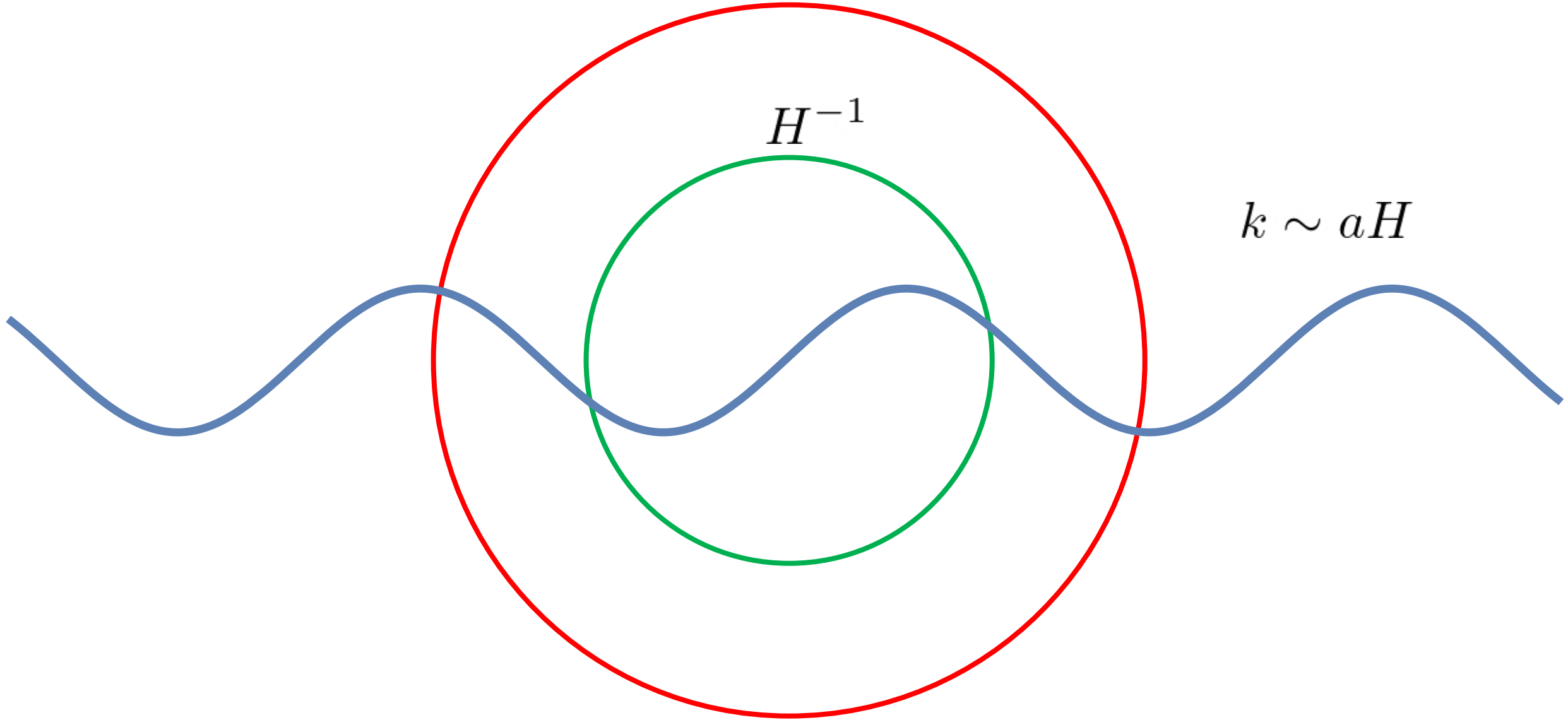
$$k \gg aH$$



$$(\sigma H)^{-1}$$

$$H^{-1}$$

$$k \sim aH$$



$$(\sigma H)^{-1}$$

$$H^{-1}$$

$$k = \sigma a H \ll a H$$

Coarse-graining exit:
Stochastic kick

Stochastic inflation

$$\phi' = \pi + \xi_\phi, \quad \pi' = - \left(3 - \frac{1}{2}\pi^2 \right) \pi - \frac{V'(\phi)}{H^2} + \xi_\pi, \quad H^2 = \frac{V(\phi)}{3 - \frac{1}{2}\pi^2}$$

$$\delta\phi_k'' = - \left(3 - \frac{1}{2}\pi^2 \right) \delta\phi_k' - \left[\frac{k^2}{a^2 H^2} + \pi^2 \left(3 - \frac{1}{2}\pi^2 \right) + 2\pi \frac{V'(\phi)}{H^2} + \frac{V''(\phi)}{H^2} \right] \delta\phi_k$$

$$\langle \xi_\phi(N) \xi_\phi(N') \rangle = \frac{1}{6\pi^2} \frac{dk_\sigma^3}{dN} |\delta\phi_{k_\sigma}(N)|^2 \delta(N - N')$$

$$\langle \xi_\pi(N) \xi_\pi(N') \rangle = \frac{1}{6\pi^2} \frac{dk_\sigma^3}{dN} |\delta\phi_{k_\sigma}'(N)|^2 \delta(N - N')$$

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$$\mathcal{R}_{<k} = \Delta N = N - \bar{N}$$

ΔN formalism

$$ds^2 = -dt^2 + a^2(t)e^{2\zeta(x,t)}dx^2$$

$$\Delta N \equiv N - \bar{N} = \mathcal{R} = \zeta$$

Stochastic ΔN formalism:

- solve stochastic system many times; include kicks up to scale k
- collect N on each run
- build statistics for coarse-grained curvature perturbation $\mathcal{R}_{<k}$

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Stochastic inflation

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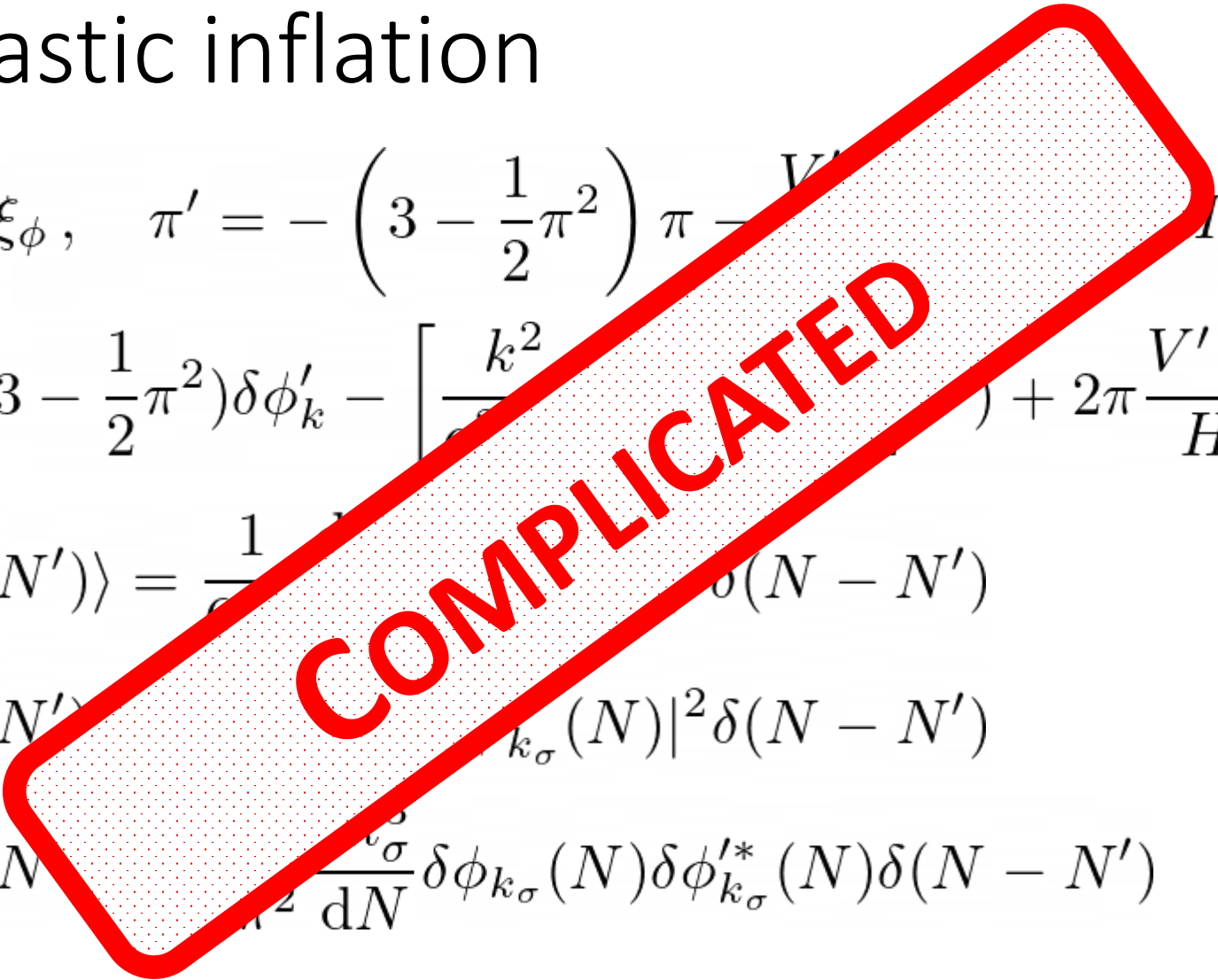
$$\delta\phi_k'' = -\left(3 - \frac{1}{2}\pi^2\right)\delta\phi_k' - \left[\frac{k^2}{\sigma^2} + 2\pi\frac{V'(\phi)}{H^2} + \frac{V''(\phi)}{H^2}\right]\delta\phi_k$$

$$\langle \xi_\phi(N)\xi_\phi(N') \rangle = \frac{1}{\sigma^2} \delta(N - N')$$

$$\langle \xi_\pi(N)\xi_\pi(N') \rangle = |k_\sigma(N)|^2 \delta(N - N')$$

$$\langle \xi_\phi(N)\xi_\pi(N') \rangle = \frac{v_\sigma}{H^2} \frac{dN}{dN} \delta\phi_{k_\sigma}(N)\delta\phi_{k_\sigma}'^*(N)\delta(N - N')$$

$$\mathcal{R}_{<k} = \Delta N = N - \bar{N}$$



How to move forward?

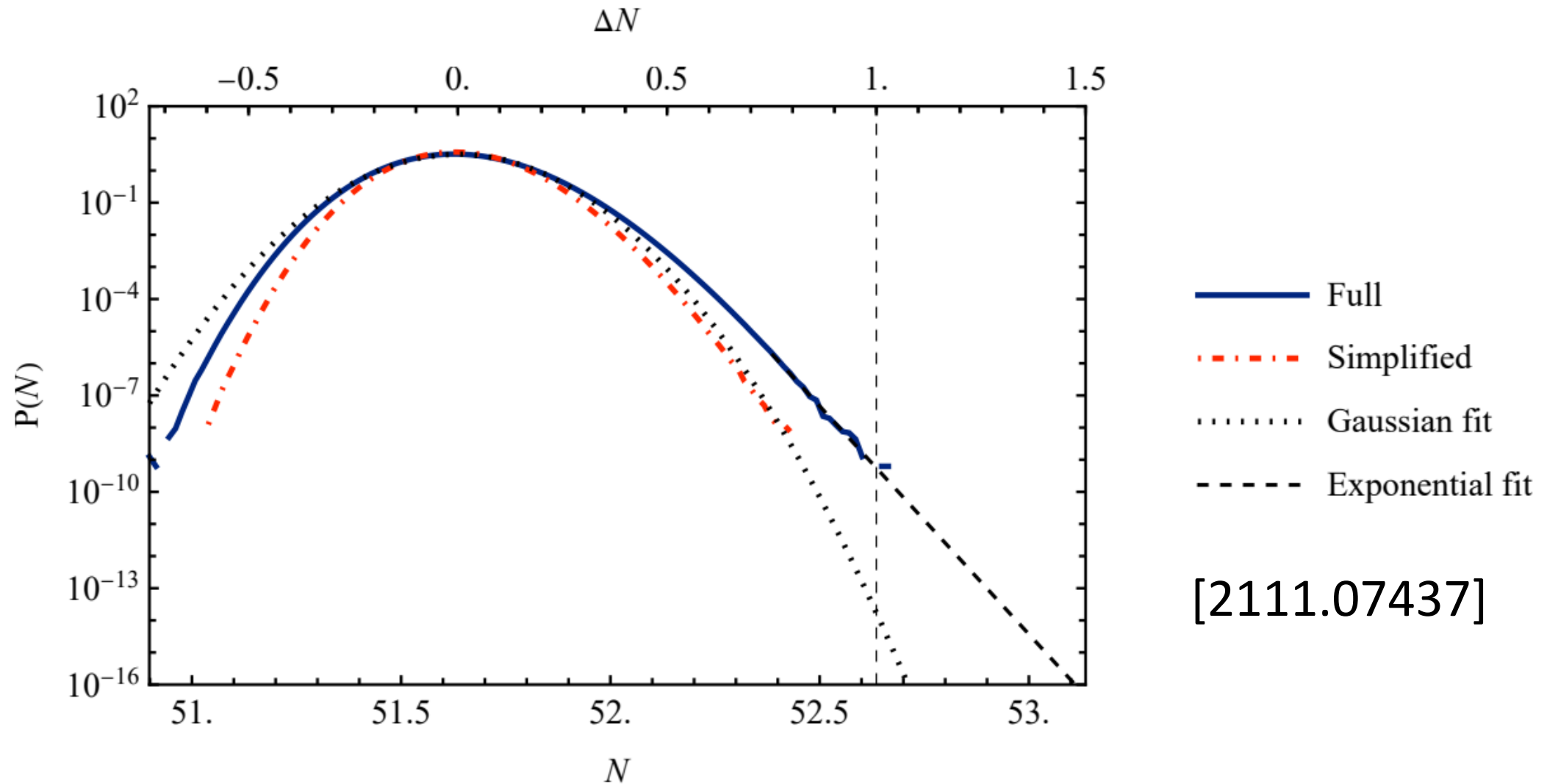
Analytical approximations?

$$\langle \xi_\phi(N) \xi_\phi(N') \rangle \approx \frac{H^2}{4\pi^2} \delta(N - N')$$

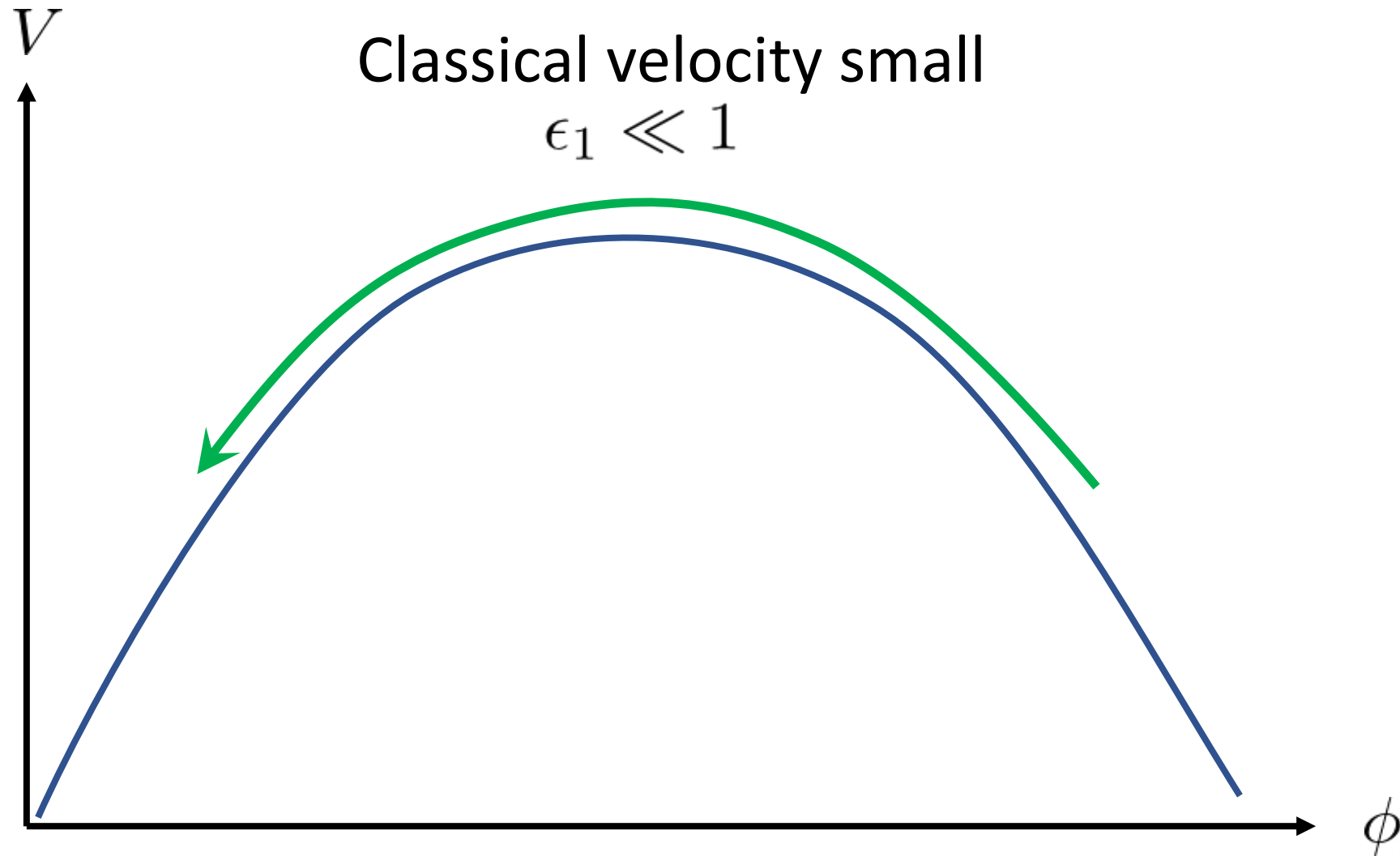
Full numerical computations?

Full numerical computations

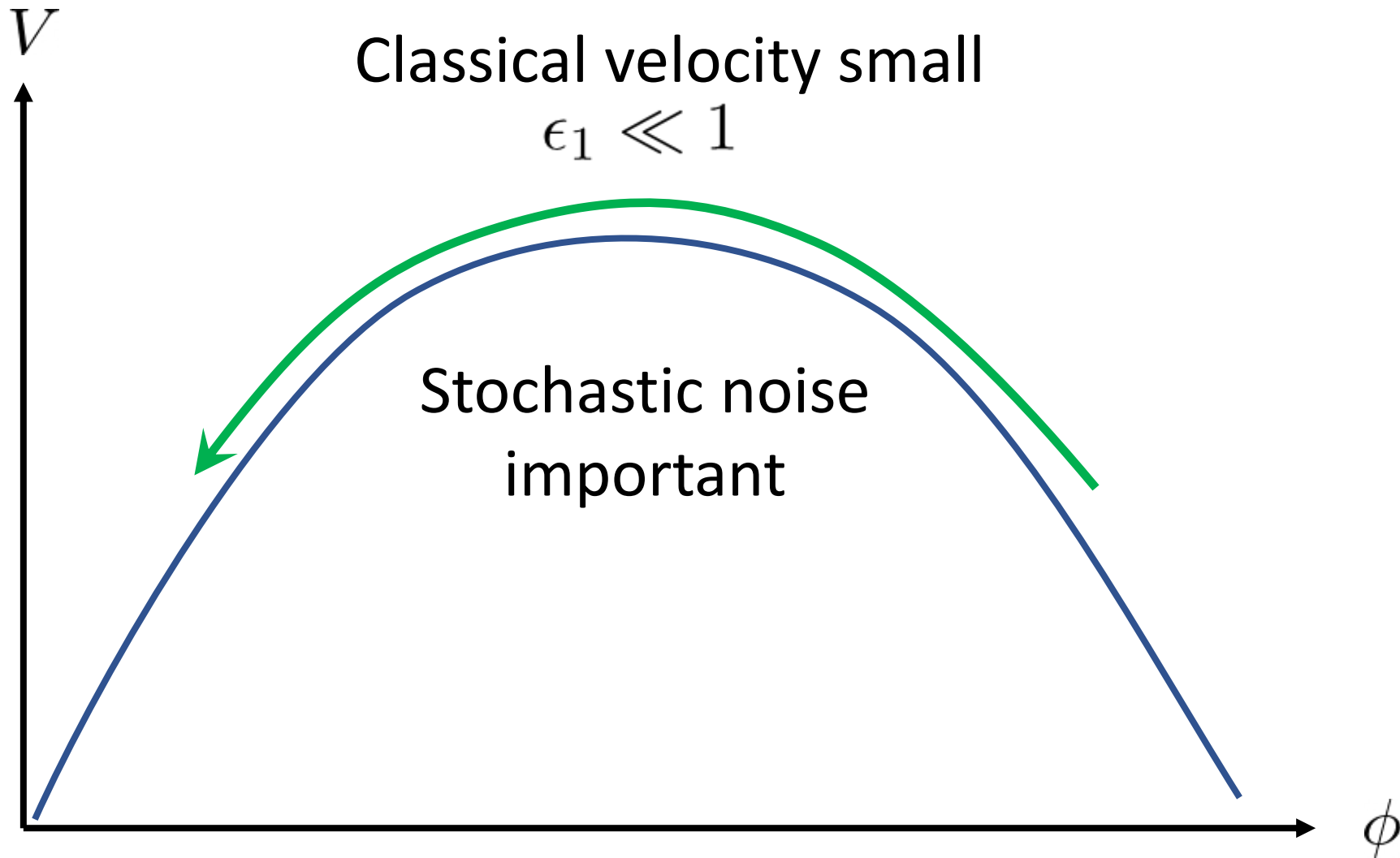
One million CPU hours



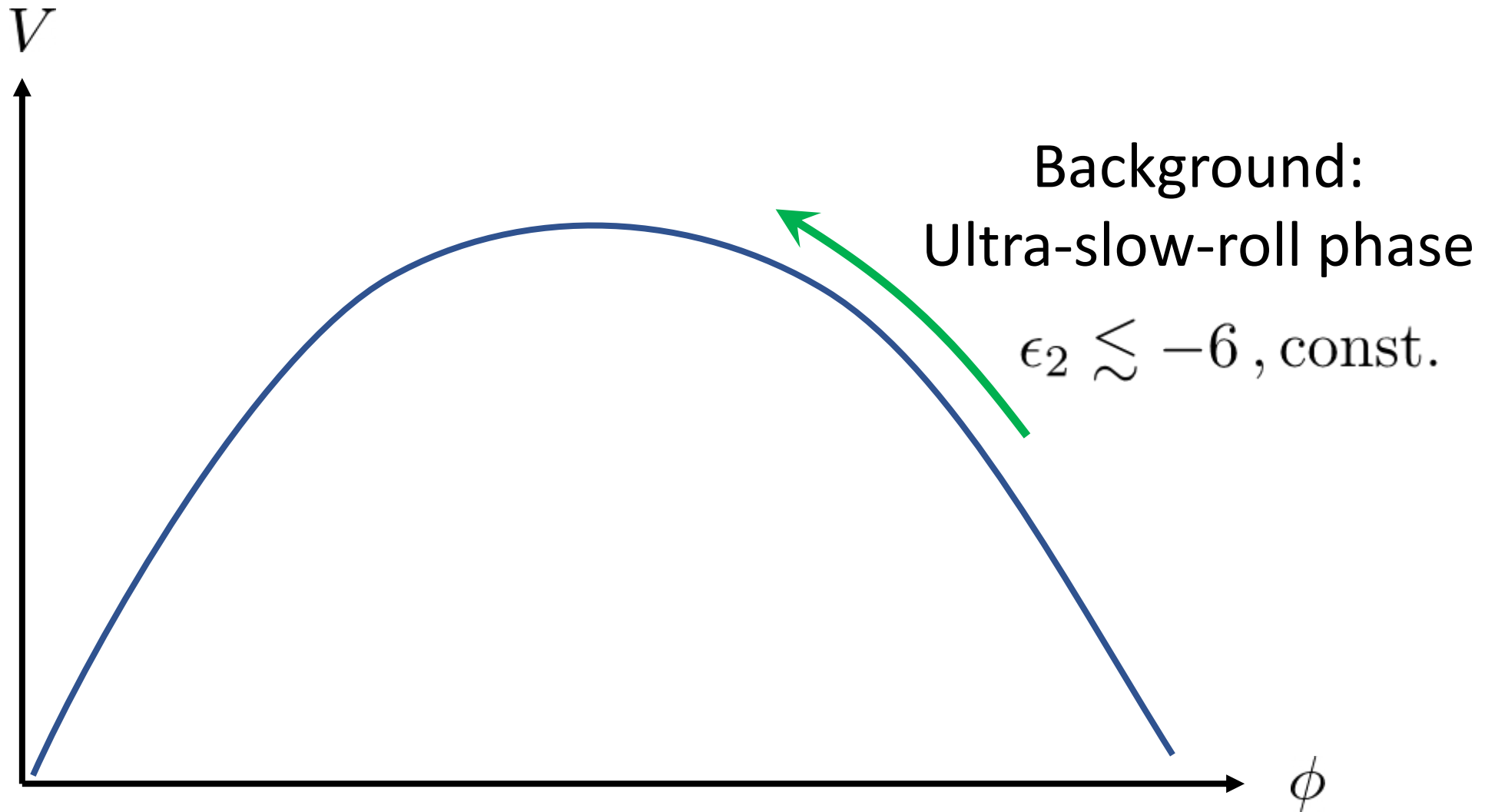
Zoom into the hilltop



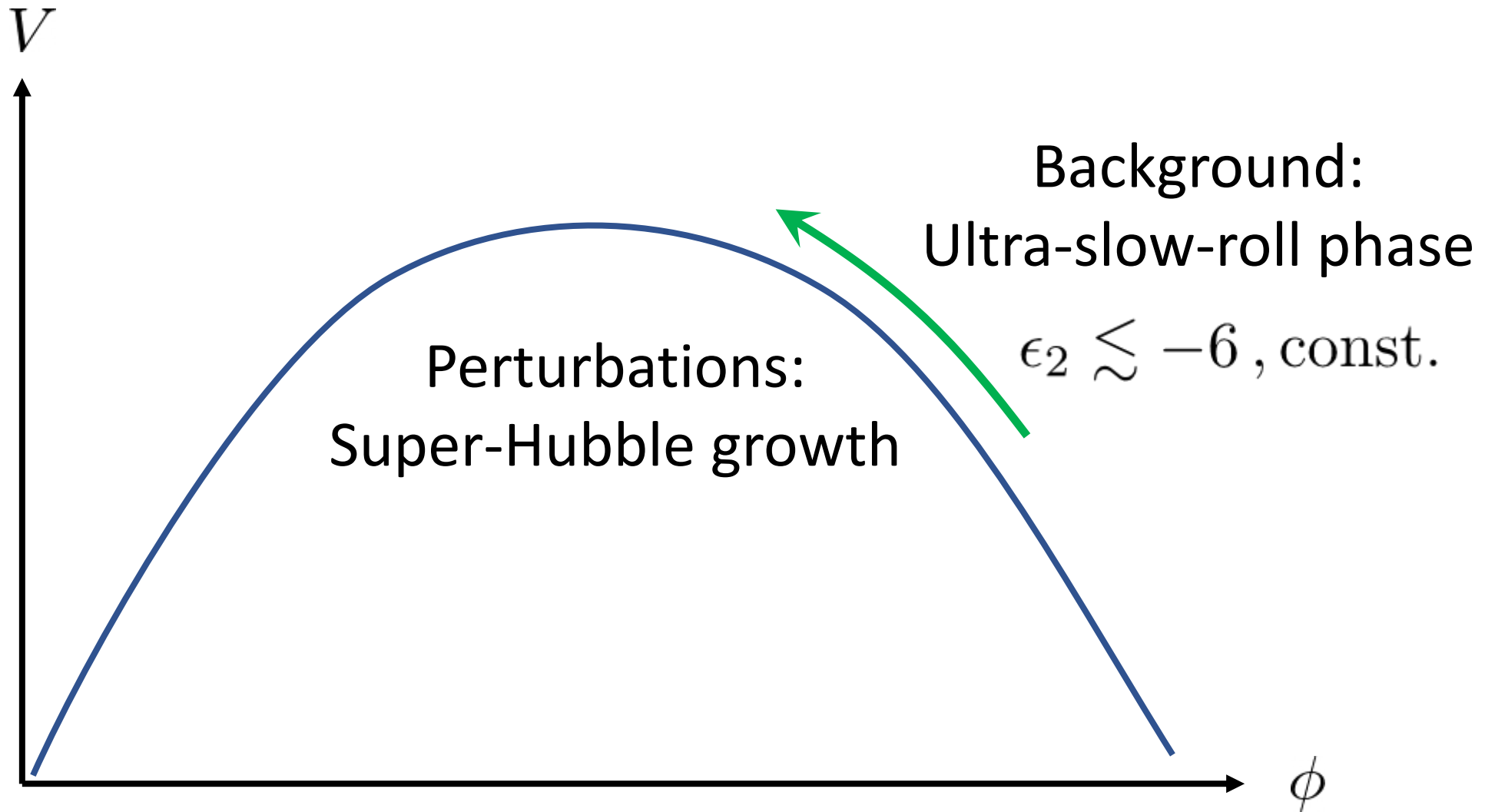
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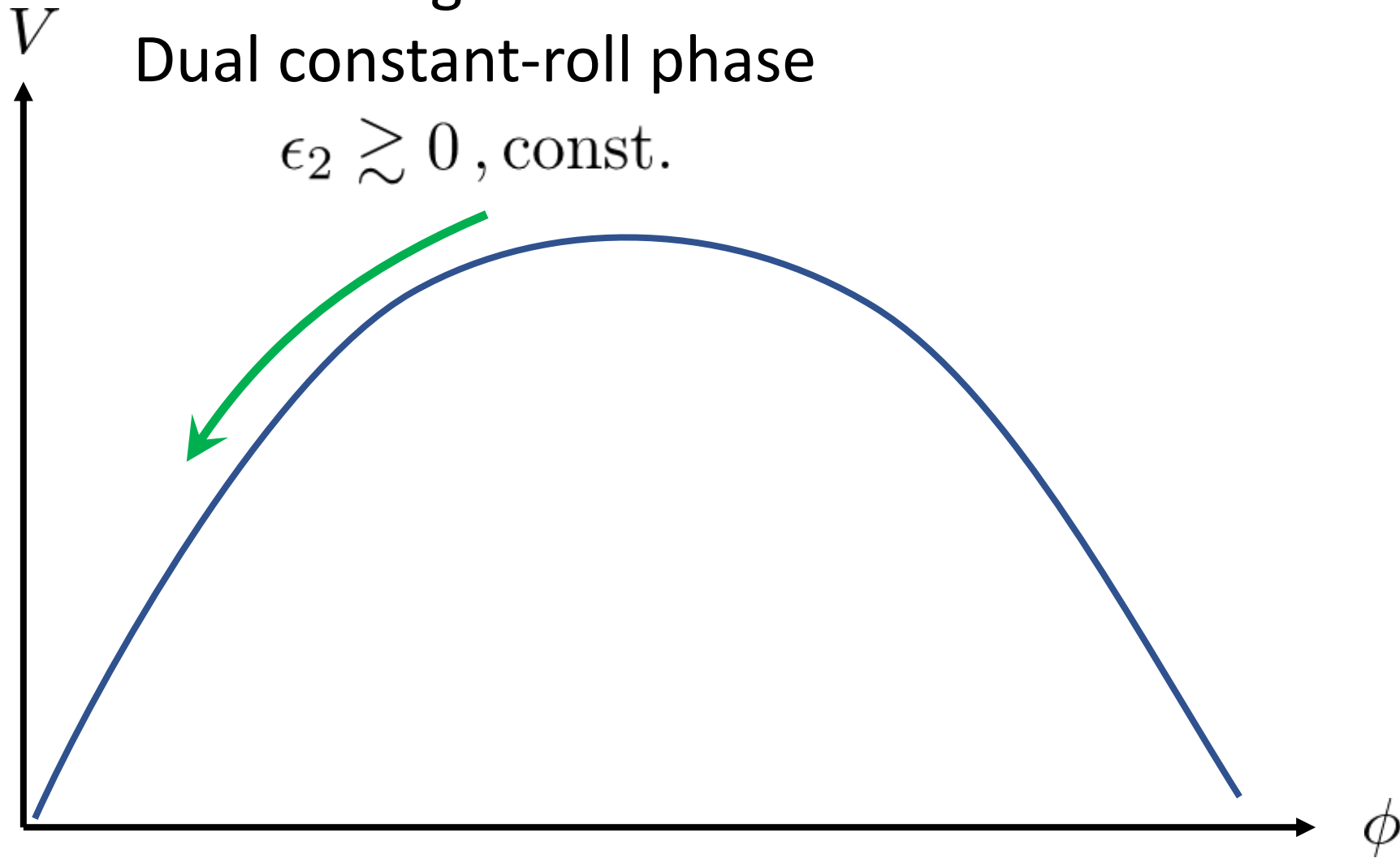


Zoom into the hilltop

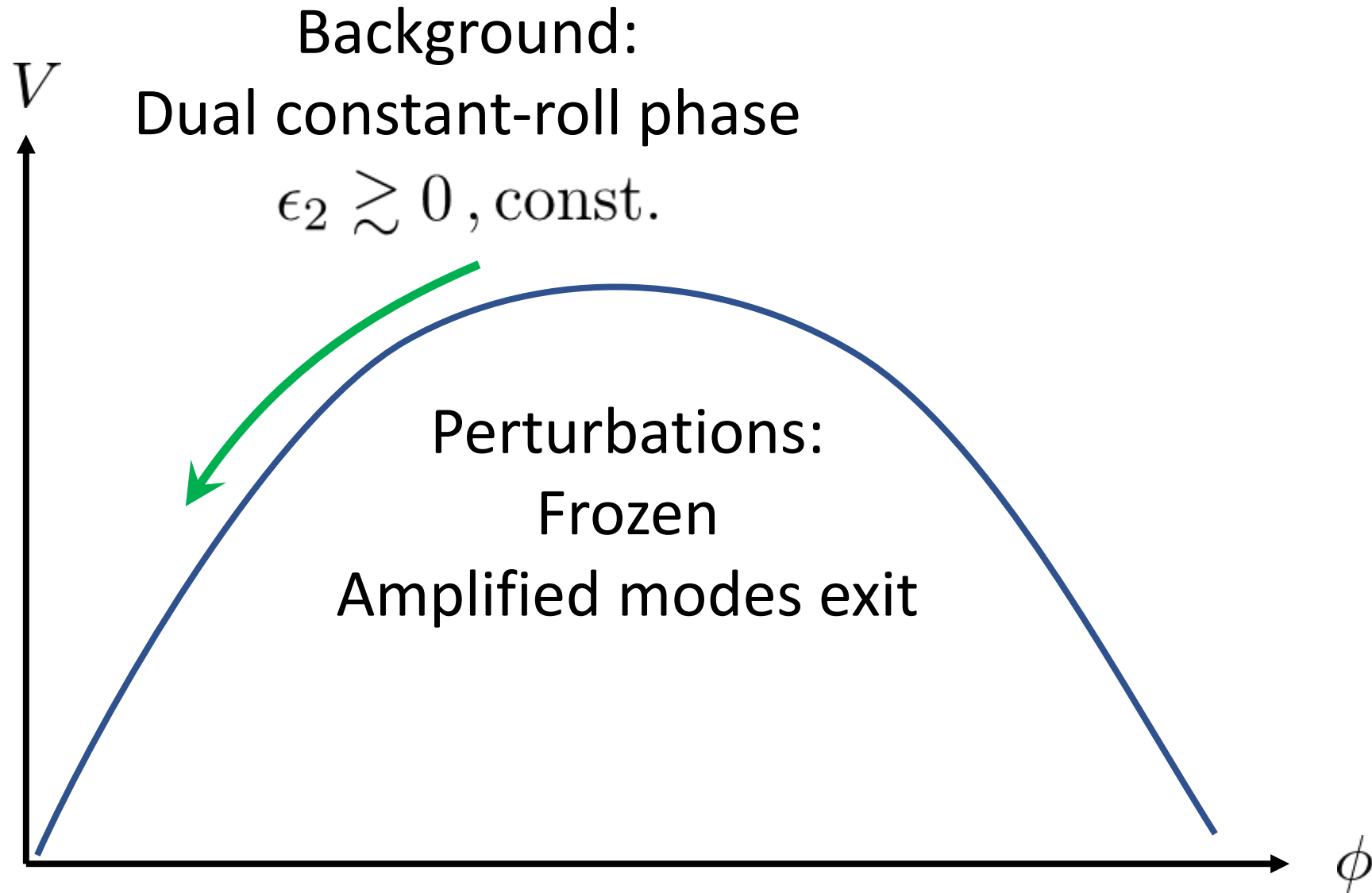
Background:

Dual constant-roll phase

$$\epsilon_2 \gtrsim 0, \text{ const.}$$



Zoom into the hilltop



Equations simplify in dual constant-roll phase

Adiabatic perturbations:


motion along classical trajectory only

Noise independent of background stochasticity:

pre-compute power spectrum

Simplified stochastic equation:


$$d\phi = \frac{\epsilon_2}{2}(\phi - \phi_0)dN + \frac{\epsilon_2}{2}\phi_0 e^{\frac{\epsilon_2}{2}N} \sqrt{\mathcal{P}_{\mathcal{R}}(k_\sigma)}dN \hat{\xi}_N$$


$$\langle \hat{\xi}_N \hat{\xi}_{N'} \rangle = \delta_{NN'}$$

Simplified stochastic equation:

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$$\phi(N) = \phi_0 \left(1 - e^{\frac{\epsilon_2}{2}N}\right) + \frac{\epsilon_2}{2}\phi_0 e^{\frac{\epsilon_2}{2}N} X_{<k_\sigma}$$

$$\langle \hat{\xi}_N \hat{\xi}_{N'} \rangle = \delta_{NN'}$$


$$X_{<k} \equiv \sum_{\tilde{k}=k_{\text{ini}}}^k \sqrt{\mathcal{P}_{\mathcal{R}}(\tilde{k})} d \ln k \hat{\xi}_{\tilde{k}}$$

ΔN distribution

$$p(X_{<k}) = \frac{1}{\sqrt{2\pi}\sigma_k} e^{-\frac{X_{<k}^2}{2\sigma_k^2}}, \quad \sigma_k^2 \equiv \int_{k_{\text{ini}}}^k \mathcal{P}_{\mathcal{R}}(\tilde{k}) d \ln \tilde{k}$$

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$$X_{<k} = \frac{2}{\epsilon_2} \left(1 - e^{-\frac{\epsilon_2}{2} \Delta N_{<k}} \right)$$

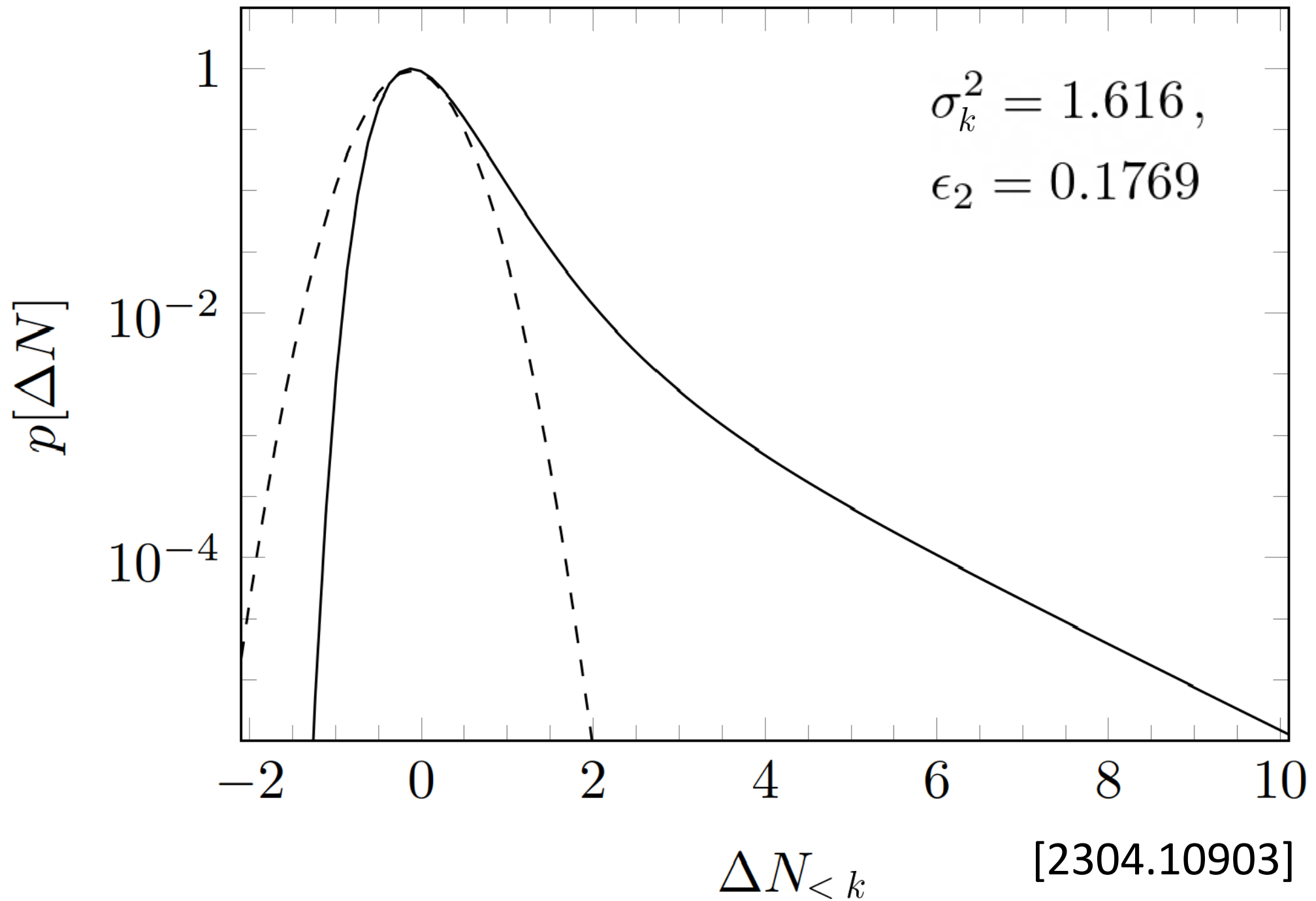
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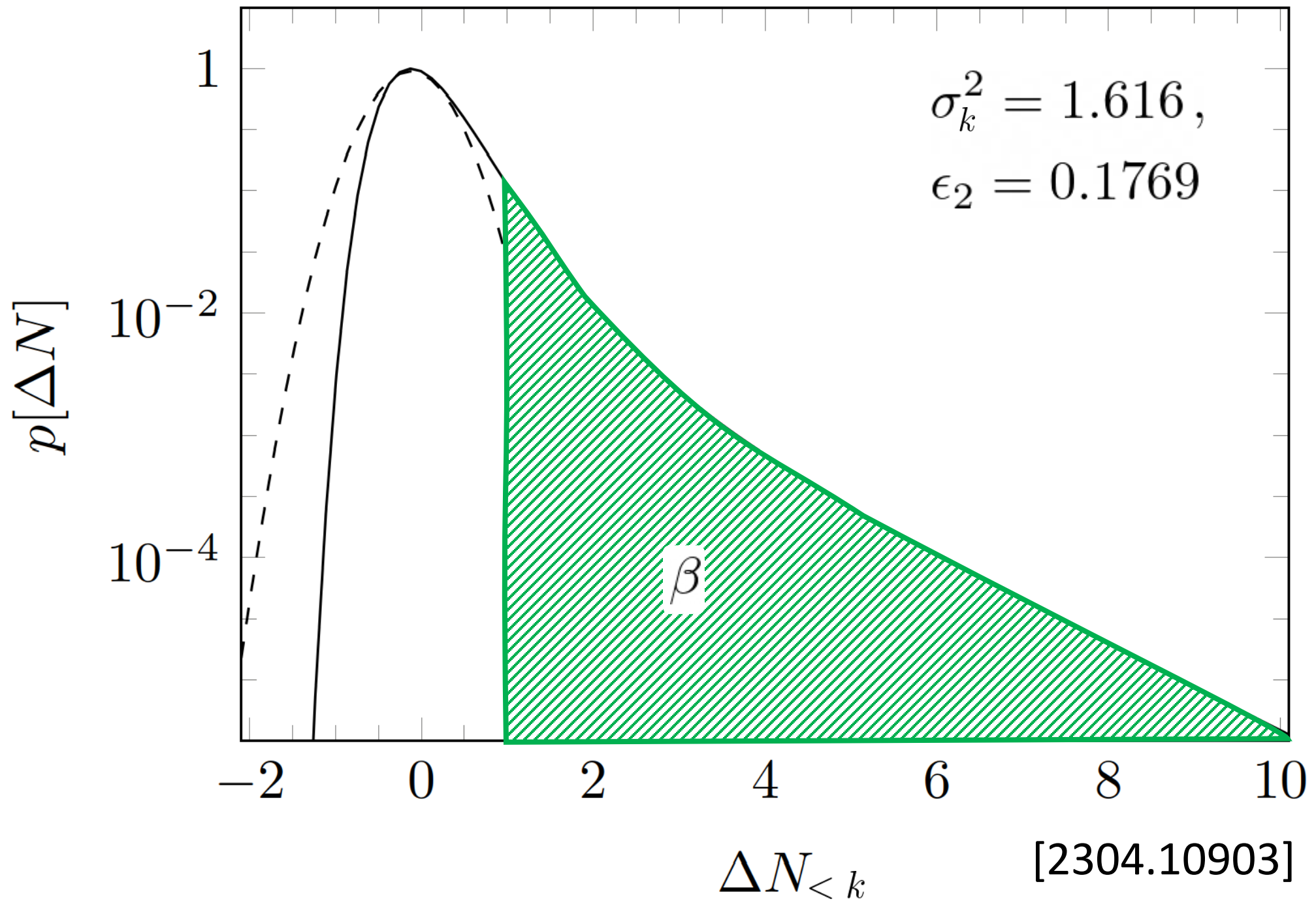
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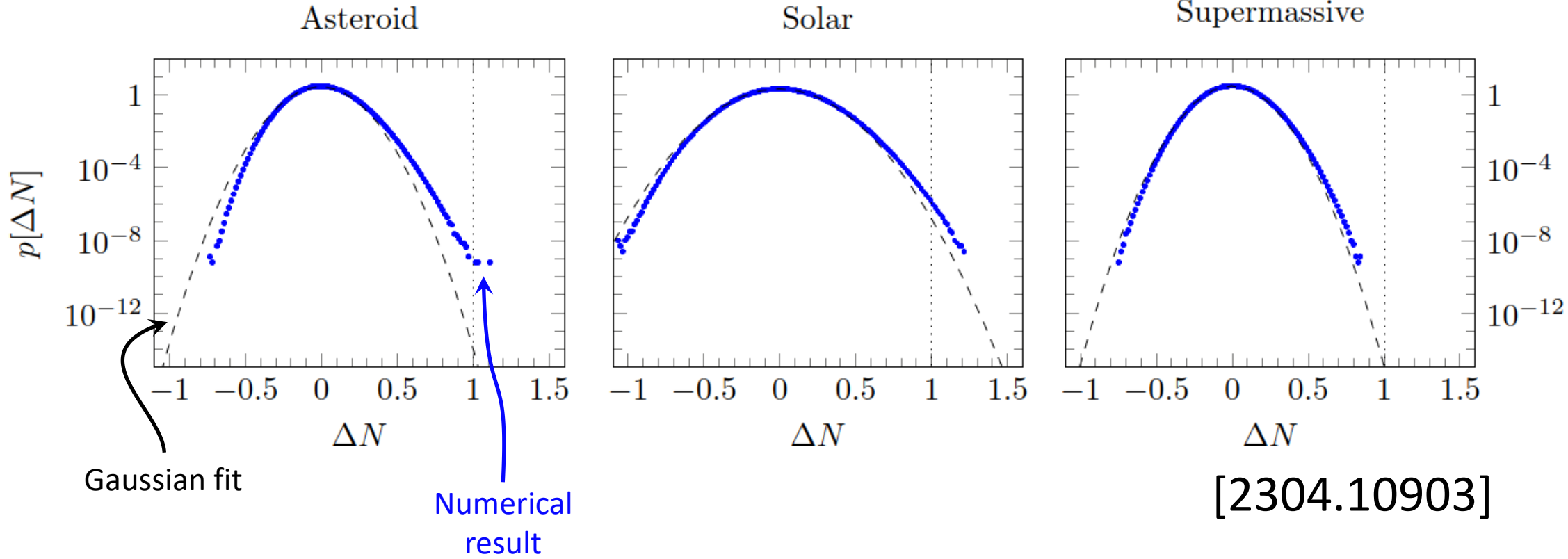
$$p(\Delta N_{<k}) = \frac{1}{\sqrt{2\pi}\sigma_k} \exp \left[-\frac{2}{\sigma_k^2 \epsilon_2^2} \left(1 - e^{-\frac{\epsilon_2}{2} \Delta N_{<k}} \right)^2 - \frac{\epsilon_2}{2} \Delta N_{<k} \right]$$

$$\Delta N_{<k} = \mathcal{R}_{<k}$$

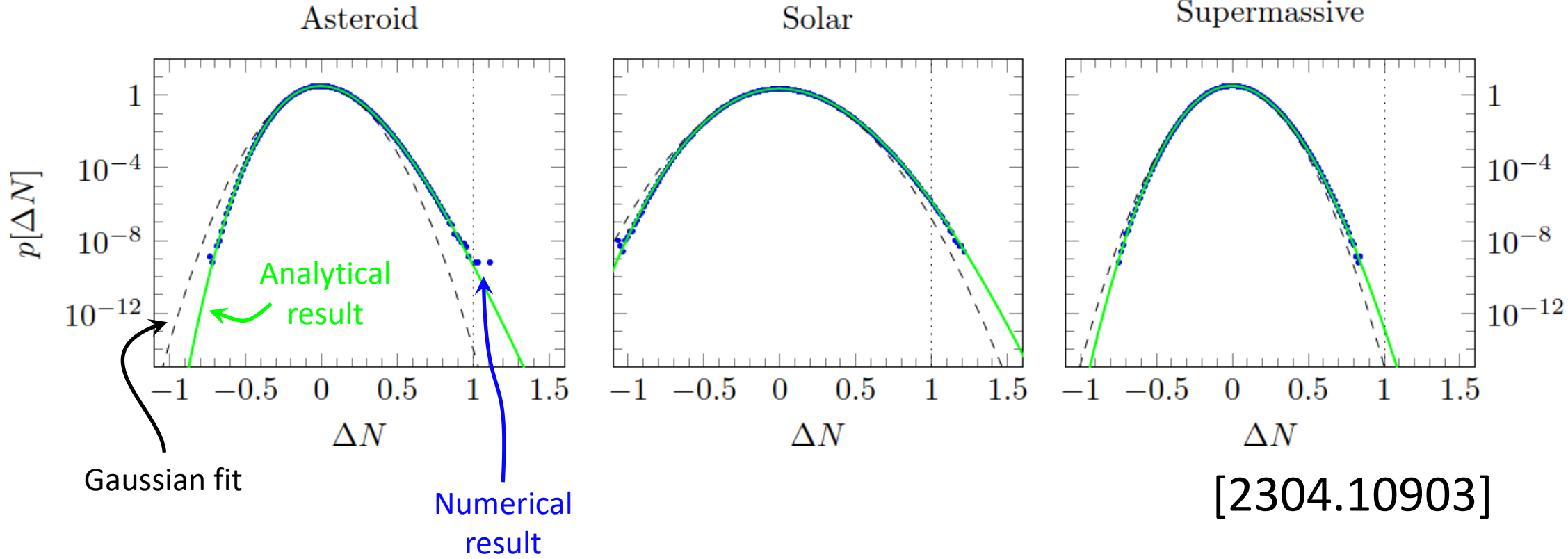




Comparison to numerics



Comparison to numerics



I. (Semi-)inflection point inflation

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Compaction function: right tool for determining the collapse threshold

$$C \equiv 2 \frac{M_{\text{MS}} - M_{\text{bg}}}{R}$$

Collapse: $C_{\text{max}} > C_c \approx 0.4$

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$$\mathcal{C} \equiv 2 \frac{M_{\text{MS}} - M_{\text{bg}}}{R}$$

Collapse: $\mathcal{C}_{\text{max}} > \mathcal{C}_c \approx 0.4$

In inflationary variables:

$$\mathcal{C}(r) = \frac{2}{3} (1 - [1 + r\zeta'(r)]^2)$$

Assume spherical symmetry

$$r\zeta'(r) = \sum_k \frac{2k^2 dk}{\sqrt{2\pi}} \zeta_k \left[\cos(kr) - \frac{\sin(kr)}{kr} \right]$$

$$\zeta_k = \frac{\sqrt{2\pi}}{2k^3} \frac{d\zeta_{<k}}{d \ln k}$$

Assume spherical symmetry

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$$\zeta_k = \frac{\sqrt{2\pi}}{2k^3} \frac{d\zeta_{<k}}{d \ln k}$$

Vary k:
Full profile
in one patch of space!



Recall: in the stochastic picture,

$$\zeta_{<k} = \Delta N_{<k} = -\frac{2}{\epsilon_2} \ln \left(1 - \frac{\epsilon_2}{2} X_{<k} \right) = -\frac{2}{\epsilon_2} \ln \left(1 - \frac{\epsilon_2}{2} \sum_{\tilde{k}=k_{\text{ini}}}^k \sqrt{\mathcal{P}_{\mathcal{R}}(\tilde{k})} d \ln \tilde{k} \hat{\xi}_{\tilde{k}} \right)$$

Master formula

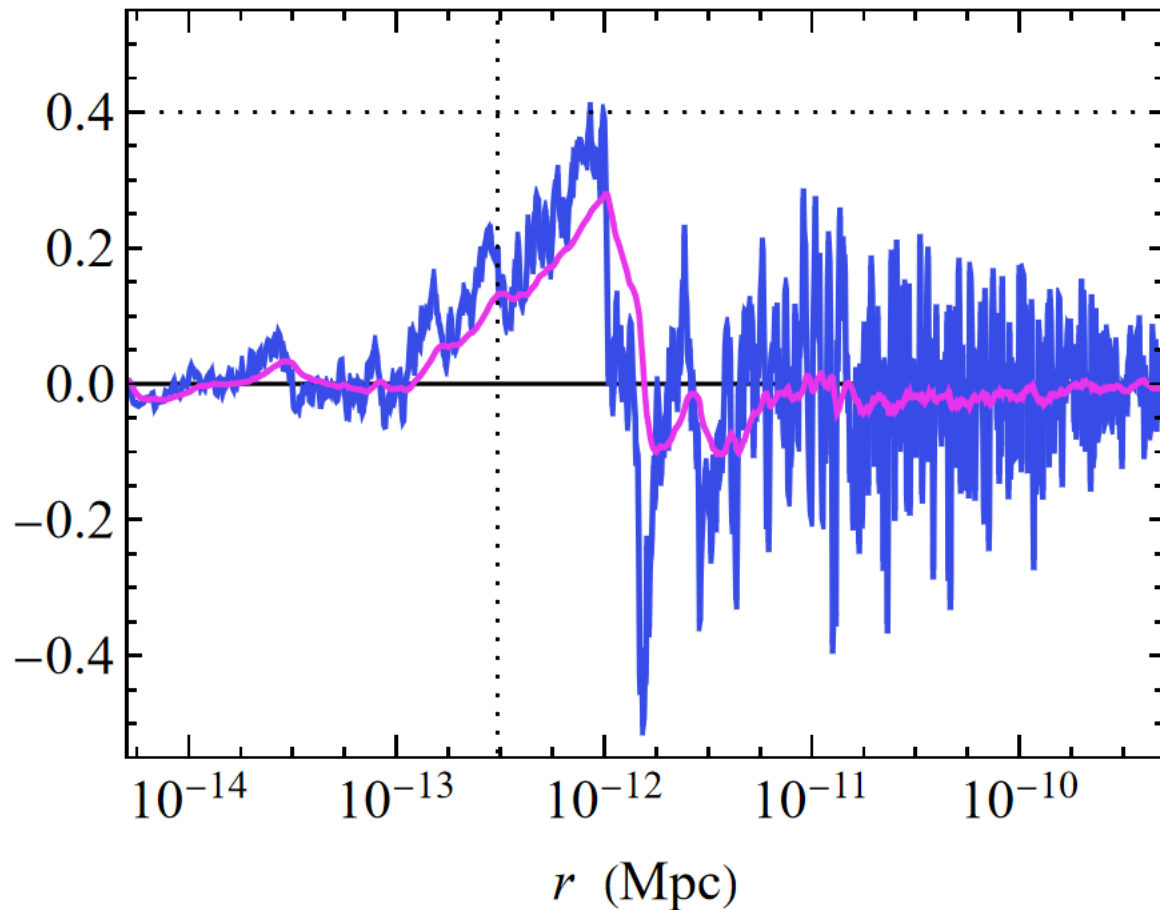
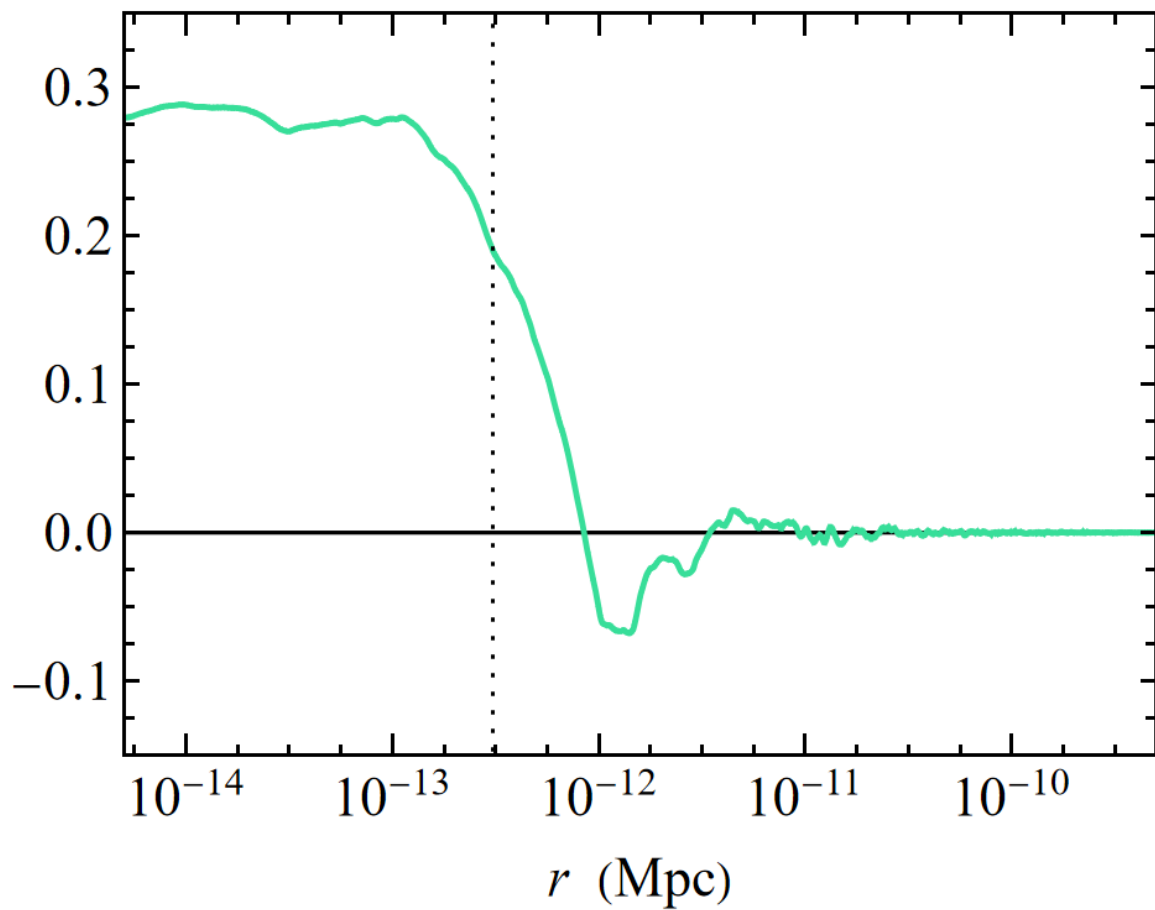
$$r\zeta'(r) = \sum_k \left[- \frac{\hat{\xi}_k}{1 - \frac{\epsilon_2}{2} X_{<k}} \sqrt{\mathcal{P}_\zeta(k)} d \ln k \right. \\ \left. + \frac{\epsilon_2}{4 \left(1 - \frac{\epsilon_2}{2} X_{<k}\right)^2} \mathcal{P}_\zeta(k) d \ln k \right] \\ \times \left[\cos(kr) - \frac{\sin(kr)}{kr} \right]$$

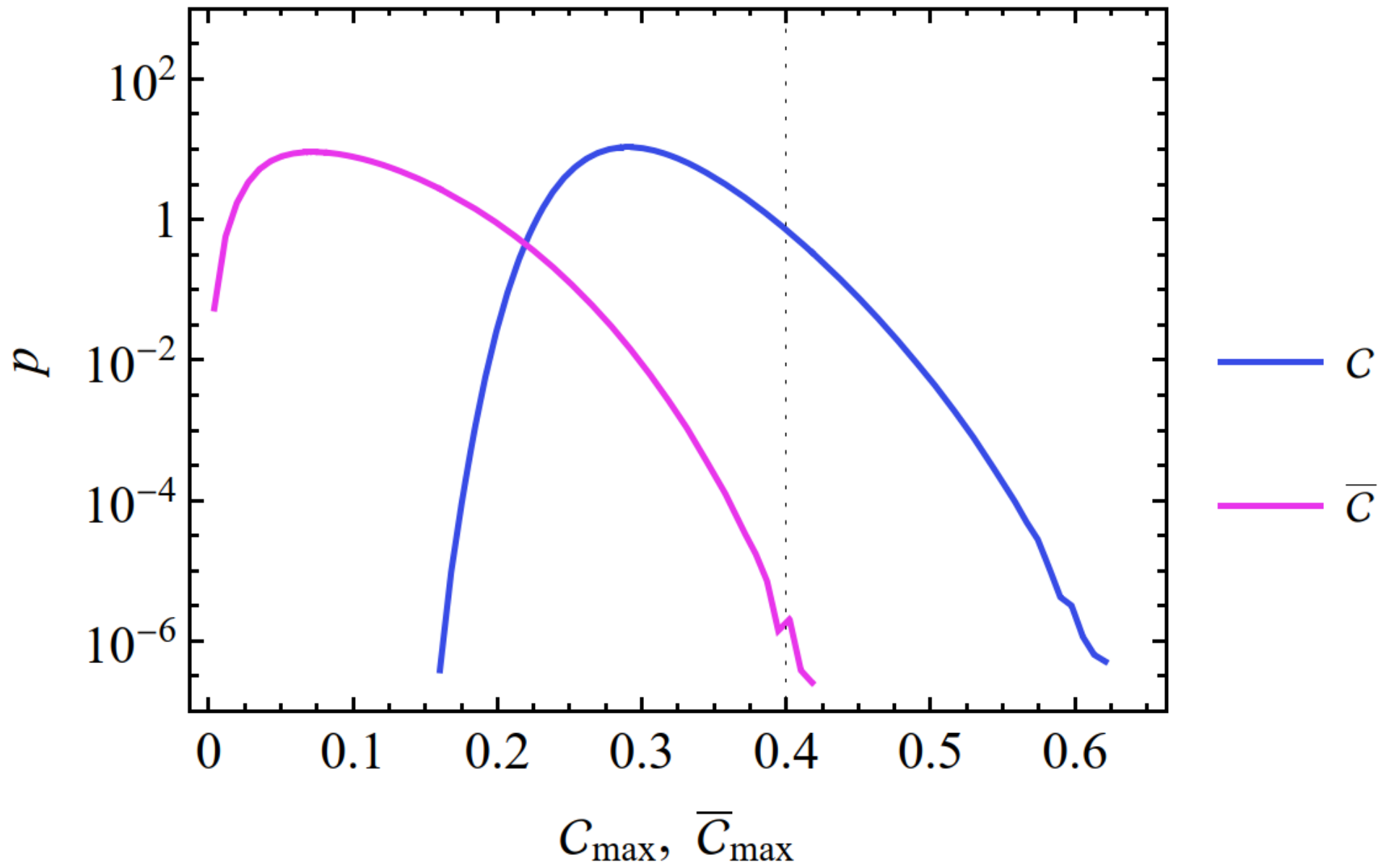
Alternative collapse measure:
averaged compaction function

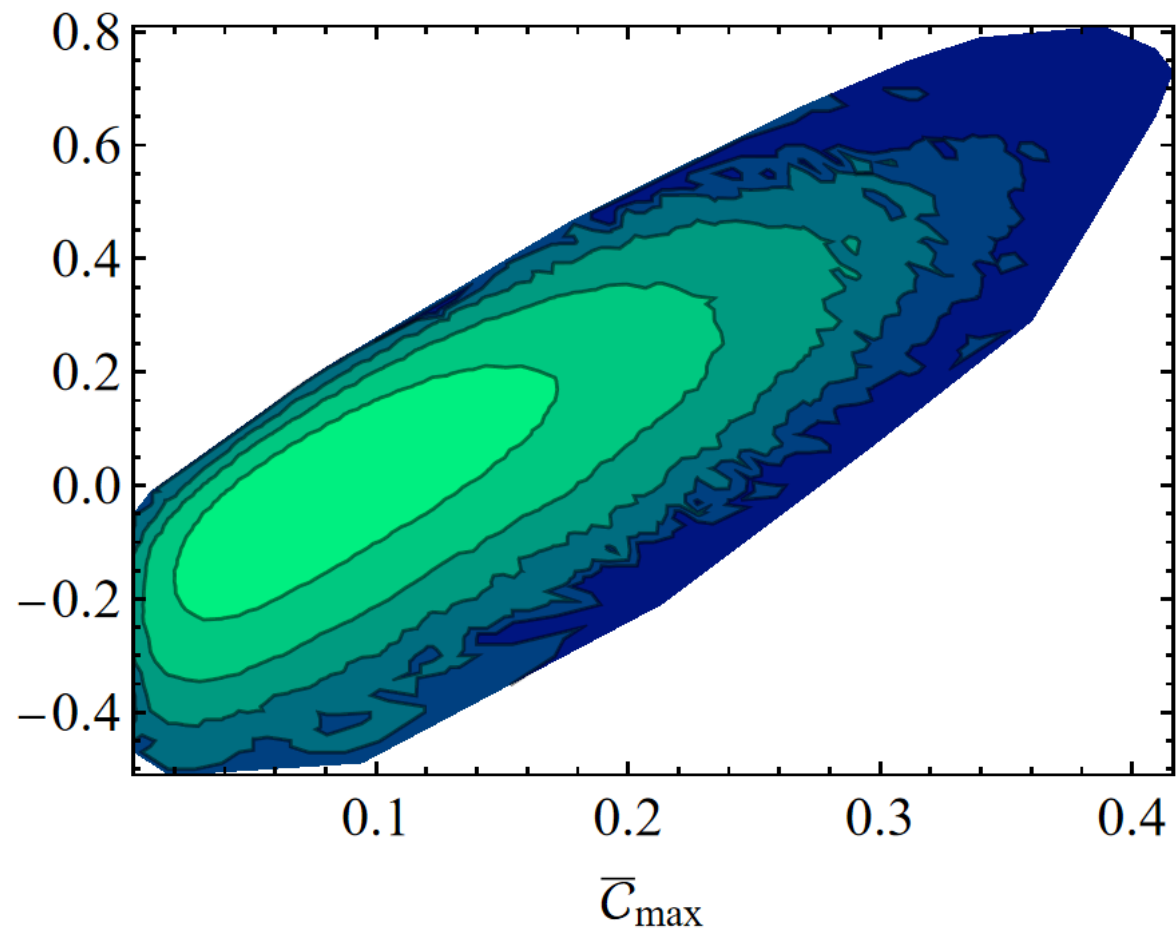
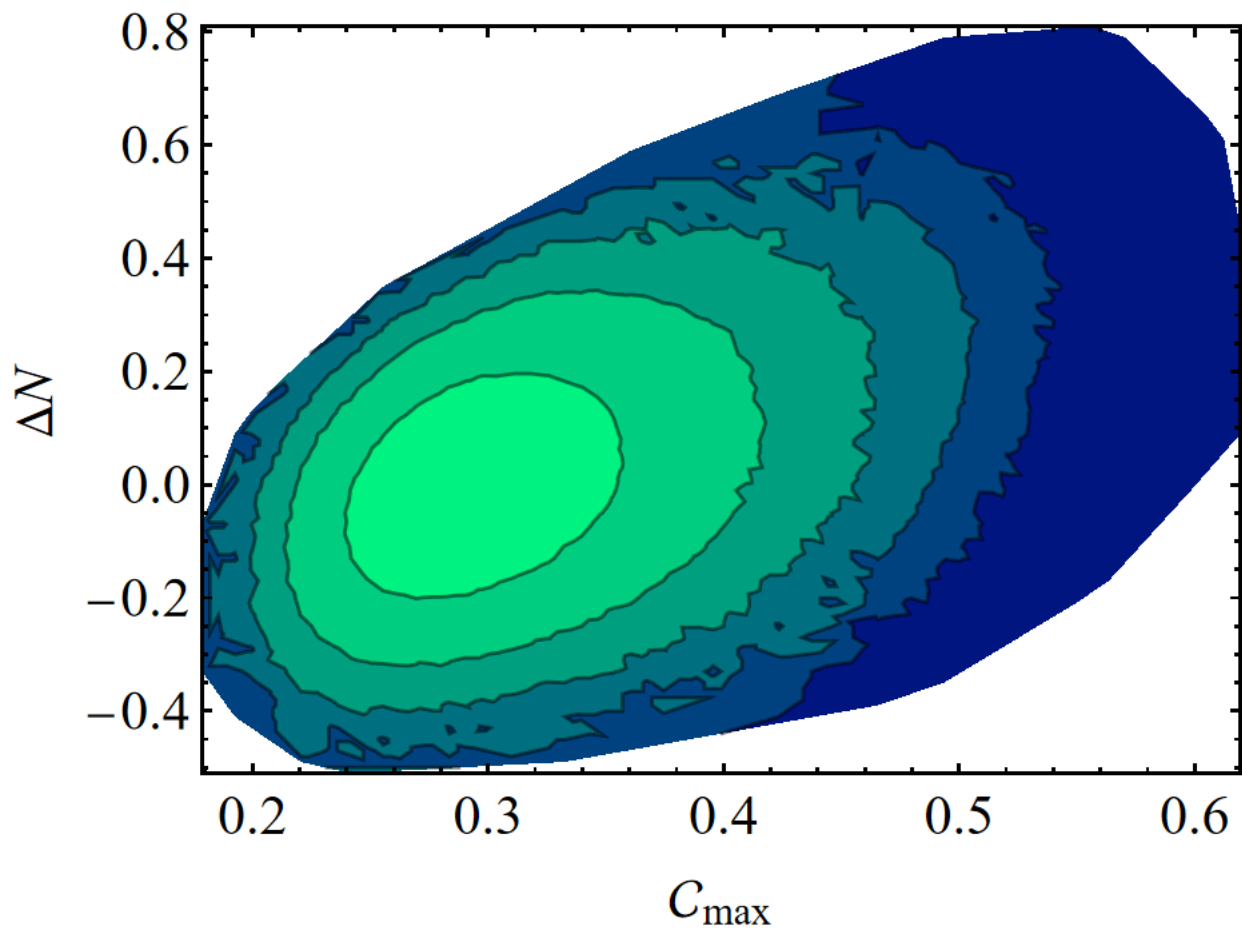
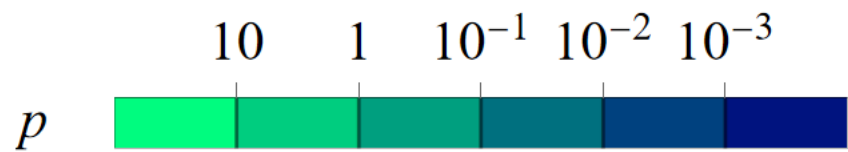
$$\begin{aligned}\bar{\mathcal{C}}(r) &\equiv \frac{3}{R(r)^3} \int_0^{R(r)} d\tilde{R} \tilde{R}^2 \mathcal{C} \\ &= -\frac{2}{r^3 e^{3\zeta(r)}} \int_0^r d\tilde{r} \tilde{r}^2 e^{3\zeta} [2\tilde{r}\zeta' + 3(\tilde{r}\zeta')^2 + (\tilde{r}\zeta')^3]\end{aligned}$$

$R = a r e^\zeta$

— ζ — c — \bar{c}







Initial PBH fractions

Gaussian approximation, $\mathcal{R}_{<k} > 1$, fixed k : $\beta \approx 5 \times 10^{-16}$

Non-Gaussian statistics, $\mathcal{R}_{<k} > 1$, fixed k : $\beta \approx 2.2 \times 10^{-11}$

$\bar{\mathcal{C}}_{\max} > 0.4$: $\beta \approx 1.4 \times 10^{-8}$

$\mathcal{C}_{\max} > 0.4$: $\beta \approx 0.016$

Problems

Collapse simulations have smooth peaks.

Us: Stochastic peaks?

- Numeric converge: ok
- Physics? Smoothing? Window functions?

Multiple peaks?

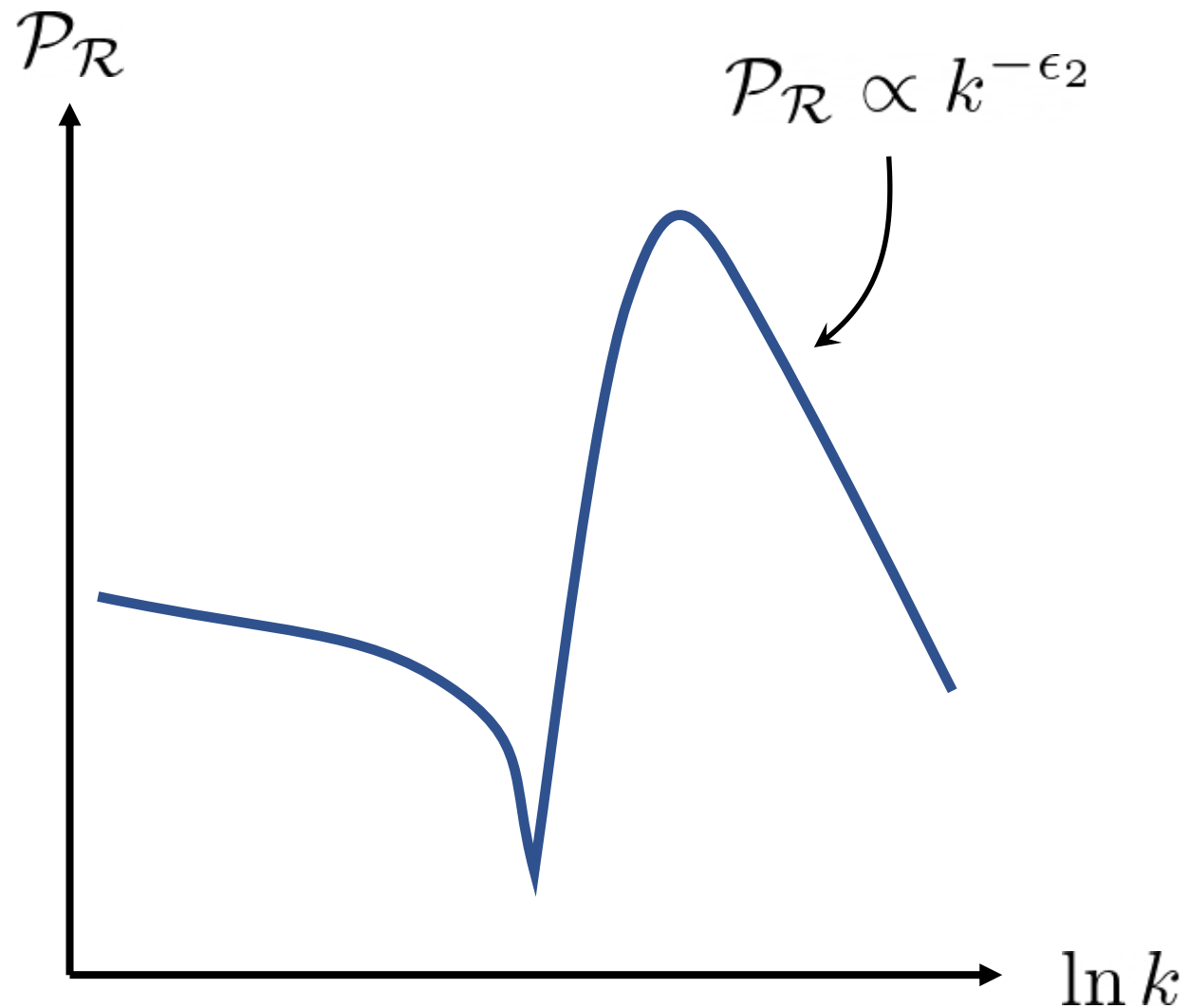
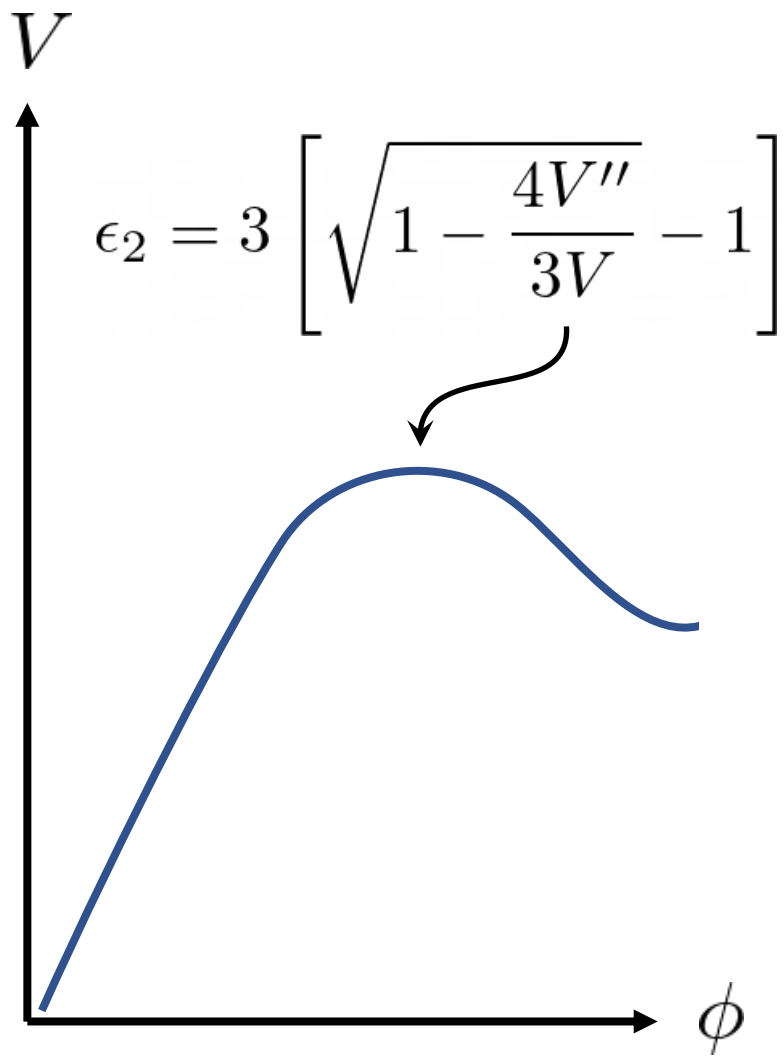
- “Outermost peak” gives final collapse?
- Overlapping peaks?

Conclusions

Stochastic inflation introduces non-Gaussian corrections to PBH statistics

Compaction function formalism needed for accurate results

Spiked radial profiles: what to do?



[2205.13540]

