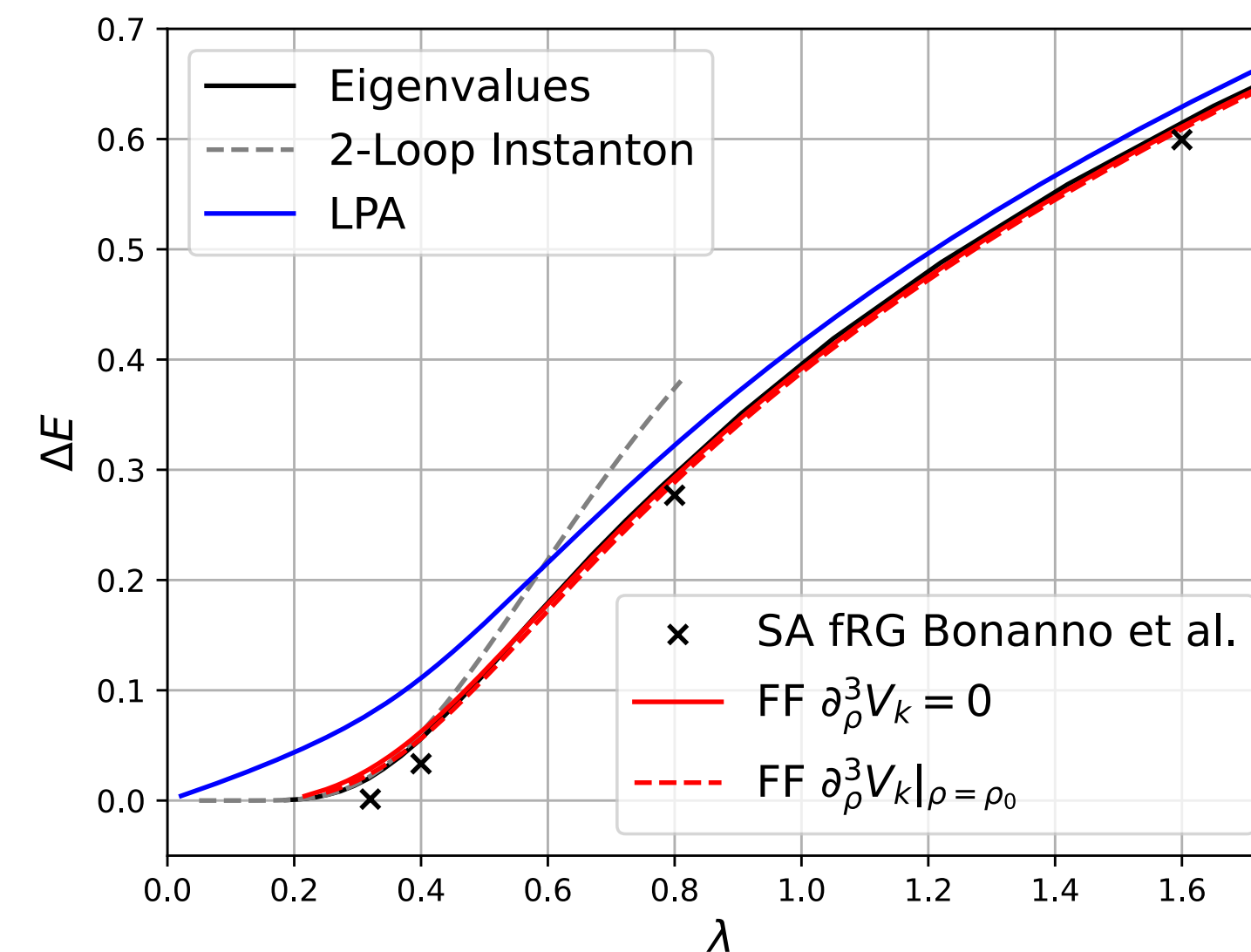
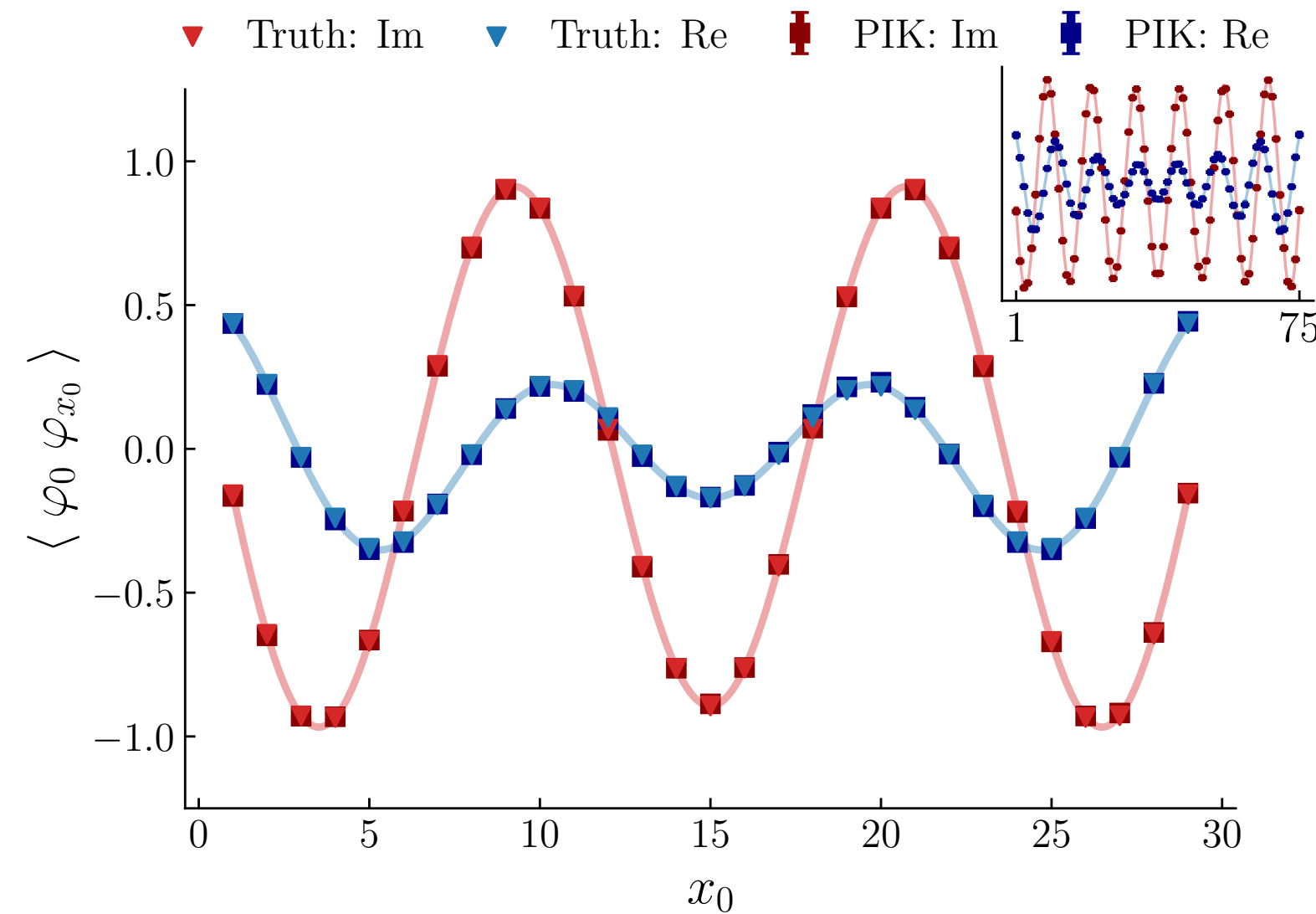
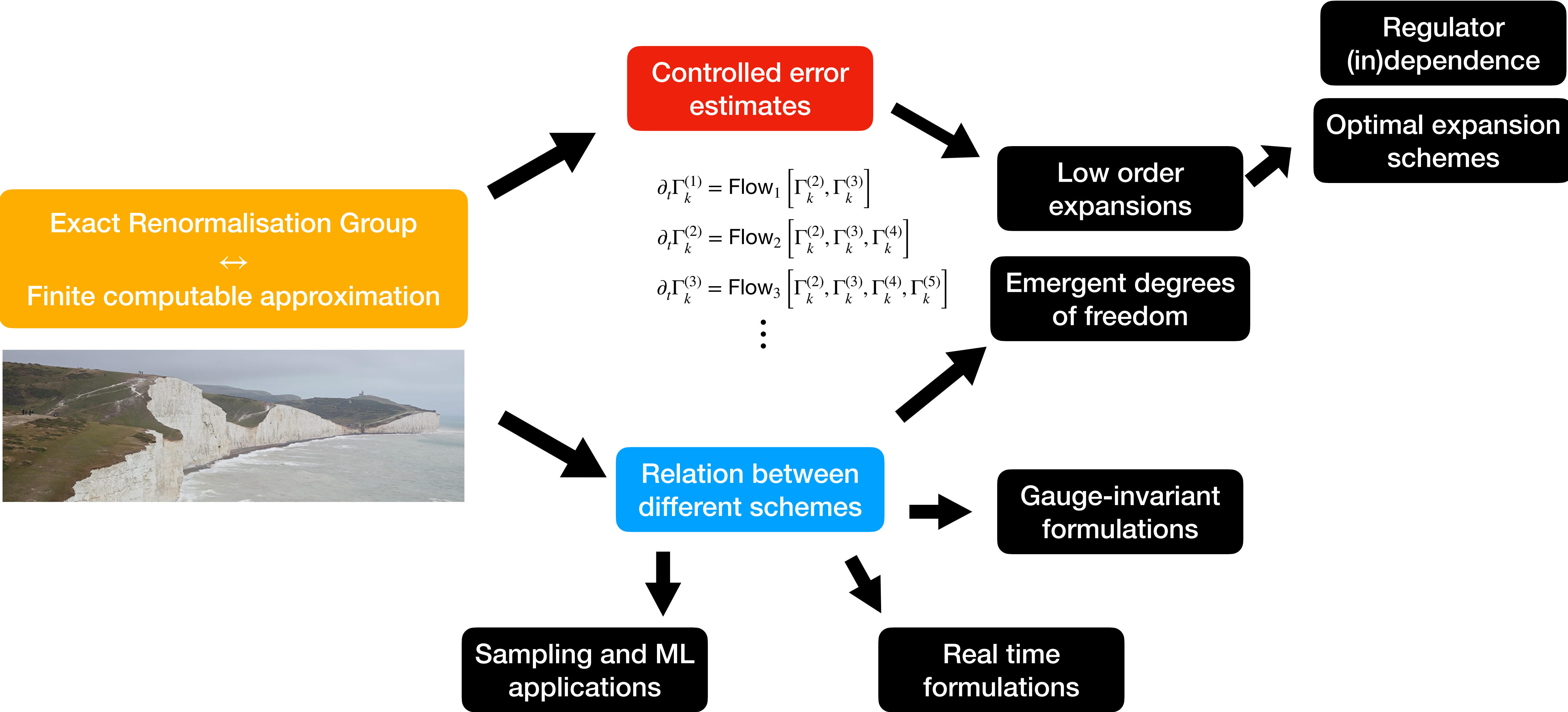


Applications of the Physics-Informed Renormalisation Group

From sampling to optimisation



Friederike Ihssen, Ruhr-Universität Bochum
05.09 - ERG 2026



The RG as a transport task

- Original idea: Compression of information under repeated block-spinning transformations $\dot{\phi} : \phi_i(x) \mapsto \phi_{i+1}(x)$
 → Keep the number of sites intact/infinite lattice

- Consider infinitesimal field reparametrisations $\phi \rightarrow \phi + \delta t \dot{\phi}(\phi)$
 → **Change of variables** in the path integral and changing distribution $p_t \rightarrow p_t + \delta t \frac{\partial}{\partial t} p[\phi]$

Wegner flow

Wegner, J. Phys. C 7 (1974) 2098

$$\frac{dp_t[\phi]}{dt} = - \int_x \frac{\partial}{\partial \phi(x)} \left\{ \dot{\phi}_t[\phi] p_t[\phi] \right\}$$

Diffusion

$$p_t[\phi] \propto e^{-S_t[\phi]}$$

→ **Continuity equation** for the overall distribution

- Most modern momentum space RG does not use block-spinning anymore:
 → (Non-linear) **Fokker-Planck** dynamics:

Polchinski-type kernel:

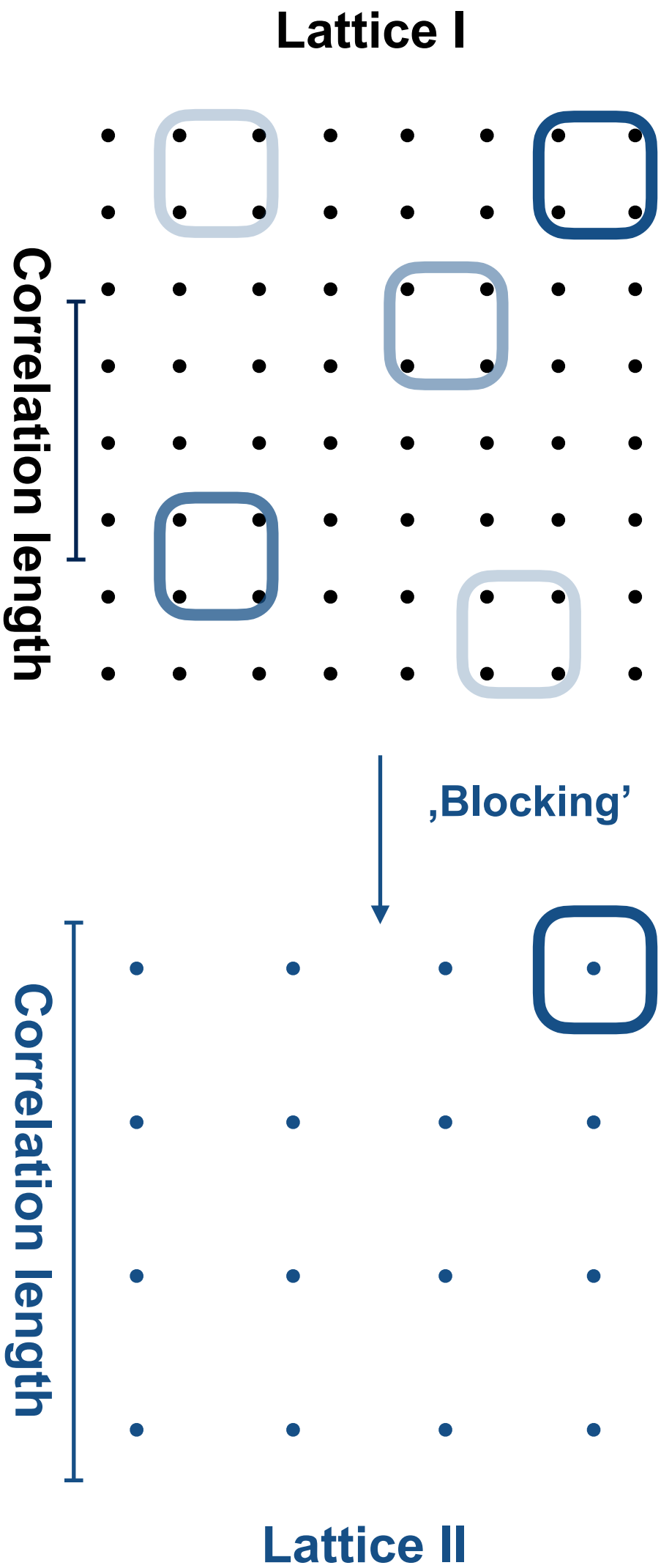
Scale transformations

Additional Reparametrisations

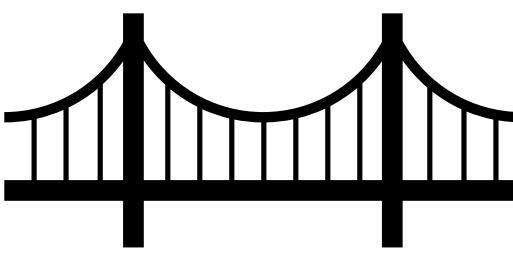
$$\dot{\phi}(\phi) = \frac{1}{2} \mathcal{C}(\phi) \frac{\partial}{\partial \phi} S_t(\phi) + \dot{\phi}_r(\phi)$$

Gradient flow

Selection/Filtering of propagating modes, e.g. $\mathcal{C} \propto -G \partial_t R_k G$



Scale transformations in the path integral



Consider the generating functional for a theory:

$$Z_t[J] = \int D\phi p_t[\phi] e^{J \hat{O}_{\phi,t}[\phi]}$$

← Current coupled to some generic operator

← Some effective degree of freedom

Scale dependent distribution: $p_t[\phi] = \frac{1}{\mathcal{N}} e^{-S_t[\phi]}$

e.g. $S_t = S + \Delta S_t$

ϕ is in the **co-moving coordinate frame** of the block spinning, i.e. $\partial_t \phi = 0$ (passive transformation)

→ Use the scale dependent (composite) operator:

$\hat{O}_{\phi,t}[\phi]$ as a **book-keeping device**

$\hat{O}_{\phi,t_0}[\phi] = \phi$

Physics-informed setup

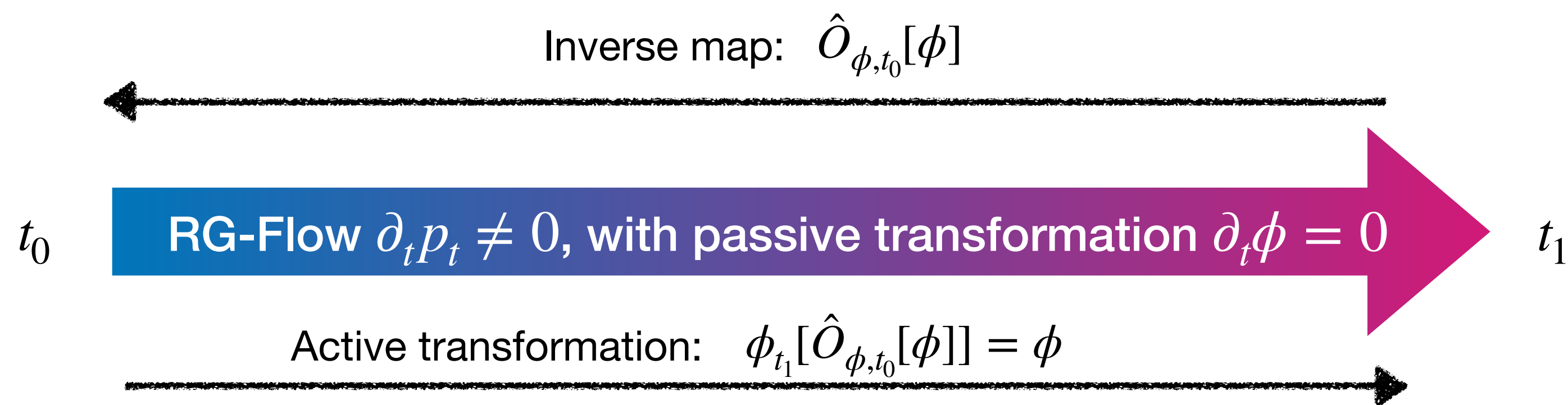
Two strategies

$$\partial_t Z_t[J] = \int D\phi \left\{ \underbrace{\partial_t p_t[\phi]}_{\text{Wegner flow}} + J \underbrace{\partial_t \hat{O}_{\phi,t} p_t[\phi]}_{\text{Change of the composite operator}} \right\} e^{J \hat{O}_{\phi,t}[\phi]}$$

1) $\partial_t Z_t \equiv 0 \Rightarrow \hat{O}_\phi$ completely absorbs the flow
Sampling

2) $\partial_t Z_t \neq 0 \Rightarrow \hat{O}_\phi$ corrects the existing flow
Modifications in fRG

Sampling with the RG-flow



Takeaway:

Integrating $\dot{\phi}[\phi]$ for microscopic transformations gives us a map for the field transformation

$$\phi_t = \phi_{t_0} + \int_{t_0}^t ds \dot{\phi}_s[\phi_s]$$

- We use $\partial_t Z = 0$ to **reconstruct the coarse-graining map**
- For general holomorphic operators $O[\phi]$ one can show

Observable in a hard theory Sample from easy distr.

$$\int D\phi p_t[\phi] O[\phi] = \int D\varphi p_0[\varphi] O[\phi_t[\varphi]]$$

Map to $t = t_1$

→ This is valid for any pair $(p_t[\phi], \dot{\phi}[\phi])$ that fulfills the Wegner equation

→ **Physics-informed kernels:** pick a suitable constraint to fix the pair, e.g.

- Langevin evolution
- Diffusion models: [Talk by L. Wang](#)
- Polchinski kernel: [Talk by Y. Ashida, M. Gubinelli](#)
- ...

Sample p_t with $\phi_t(\phi_0)$, where $\phi_0 \sim p_0$

The physics-informed mindset

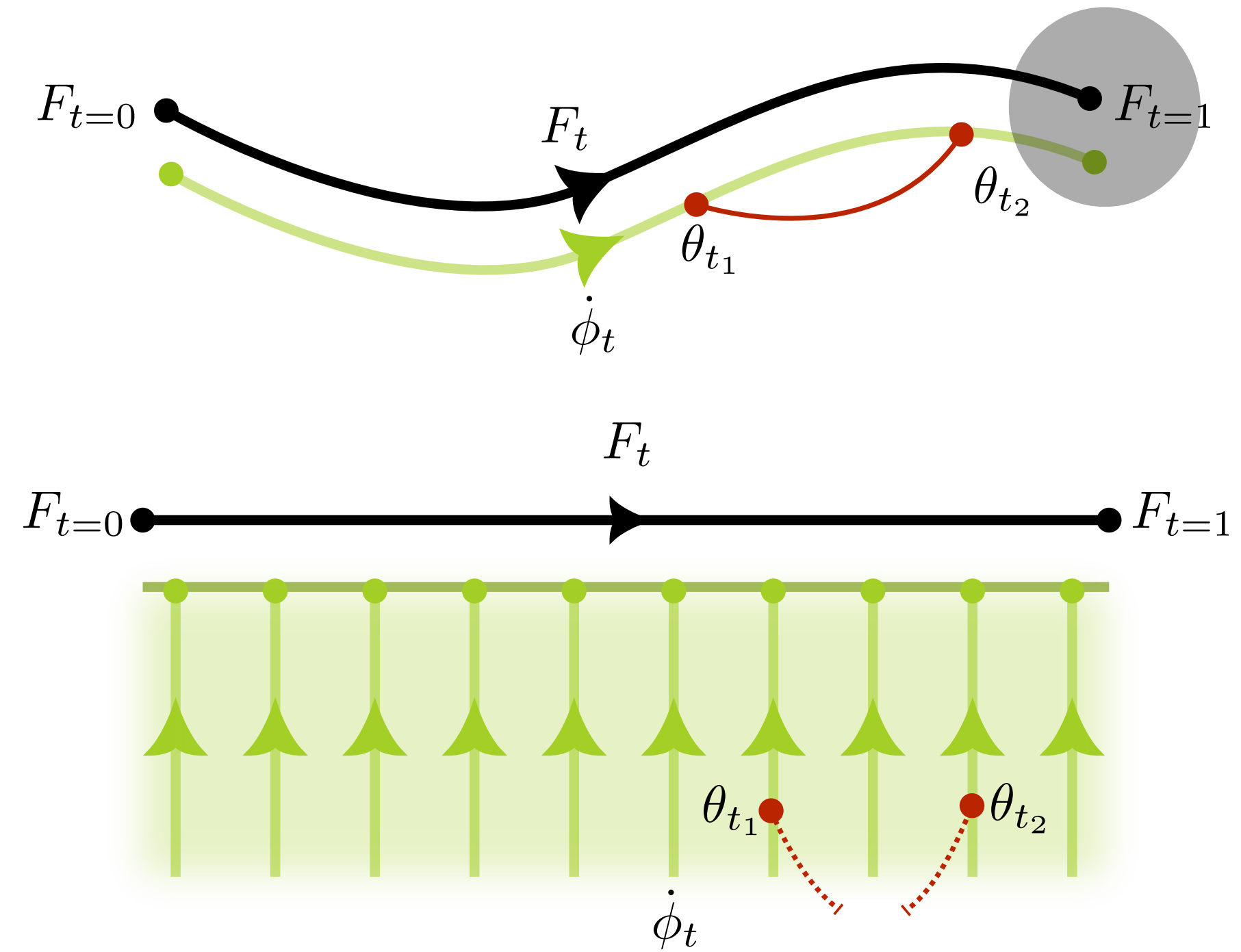
Allowing general time-dependent paths for distributions

$F_t = \{S_t, p_t, \dots\}$ does not guarantee the desired physical distribution

...so we impose a trajectory analytically. Then the Wegner equation

$$\partial_t S_t + \dot{\phi} \frac{\delta}{\delta \phi} S_t = \frac{\delta}{\delta \phi} \dot{\phi} \quad \longrightarrow \quad RG_1 + \dot{\phi} RG_2 = \frac{\delta}{\delta \phi} \dot{\phi}$$

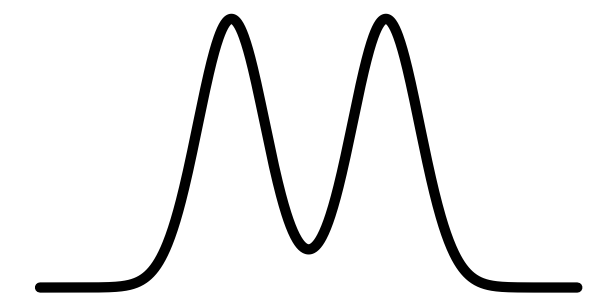
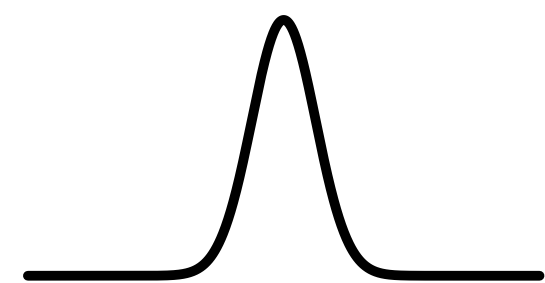
...is a differential equation at every time-slice!



Example: zero-dimension ϕ^4 - theory

$$\hat{S}_t(\phi) = \frac{1}{2} m^2(t) \phi^2 + \frac{\lambda(t)}{4} \phi^4$$

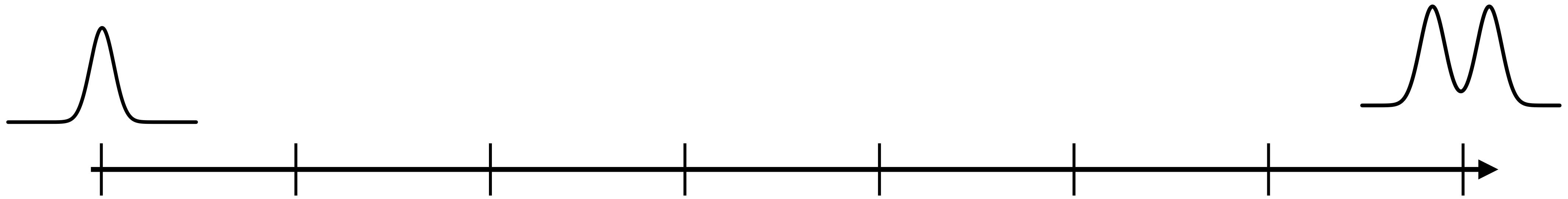
Task: solve an ODE in ϕ at every time-slice



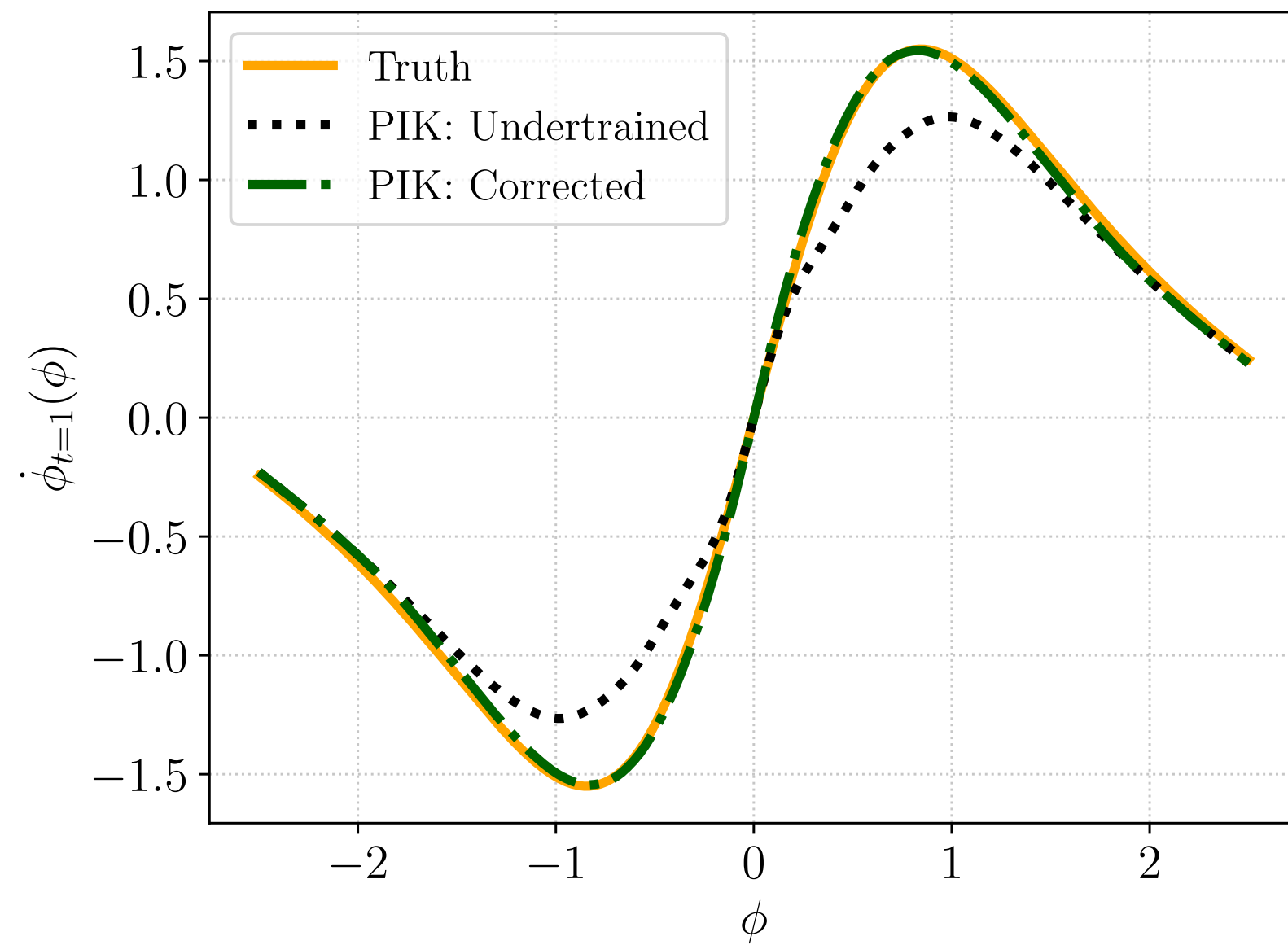
$$(m_0^2, \lambda_0) = (1, 0) \quad \longrightarrow$$

Analytic interpolation between couplings

$$\longrightarrow (m_1^2, \lambda_1) = (-2, 1/2)$$



$$(m_0^2, \lambda_0) = (1, 0) \xrightarrow{\text{red arrow}} (m_t^2, \lambda_t) = (m_0^2 + t(m_1^2 - m_0^2), \lambda_0 + t(\lambda_1 - \lambda_0)) \xrightarrow{\text{red arrow}} (m_1^2, \lambda_1) = (-2, 1/2)$$

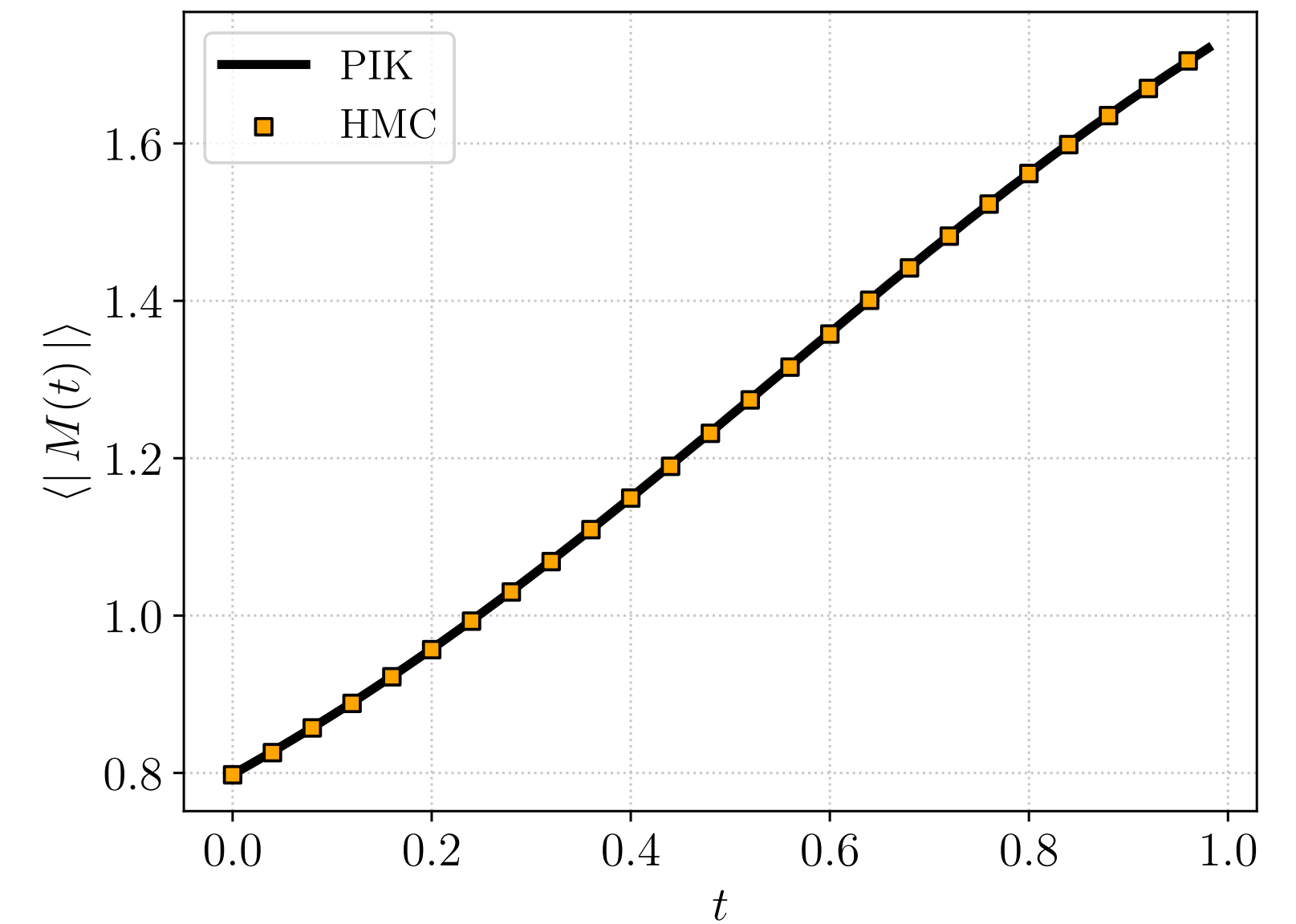
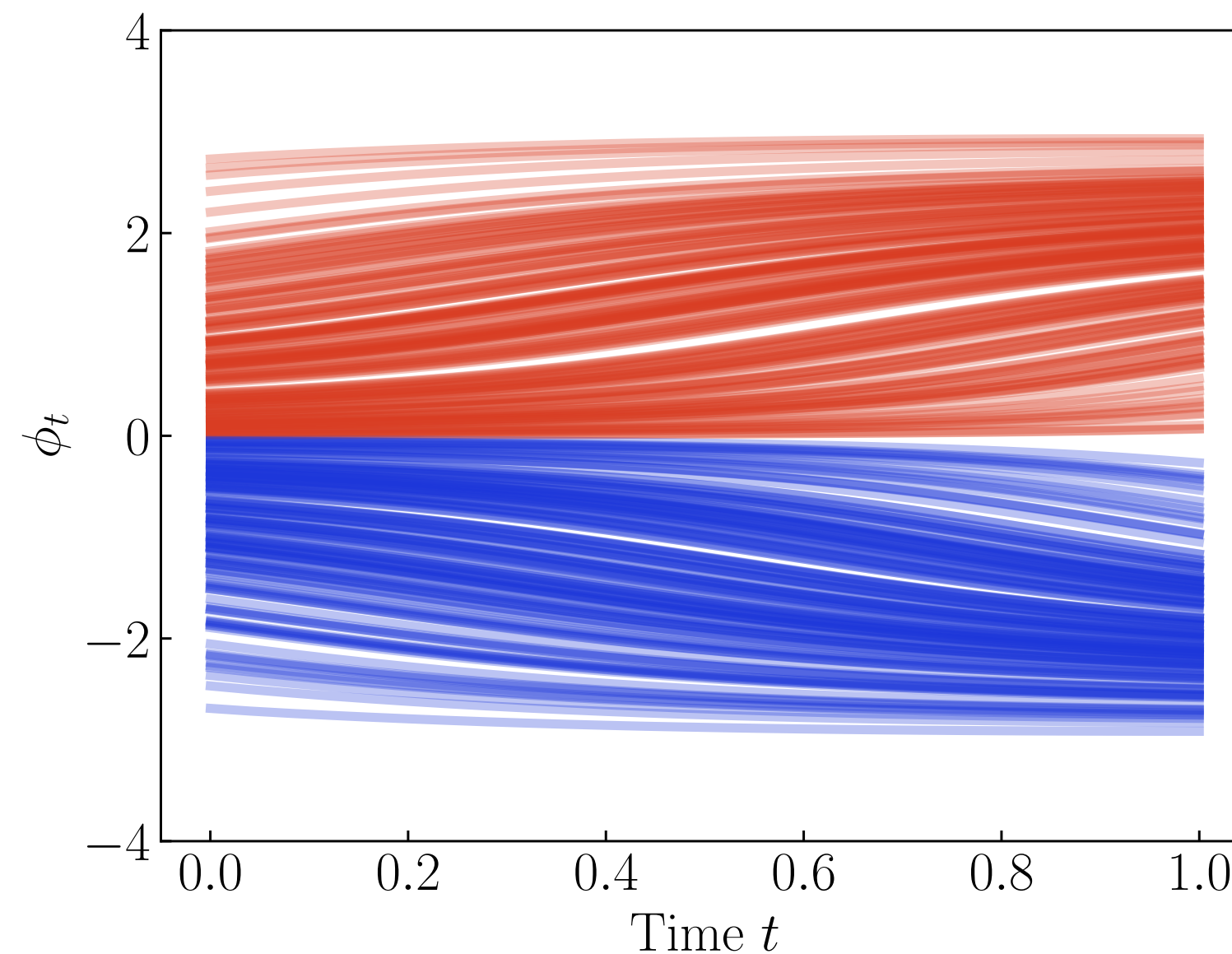


- Compute PIKs separately at every time-slice
- Linearity of the equation allows for systematic improvements

- Evolve samples separately with

$$\phi_t = \phi_{t_0} + \int_{t_0}^t ds \dot{\phi}_s[\phi_s]$$

- This is an **optimal transport flow**



- Sample operators along this trajectory

$$\langle O \rangle_t = \frac{1}{N} \sum_n O[\phi_{n,t}]$$

Part 1 (Strategy 1): Conclusion and outlook

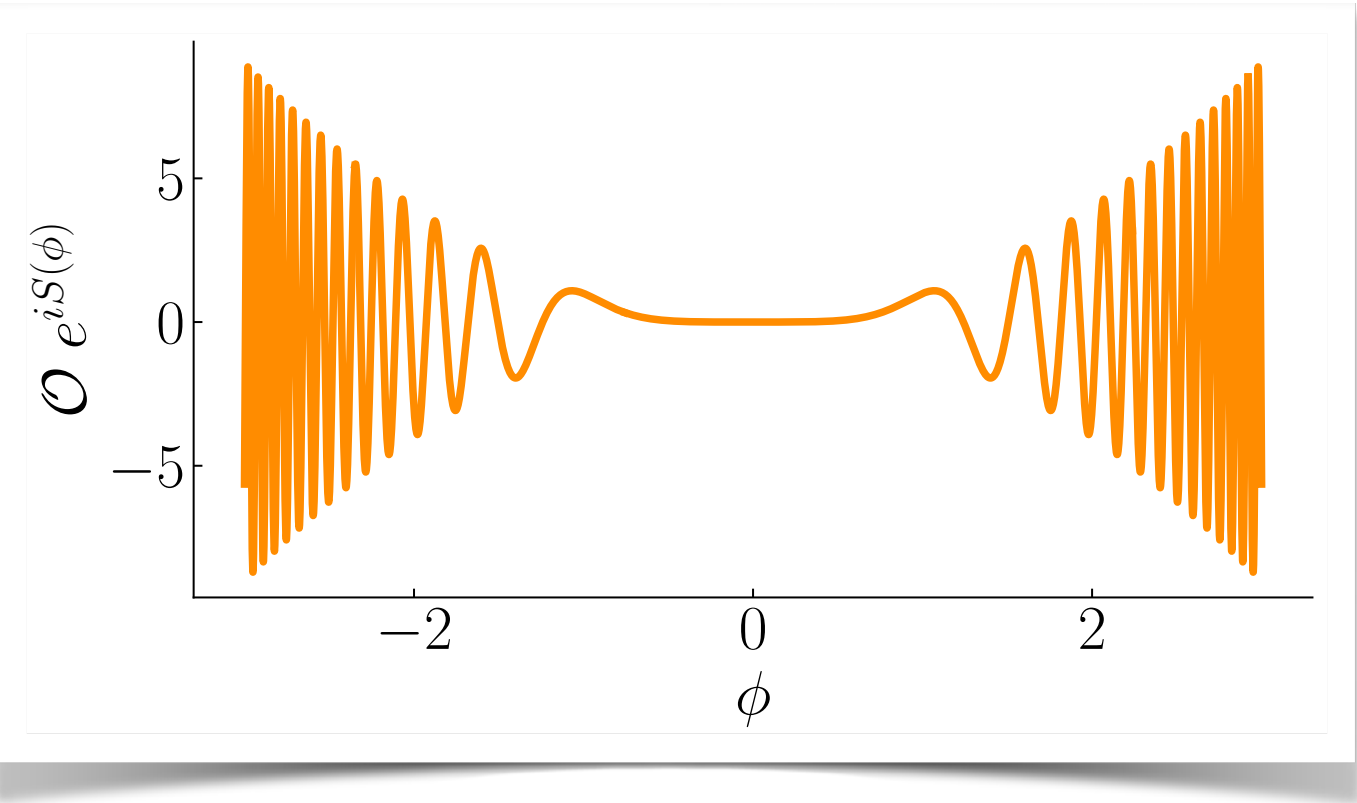
PIK is an umbrella for sampling based strategies:

- Fixing a target action S_t allows to sample this distribution directly $\leftrightarrow$ Contrast to the Polchinski flow or complex Langevin
- We have only used **general reparametrisation freedom**: 0d does not know about scale transformations
- Particularly interesting for complex actions/systems with a sign problem

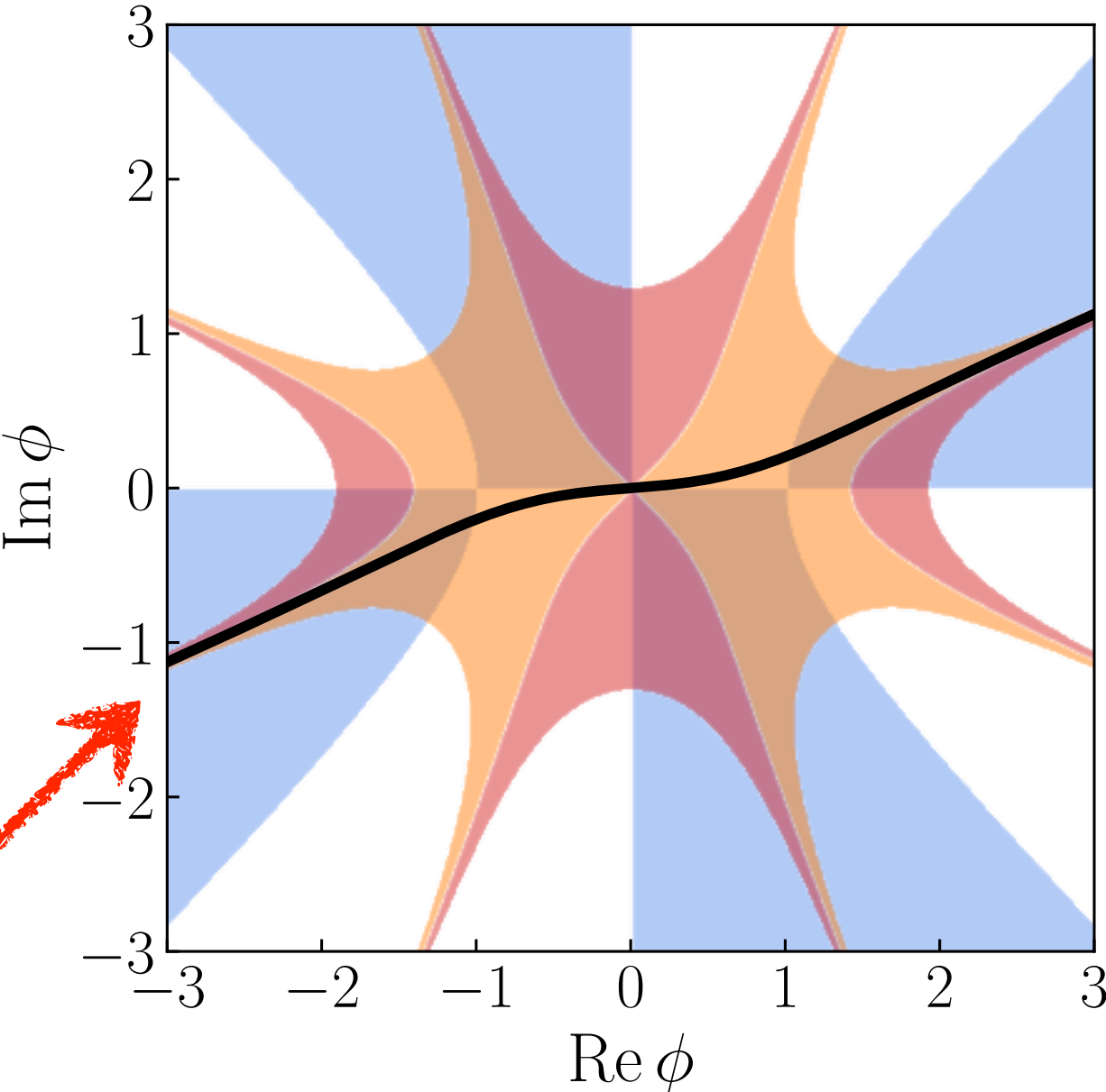
Weight preservation:

$$p_0 d\mu_0 \in \mathbb{R} \longrightarrow p_t d\mu_t \in \mathbb{R}$$

0-dim toy model of a real time theory

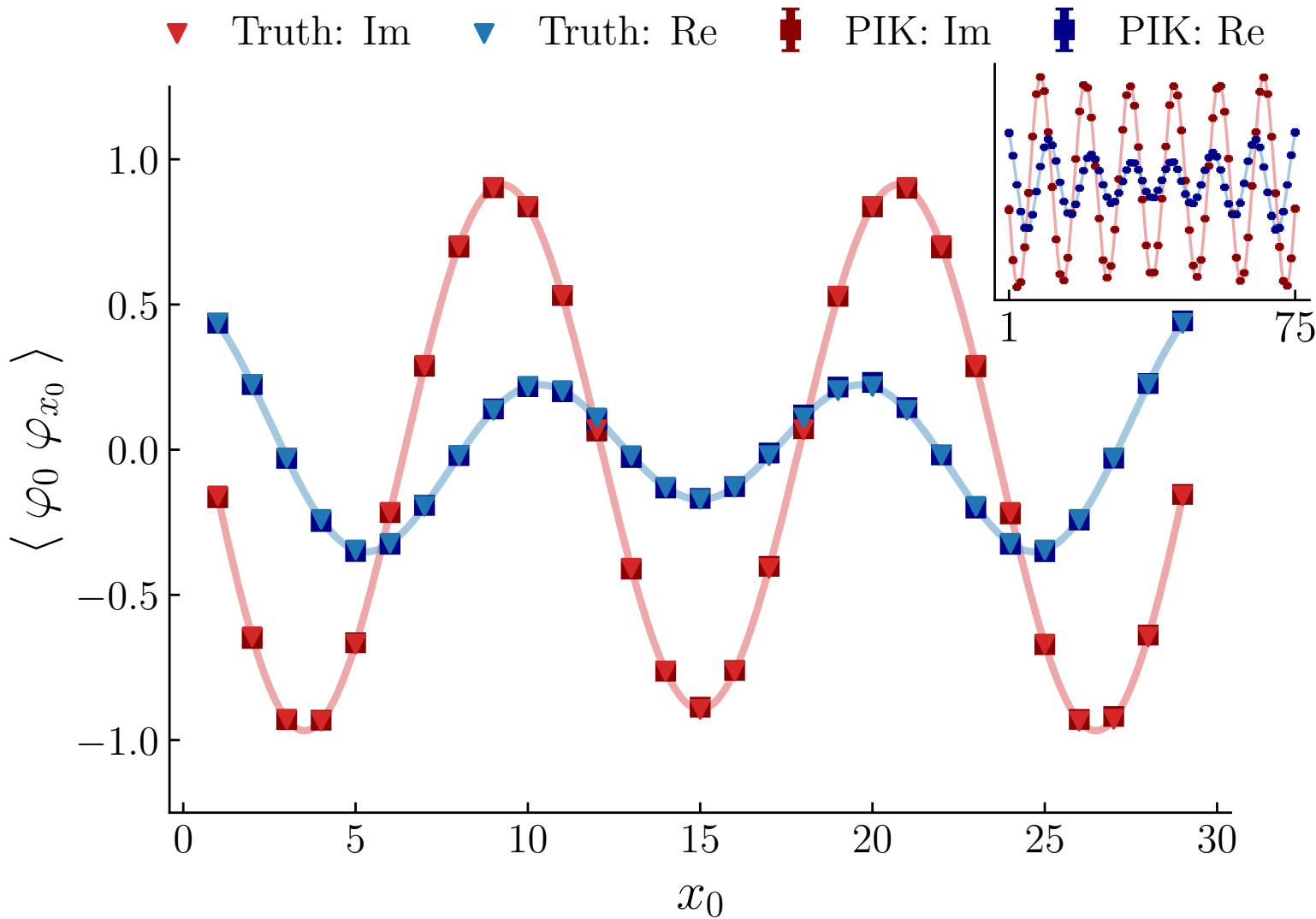


No guessing involved!



PIK implements a contour deformation to remain weight preserving

Complex contour deformation



1+0 dim theory:
Quantum harmonic oscillator

Generalised RG flows: Strategy 2

Scale transformations

$$\partial_t Z_t[J] = \int D\phi \left\{ \underbrace{\partial_t p_t[\phi]}_{\text{Wegner flow}} + J \underbrace{\partial_t \hat{O}_{\phi,t}}_{\text{Change of the composite operator}} p_t[\phi] \right\} e^{J \hat{O}_{\phi,t}[\phi]}$$

Reparametrisations

Implement the **scale suppression/Regulator** as usual

$$p_t[\phi] = \frac{1}{\mathcal{N}} e^{-S[\phi] - \Delta S_t[\phi]}$$

By allowing $\partial_t Z_t \neq 0$ we remain with a very general reparametrisation freedom

$$\partial_t \hat{O}_{\phi,t}$$

Sampling

$$\dot{\phi} = \langle \partial_t \hat{O}_{\phi,t} \rangle, \quad \phi = \langle \hat{O}_{\phi,t} \rangle$$

Direct computation & optimisation

$$\frac{d p_t(\phi)}{dt} = - \int_x \frac{\partial}{\partial \phi(x)} \left[\dot{\phi}_t(\phi) p_t(\phi) \right]$$

Legendre transform

$$\left(\partial_t + \int_x \dot{\phi}[\phi] \frac{\delta}{\delta \phi} \right) \Gamma_\phi[\phi] = \frac{1}{2} \text{Tr} \left[G_\phi[\phi] \left(\partial_t + 2 \frac{\delta \dot{\phi}}{\delta \phi} \right) R_k \right]$$

⋮

Wegner flow

Generalized flow equation

Wetterich equation $\dot{\phi} = 0$

Wegner, J. Phys. C 7 (1974) 2098

Pawlowski, *Annals Phys.* 322 (2007) 2831

Wetterich, PLB 301 (1993) 90

General field transformations

Scale transformation implemented with regulator → Remaining reparametrisation freedom

Lets go back to the generalised RG...

$$\left(\partial_t + \int_x \dot{\phi}[\phi] \frac{\delta}{\delta \phi}\right) \Gamma_{\phi}[\phi] = \frac{1}{2} \text{Tr} \left[G_{\phi}[\phi] \left(\partial_t + 2 \frac{\delta \dot{\phi}}{\delta \phi}\right) R_k \right]$$

$\dot{\phi} = 0$
 $\dot{\phi} \neq 0$

Implements Wetterich RG (previous example)

Corrections of loop-order allow for optimization

Tree-level corrections change the theory

- Key insights:

 - (1) $\dot{\phi}[\phi]$ is a functional and can be truncated the same as the effective action $\Gamma[\phi]$
 - (2) This additional freedom allows to implement constraints in the RG flow

- Example: Rebosonization

Talk by F. Rennecke, F. R. Sattler, ...

Absorb the divergence into a Yukawa coupling $\propto h\phi_a\bar{\psi}M_a\psi$ ∂_t

Ansatz: $\dot{\phi} = A_a(\phi) \bar{\psi}M_a\psi$

$A_a(\phi) \rightarrow \partial_t g_a = 0$

Introduces corrections to the Yukawa coupling!

For linear transformations

Gies, Wetterich, PRD 65 (2002) 065001

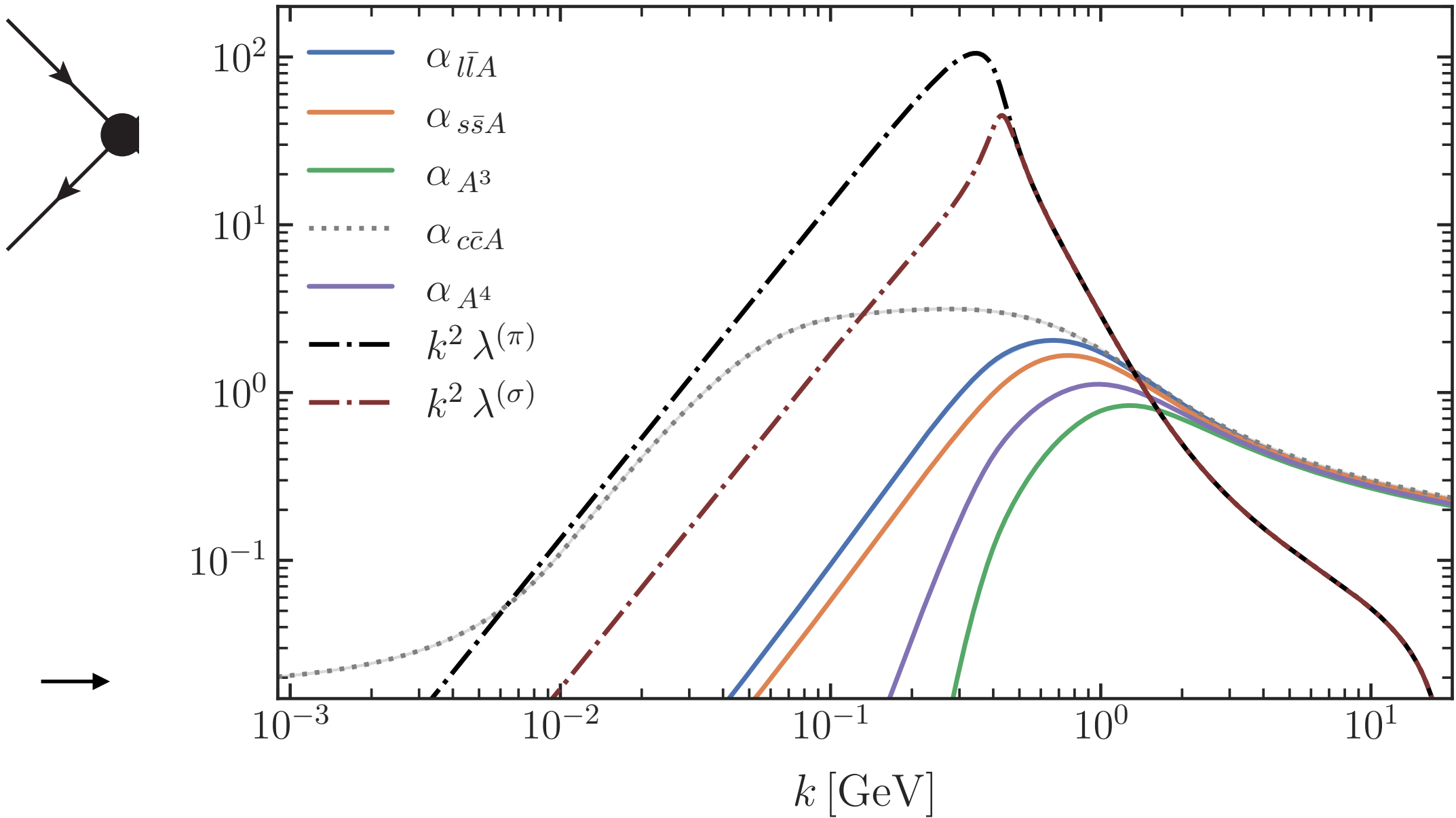
Implementation in QCD

Fu, Pawłowski, Rennecke, PRD 101 (2020) 5, 054032
FI, Pawłowski, Sattler, Wink, PRD 113 (2026) 9, 094038

Effectively, a four-fermi vertex with a non-trivial momentum dependence is considered

→ No divergence

$$\lambda_i = \frac{h^2}{Z_{\psi}^2(m_i^2 + p^2 + R_k)}$$



...but there is more than that

Physics-informed RG

$(\Gamma_\phi[\phi], \dot{\phi}[\phi])$

Generally we can consider the pair and implement some physical constraint with it:

PIRG idea

$\Gamma_{\phi,t_0}[\phi]$

$(\Gamma_{\phi,t}[\phi], \dot{\phi}_t[\phi])$

$\Gamma_{\phi,t_1}[\phi]$

$$\left(\partial_t + \int_x \dot{\phi} \frac{\delta}{\delta \phi}\right) \Gamma_T[\phi] = \text{RG}_1[\Gamma_T] + \text{RG}_2[\Gamma_T] \frac{\delta \dot{\phi}}{\delta \phi}$$

Reconstruction & observables: Generalised operator flow

$\left(\partial_t + \int_x \dot{\phi} \frac{\delta}{\delta \phi}\right) \mathcal{O}_t[\phi]$

$= -\frac{1}{2} \text{Tr} \left[G_T \left(\partial_t + 2 \frac{\delta \dot{\phi}}{\delta \phi} \right) G_T \frac{\delta^2}{\delta \phi^2} \mathcal{O}_t[\phi] \right]$

$+ F_\phi$

observable

propagator insertion

DoF for optimisation

Example: Direct Computation

Step 1: Solve approximate flow for $\Gamma_{t,\text{app}}$

Step 2: Use $\Gamma_T = \Gamma_{t,\text{app}}$ as target action

Solve generalized flow for $\dot{\phi}_t$

Advantages

- (i) Trade intricate diffusion eq. for a simple linear diff. eq.
- (ii) Optimised and more versatile expansion schemes
 - (a) Ground state expansion
 - (b) Feed-down flows

⋮

Disadvantage: Reconstruction!

The ground state-expansion

Topology in the anharmonic oscillator

- Quantum mechanics: i.e. 1+0 dimensional QFT

→ Setup with **tunneling**

→ Consider the **energy gap** between ground state and the first excited state ΔE

$$\langle \hat{q}(0) \hat{q}(\tau) \rangle = \sum_n |\langle 0 | q | n \rangle|^2 \exp(-(E_0 - E_n)\tau)$$

→ Behavior at large imaginary times

$$\langle \hat{q}(0) \hat{q}(\tau) \rangle \propto \exp(-\Delta E \tau)$$

This is all known from the **Schrödinger equation**

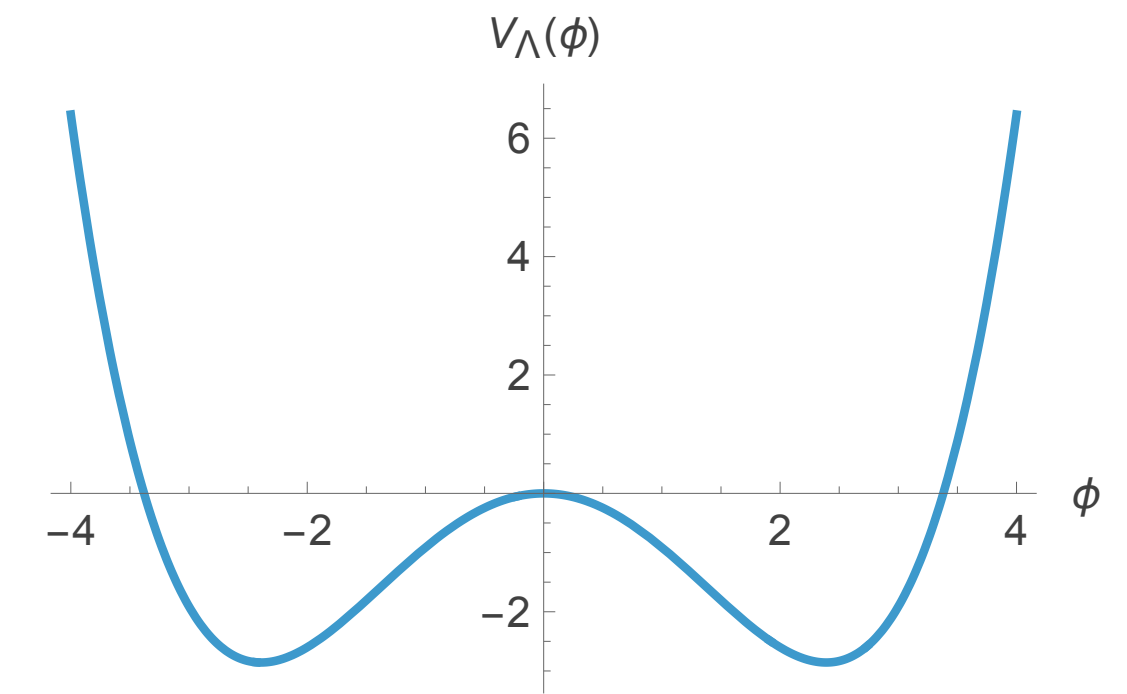
→ but also: Expansion about **Instanton solution**

$$\Delta E(\lambda_\varphi \rightarrow 0) \propto \frac{1}{\sqrt{\lambda_\varphi}} e^{-a_{\text{inst}}/\lambda_\varphi}$$

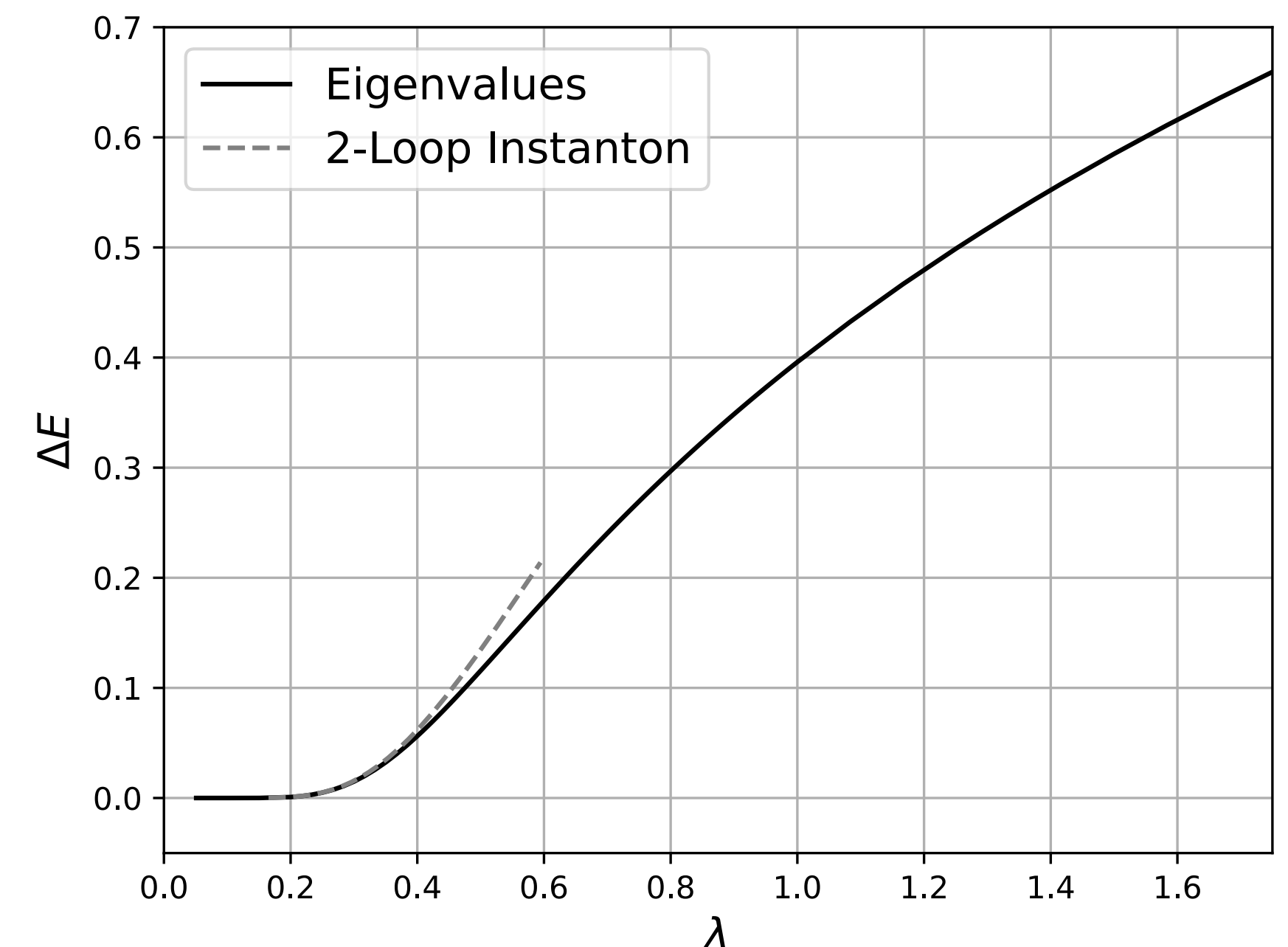
BKT-type scaling

Testing ground for novel methods with non-perturbative physics

→ In particular where fRG has struggled before



$$V_\Lambda(\hat{q}) = -\frac{1}{2}\hat{q}^2 + \frac{\lambda}{8}\hat{q}^4$$



The anharmonic oscillator in the fRG

Chose a simple truncation with the derivative expansion (LPA)

$$\Gamma_k[\phi] = \int_{\tau} \left[\frac{1}{2} (\partial_{\tau} \phi)^2 + V_k(\phi) \right]$$

Solve the PDE for the [Wetterich equation](#), and extract $\Delta E = \sqrt{V^{(2)}(\phi_{\text{eom}})}$

Solve the [generalized equation](#) in the same truncation for Γ_k

With an **optimized kernel**:
(Ground state expansion)

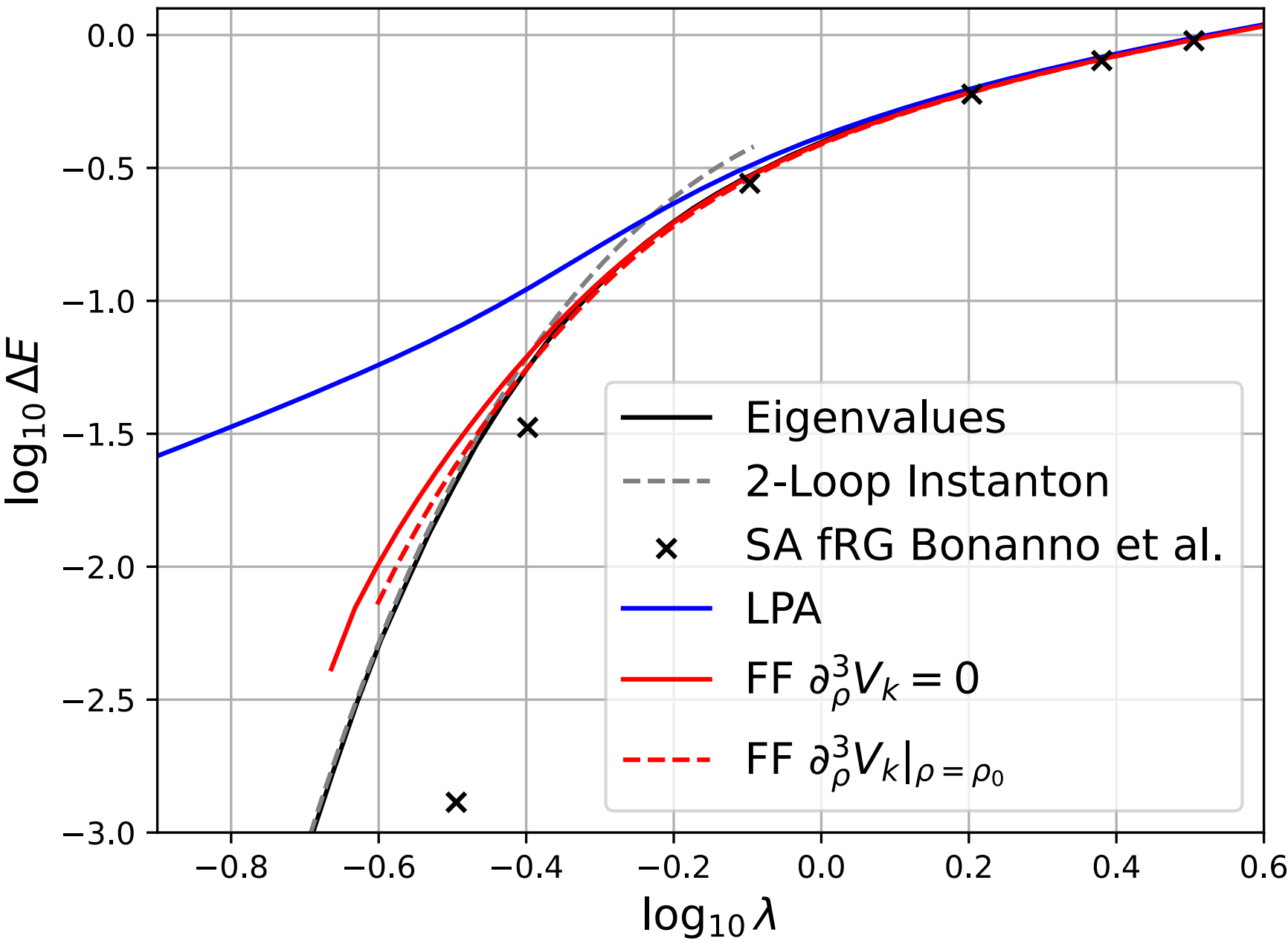
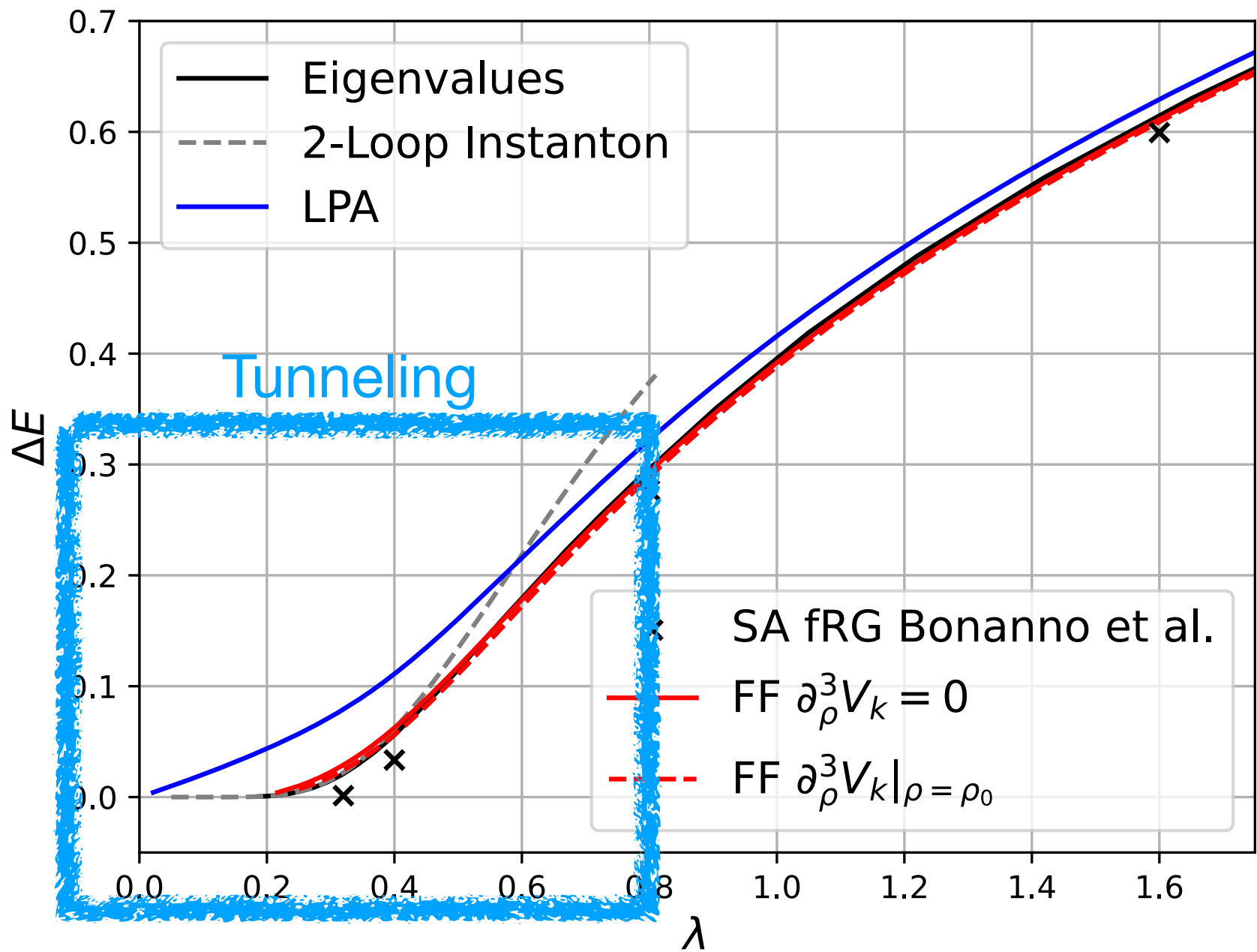
$$\dot{\phi}(\phi) \rightarrow \partial_t (\partial_{p^2} \Gamma_k^{(2)}[\phi]) = 0$$

The optimized kernel **eliminates the next order of the expansion scheme**.

Results in a field dependent anomalous dimension:

This allows the extraction of the **instanton scaling** up to 1-2% exactness!

$a_{inst,ex} = 1.886....$ $a_{inst,FF} = 1.910(2)$

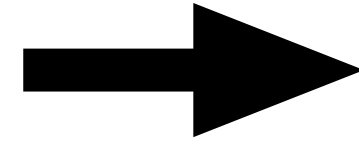


The ground-state expansion

Consider the standard derivative expansion:

$$\Gamma_k[\phi] = \int_{\tau} \left[\frac{1}{2} (\partial_{\tau} \phi)^2 + V_k(\phi) \right]$$

Zeroth-order expansion



$$\Gamma_k[\phi] = \int_{\tau} \left[\frac{1}{2} Z(\phi) (\partial_{\tau} \phi)^2 + V_k(\phi) \right]$$

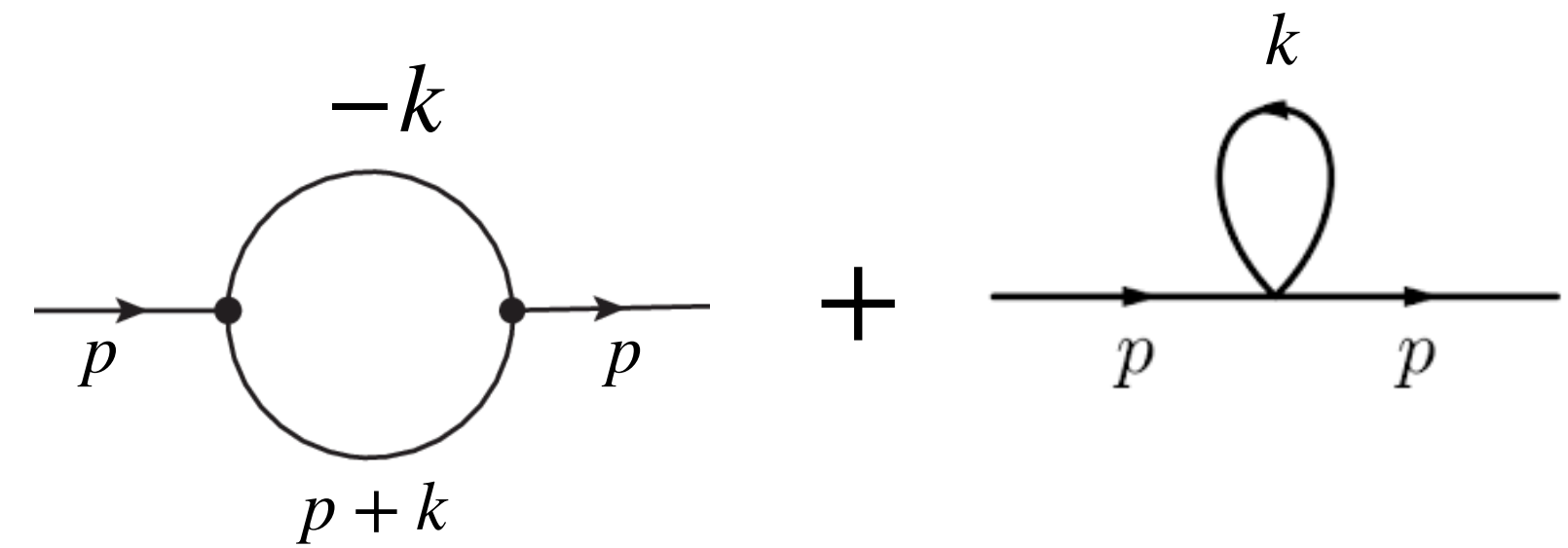
First-order expansion

RG-flow now accounts for fluctuation physics of order p^2

In this formulation the self-energy diagrams get momentum dependent corrections:

→ It creates e.g. a momentum dependent Tadpole diagram for all n-point functions

$$\partial_t \Gamma_k^{(2)}[\phi] \propto$$



$$\propto Z^{(2)}(\phi) p^2$$

Consider an expansion about the classical dispersion:

$$\Gamma_k^{(2)} = p^2 + \tilde{m}^2(\phi)$$



$$\dot{\phi}(\phi) \rightarrow \partial_t Z(\phi) \equiv 0$$

Effective action looks zeroth-order again, but **includes momentum dependent physics of all orders of tadpole diagrams** via $\dot{\phi}$



$$Z(\phi) \equiv 1$$

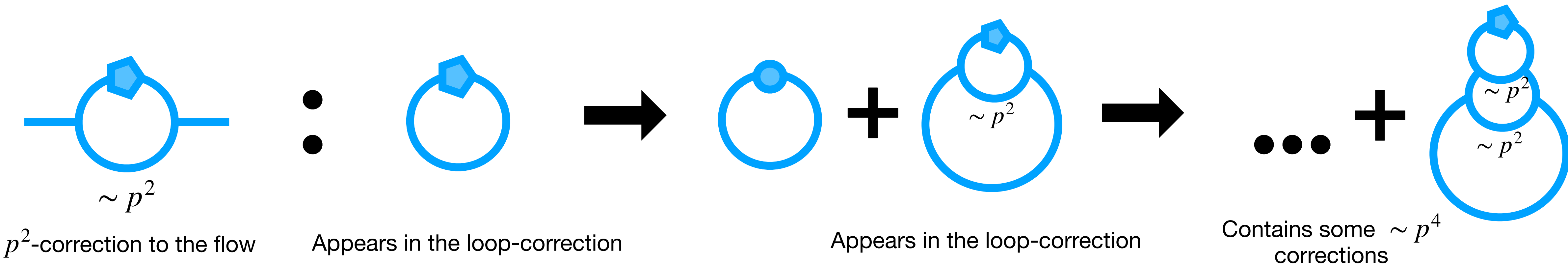
Vertex has no field dependence → No momentum dependent tadpole

The ,absorption strategy‘

Loop-correction

 =  + $\frac{\delta \dot{\phi}}{\delta \phi} R_k$

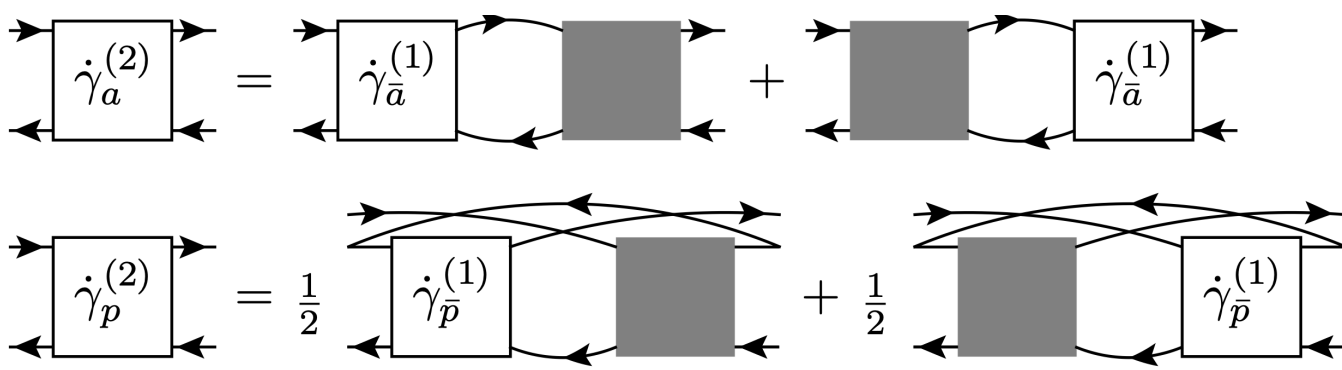
The ground-state expansion can be thought of as a **multi-loop scheme** (to infinite order):



In this sense it is related to the very successful **multi-loop fRG** (Fermionic 6-pt vertex is absorbed)

 Kugler, von Delft, PRL 120, 057403 (2018)
 Kugler, von Delft, PRB 97, 035162 (2018)

→ But automatically **includes all orders** (resummed scheme)



Application to Gauge theories (and other symmetry constraints): **Absorb terms that break the gauge invariance of the scheme!**

→ while feeding back their flow via implicit higher loop-corrections

E.g. Hard-wiring the modified Nielsen-Identities

 FI, Pawłowski, PRD 112 (2025) 10, 105005

Diffeomorphism-invariant Approach to Asymptotically Safe Quantum Gravity

Friederike Ihssen ^{1, 2} Benjamin Knorr ² Silas Mezger ^{3, 2} Jan M. Pawłowski ^{2, 4} and Paul P. Sprenger ²

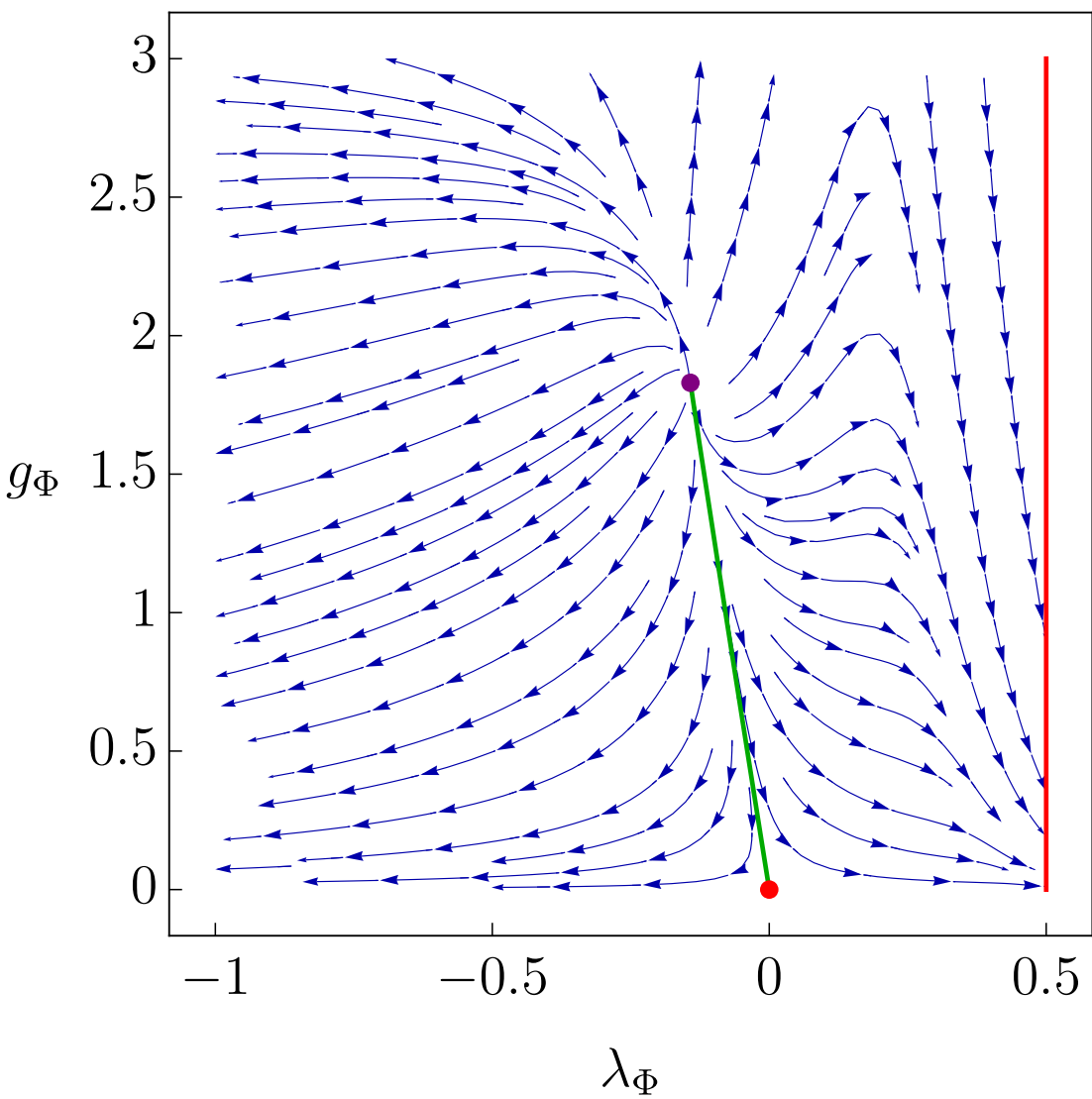
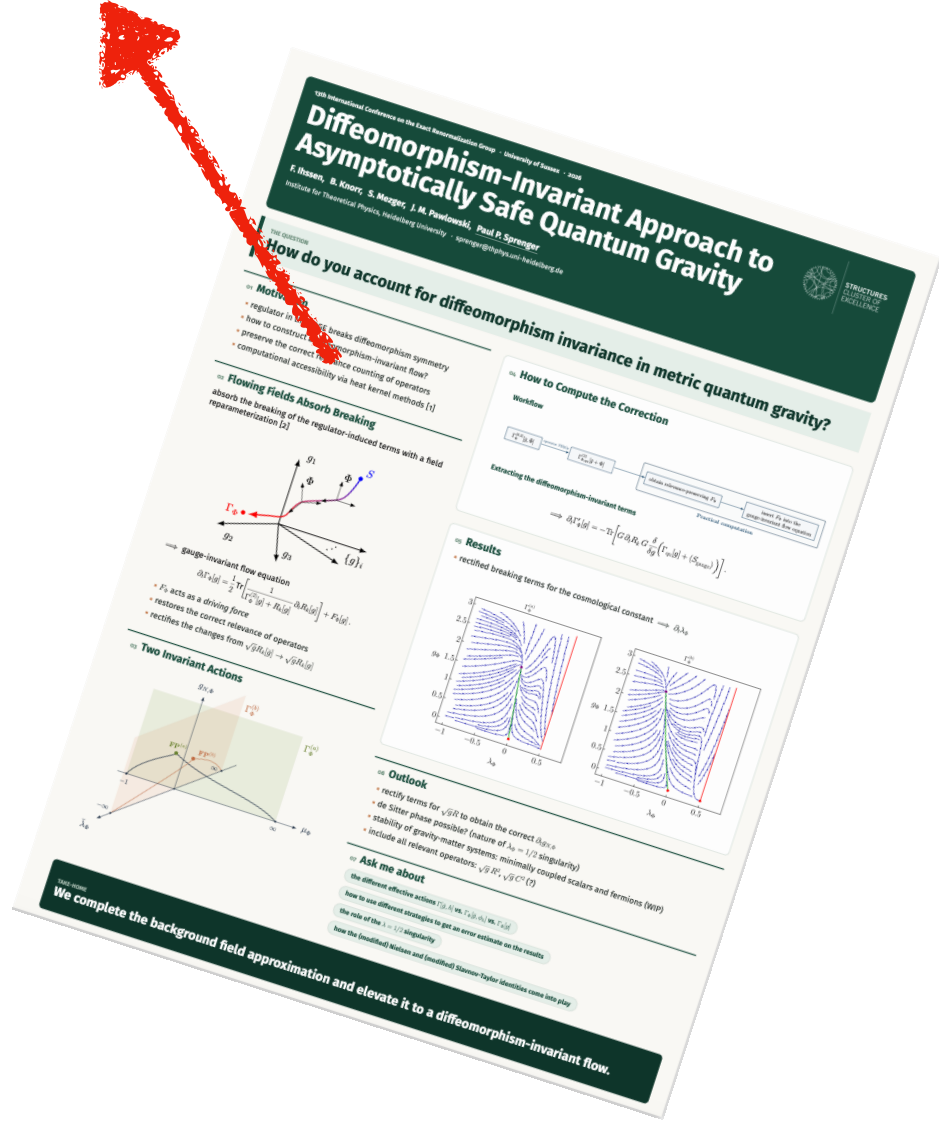
On the web tomorrow...

...since an entire week now

The regulator breaks gauge invariance.
Exact relations are around, e.g. modified Ward identities, Slavnov-Taylor, modified Nielsen Identities.

- ⇒ Embed a gauge invariant flow of your choice in the PIK
- ⇒ Compute difference to the exact solution with $\dot{\Phi}$

- ← Constraint fixes Γ_k
- ← Contains relevant and symmetry breaking terms



$\lambda_k^{(a,b)}$	$(\lambda_\Phi^*, g_\Phi^*)$	$\{\theta_i\}$
$\Gamma_{\Phi, \text{qu}}^{(a)}[g]$	$(-0.142, 1.830)$	$\{3.86, 1.95\}$
$\Gamma_{\Phi, \text{qu}}^{(b)}[g]$	$(-0.473, 2.369)$	$\{4.66, 1.89\}$

Reinsert

$$\partial_t \Gamma_{\Phi, rp}[\bar{g} + \Phi] = \frac{1}{2} \text{Tr } G_\Phi[\bar{g} + \Phi] \partial_t R_k[g] + F_\Phi[\bar{g} + \Phi]$$

Identify
Extract

Conclusion

- Generalised flows connect many of the topics of the conference!
 - Sampling strategies
 - Flows of the 1PI effective action
- **Analytical access** is very interesting to optimize the sampling task and understand AI algorithms
 - Complex actions + sign problem
- **Optimisation** in the Wetterich equation
 - Test approximations
 - Absorption of couplings for faster convergence
 - Relevance counting preserved in gauge invariant formulations

Thank you for your attention!

