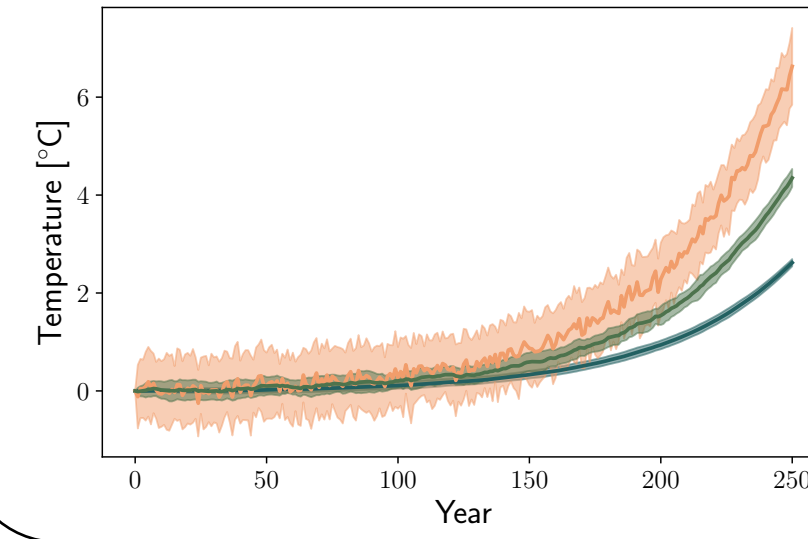


**Goal:** Emulate the statistics of a chaotic system

**1**

Select a climate variable of interest,  $w$



$$\frac{\partial w}{\partial t} = \mathcal{N}(w, F) + \epsilon \xi,$$

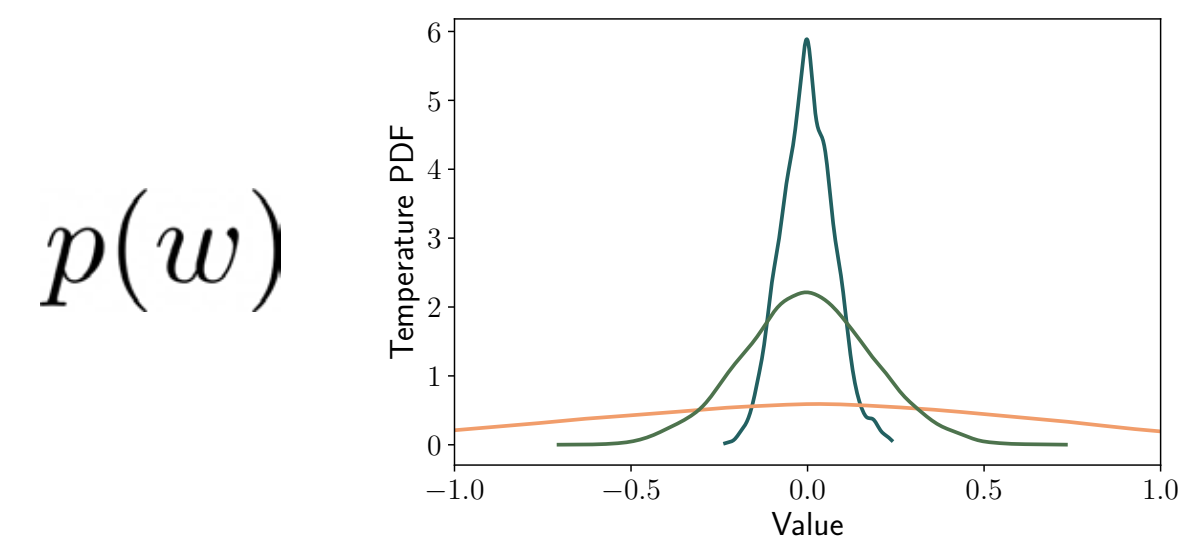
Option 1:

Option 2:

## Section 2.1: Theoretical Framework for Climate Emulation

**2a**

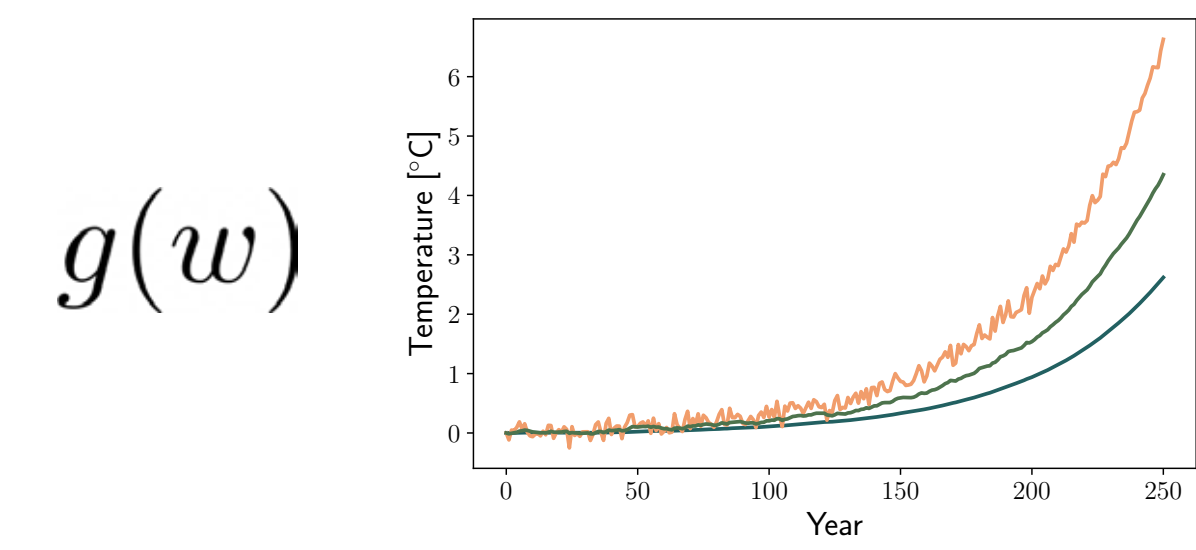
Target the full probability distribution



Two Sides of the Same Coin  
(Duality)

**2b**

Target a statistical quantity (e.g. mean/variance)



**3a**

Approximate Fokker-Planck Operator

$$\mathcal{F}(\cdot) = \frac{\partial}{\partial w} \left[ D \frac{\partial}{\partial w} (\cdot) - \mathcal{N}(w, F)(\cdot) \right]$$

Appendix A4:  
Theory Connection  
(Adjoint)

**3b**

Approximate Koopman Operator

$$\mathcal{K}(\cdot) = \mathcal{N}(w, F) \frac{\partial (\cdot)}{\partial w} + D \frac{\partial^2 (\cdot)}{\partial w^2}$$

## Connecting to Linear Response Theory

**3c**

Appendix A6:  
Connecting response theory to the Fokker-Planck operator

Theory Connection  
(FDT)

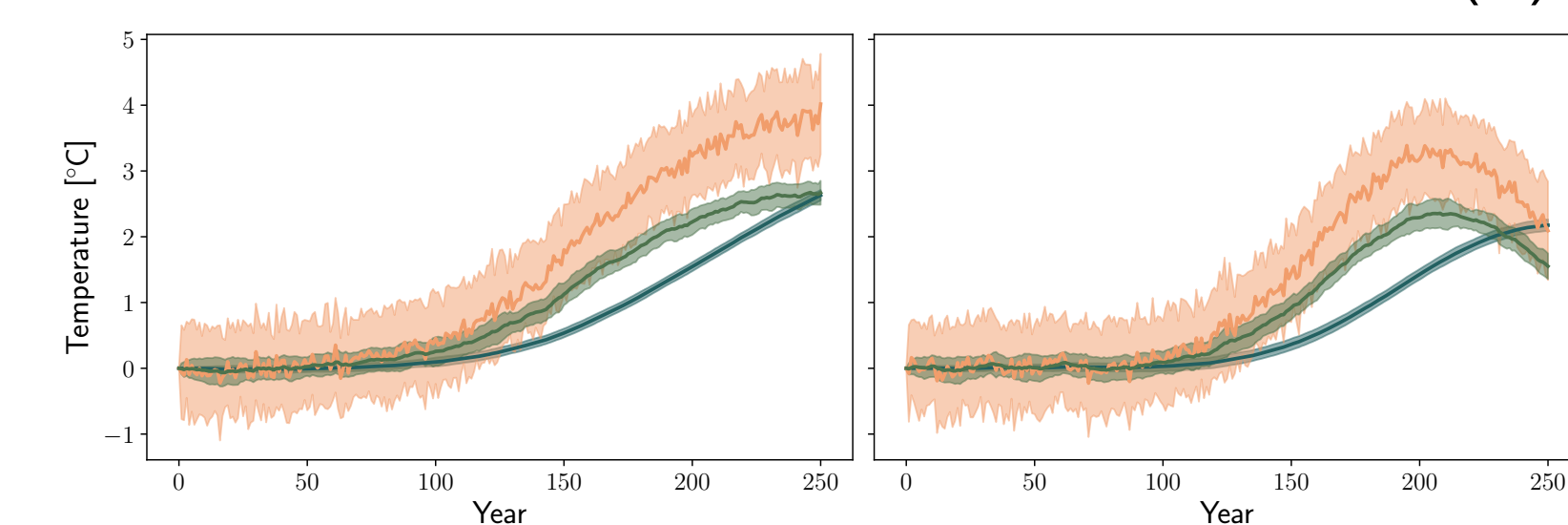
**3d**

Fluctuation Dissipation Theorem (FDT)

$$\langle g(w) \rangle = \langle g(w) \rangle_0 + \int_{-\infty}^t e^{\mathcal{K}s} F(t-s) ds$$

**4**

Emulate variable with new scenario(s)



## Section 2.2.3: Operator-Based Emulation

Method VI: EDMD

$$\phi(T_{n+1}) = \tilde{\mathcal{K}}\phi(T_n) + \psi(F_n)$$

Linear Basis

Method V: DMD

$$T_{n+1} = \mathcal{L}T_n + F_n$$

Quasi-Equilibrium

## Section 2.2.1: Pattern Scaling and its Immediate Extensions

General Quasi-Equilibrium

$$T(\mathbf{x}, t) = \mathcal{L}^{-1}(\mathbf{x}, \mathbf{x}') [F(\mathbf{x}', t)]$$

Forced with GMT

Method I: Pattern Scaling

$$T(\mathbf{x}, t) = a(\mathbf{x})\bar{T}(t)$$

## Section 2.3.2: Impulse Response Emulators

Method II: FDT

$$R(\mathbf{x}, \mathbf{x}', t) = \frac{\langle T_\delta(\mathbf{x}, t) - T_0(\mathbf{x}, t) \rangle}{|\delta(\mathbf{x}')|}$$

No Access to Governing Eq.

Method III: Deconvolution

$$\mathbf{R} = \frac{\mathbf{F}^{-1}\mathbf{T}}{\Delta t}$$

Exponential Decay

Method IV: Mode Fitting

$$R(t) = \sum_{i=1}^3 \alpha_i e^{\lambda_i t}$$

## Probabilistic Models

Bayesian inference, particle filters, deep-learning/diffusion models

Connect to FDT

Score-based response function

$$R(\mathbf{x}, \mathbf{x}', t) = -\langle T(\mathbf{x}, t) s(T(\mathbf{x}', 0)) \rangle$$