

Identifying disentangled climate prototypes of forest mortality with generative deep learning

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Introduction

- Forest are global carbon sink
- Climate change leads to higher forest mortality
- Forest mortality is complex
 - Multiple drivers
 - Long temporal scale
 - State dependency

Introduction

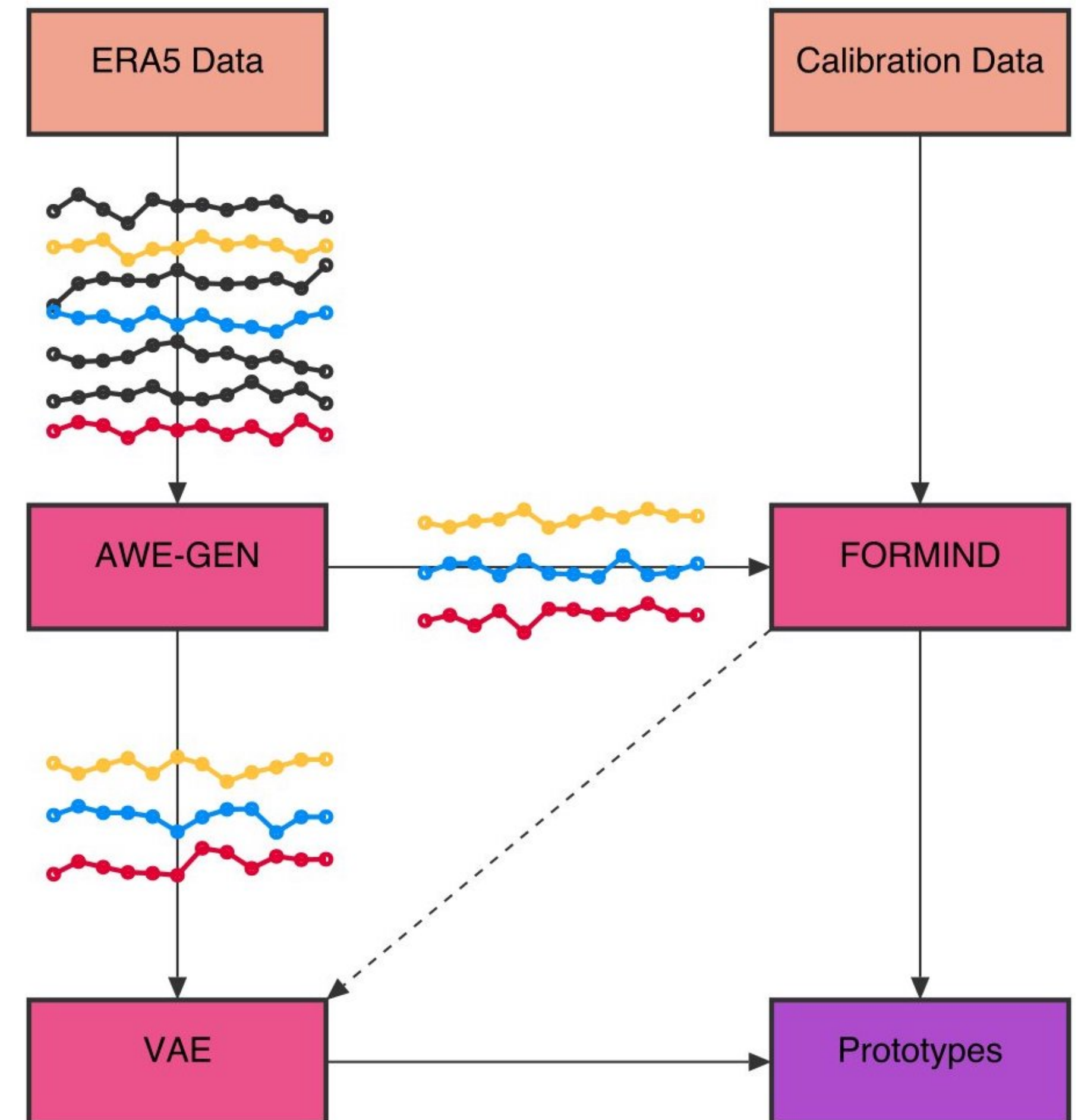
Why?

- Understanding the climatic drivers of forest mortality
 - How forest mortality will change with changing climate?
 - Important for forest management and policy makers

Methodology

Flow

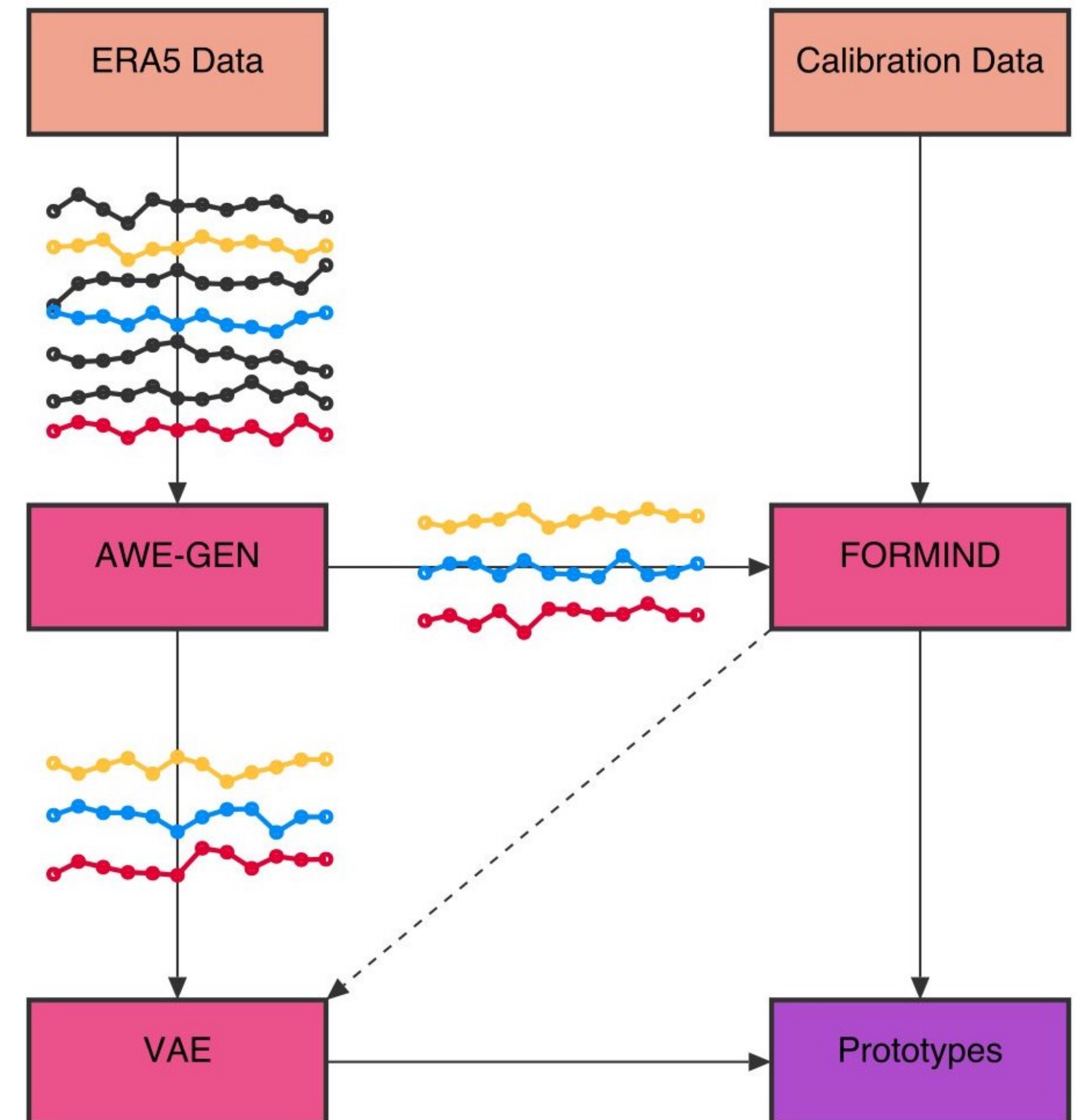
- ERA5 reanalysis dataset
- Variables
 - Precipitation
 - Cloud cover
 - Shortwave-radiation
 - Relative humidity
 - Wind speed
 - Atmospheric pressure
- 1979-2019



Methodology

Flow

- AWE-GEN
 - Stochastic weather generator
 - Radiation, Precipitation and Temperature
- FORMIND
 - Process and individual based forest model
- Variational Autoencoders (VAE)
 - Generative deep learning model



Methodology

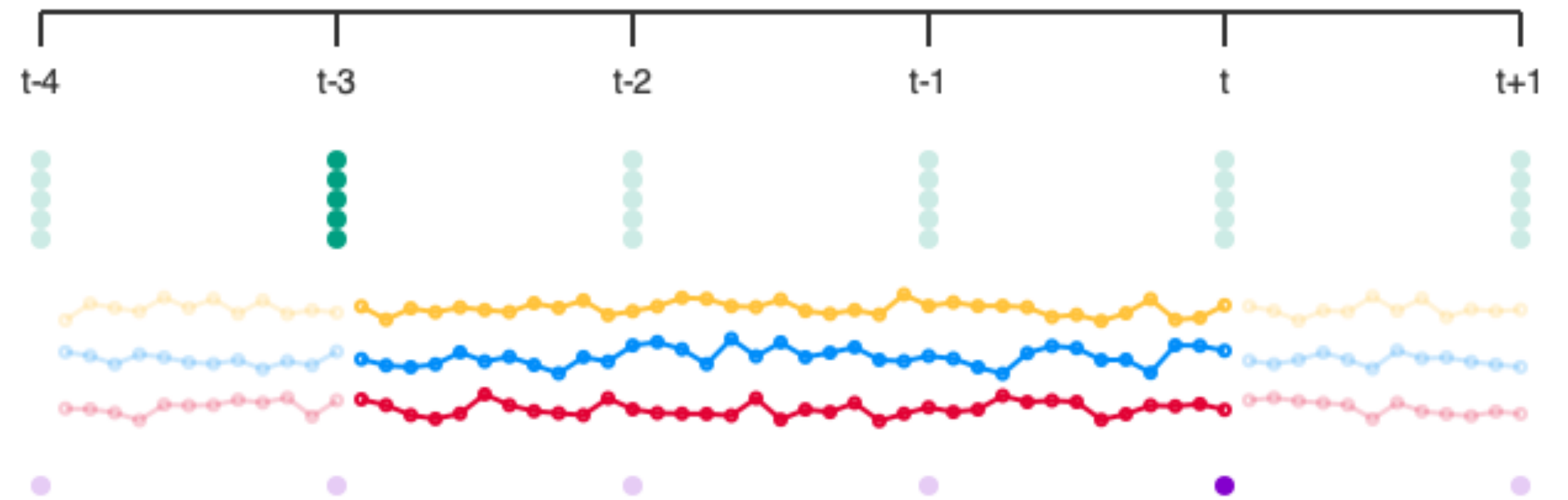
Data Preparation

- FORMIND Calibration
 - Hainich Beech Forest, Germany
 - 51.08 degrees N, 10.51 degrees E
- 3 separate simulations
 - Beech, Pine and Spruce
 - Burn in period 2000 years
 - 160,000 years of simulations
 - Area 200 ha

Methodology

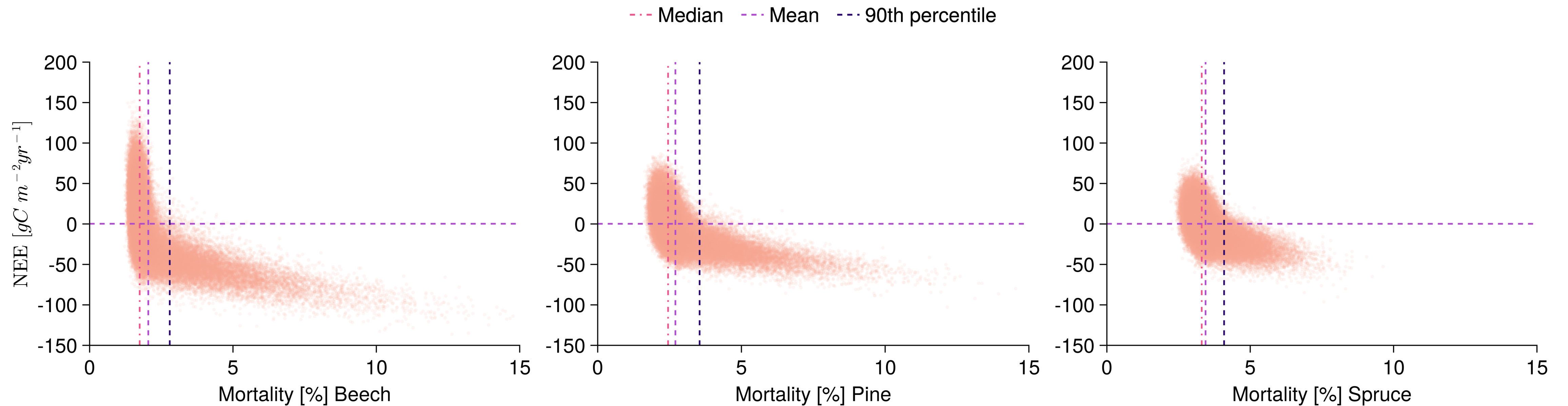
Data Preparation

- State variables (x_s)
 - Yearly values - $year_{t-3}$
- Dynamic variables (x_d)
 - Monthly values - $year_{t-2}$, $year_{t-1}$, and $year_t$
- Biomass Mortality Rate (MBR) (y)
 - Yearly Value - $year_t$
 - Binary (90th percentile)



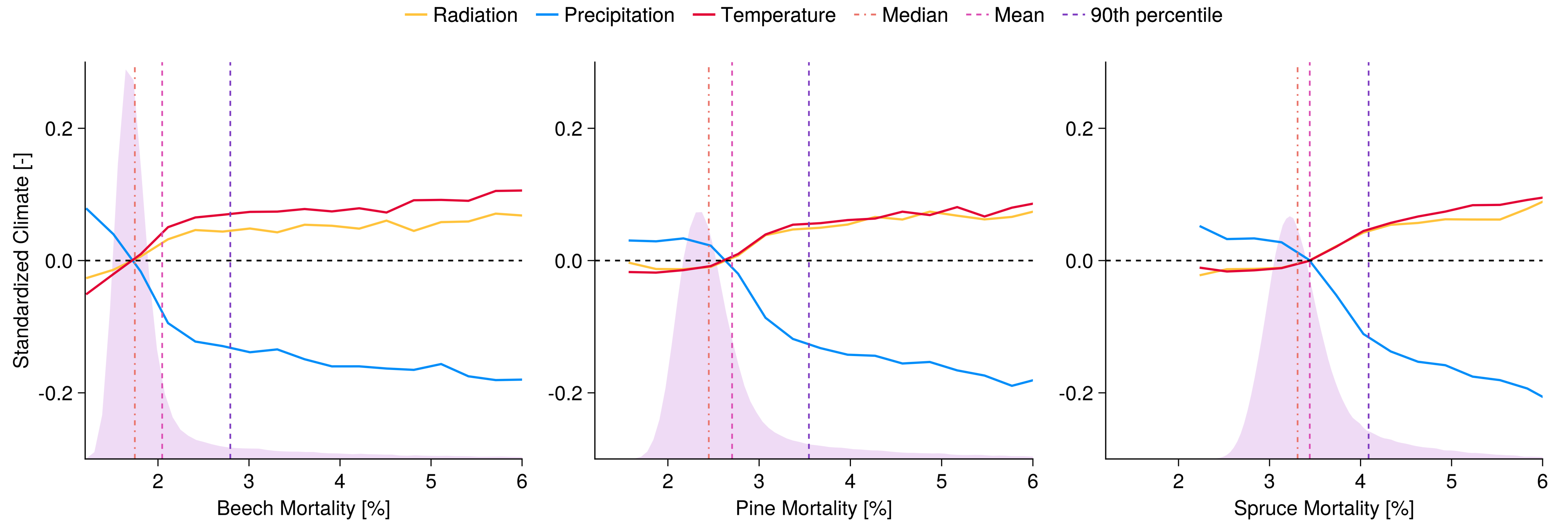
FORMIND Simulation

Net Ecosystem Exchange vs Mortality



FORMIND Simulation

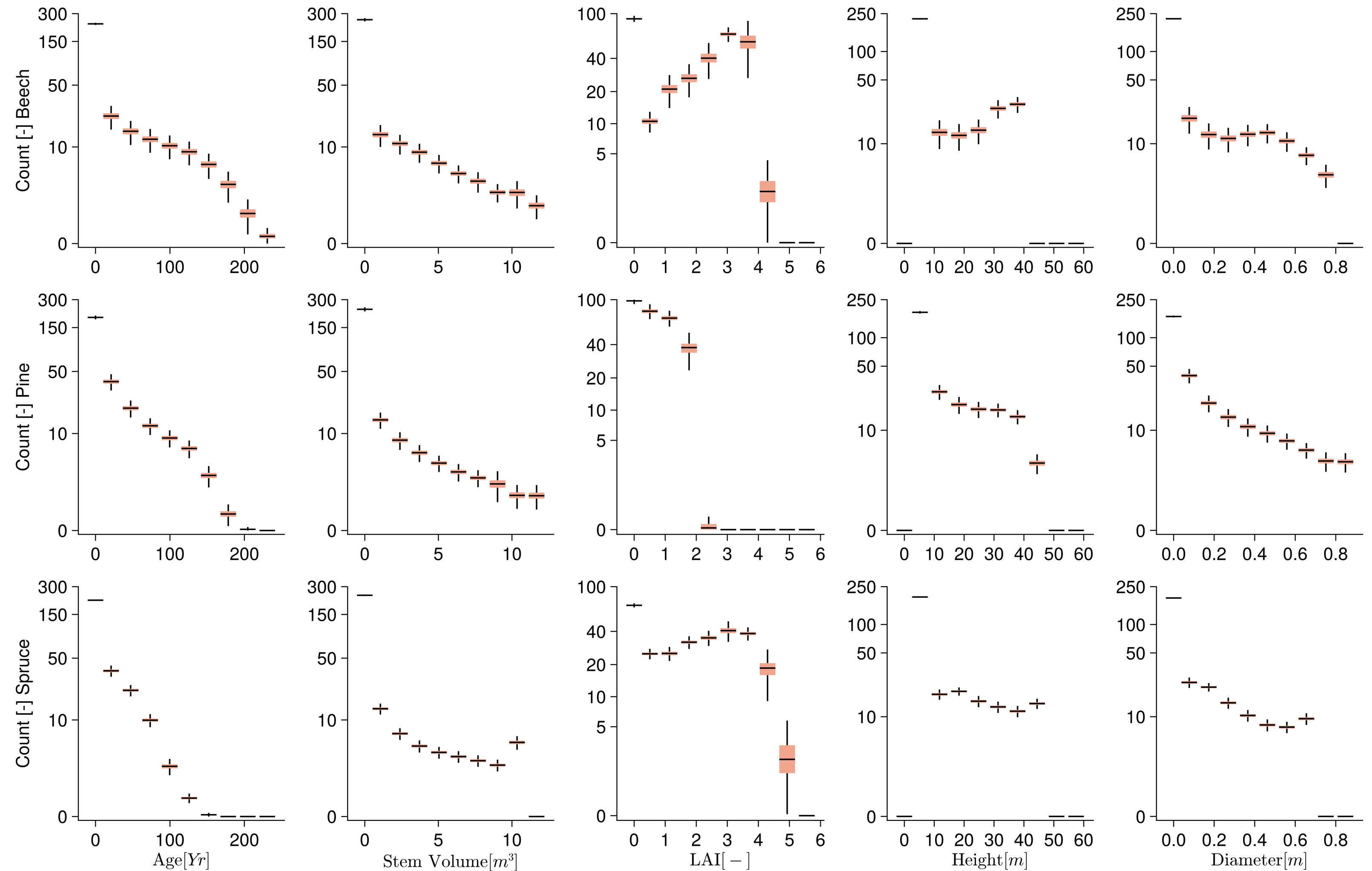
Forest Mortality



FORMIND Simulation

State Variables

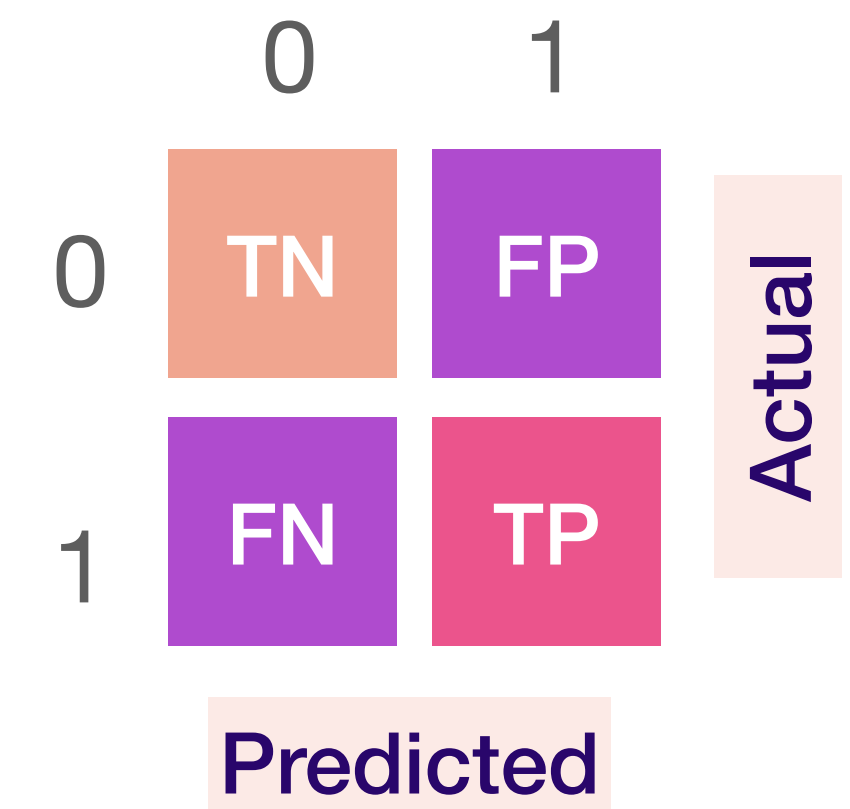
- State variables are PFT dependent
- LAI has a very different pattern



Extreme Mortality

Logistic Regression

Metrics	CSI	f1-score	Average Precision
Beech Logistic Regression with x_d	0.43	0.60	0.65
Beech Logistic Regression with x_d and x_s	0.53	0.69	0.77
Pine Logistic Regression with x_d	0.46	0.64	0.70
Pine Logistic Regression with x_d and x_s	0.60	0.75	0.85
Spruce Logistic Regression with x_d	0.46	0.63	0.69
Spruce Logistic Regression with x_d and x_s	0.59	0.74	0.84



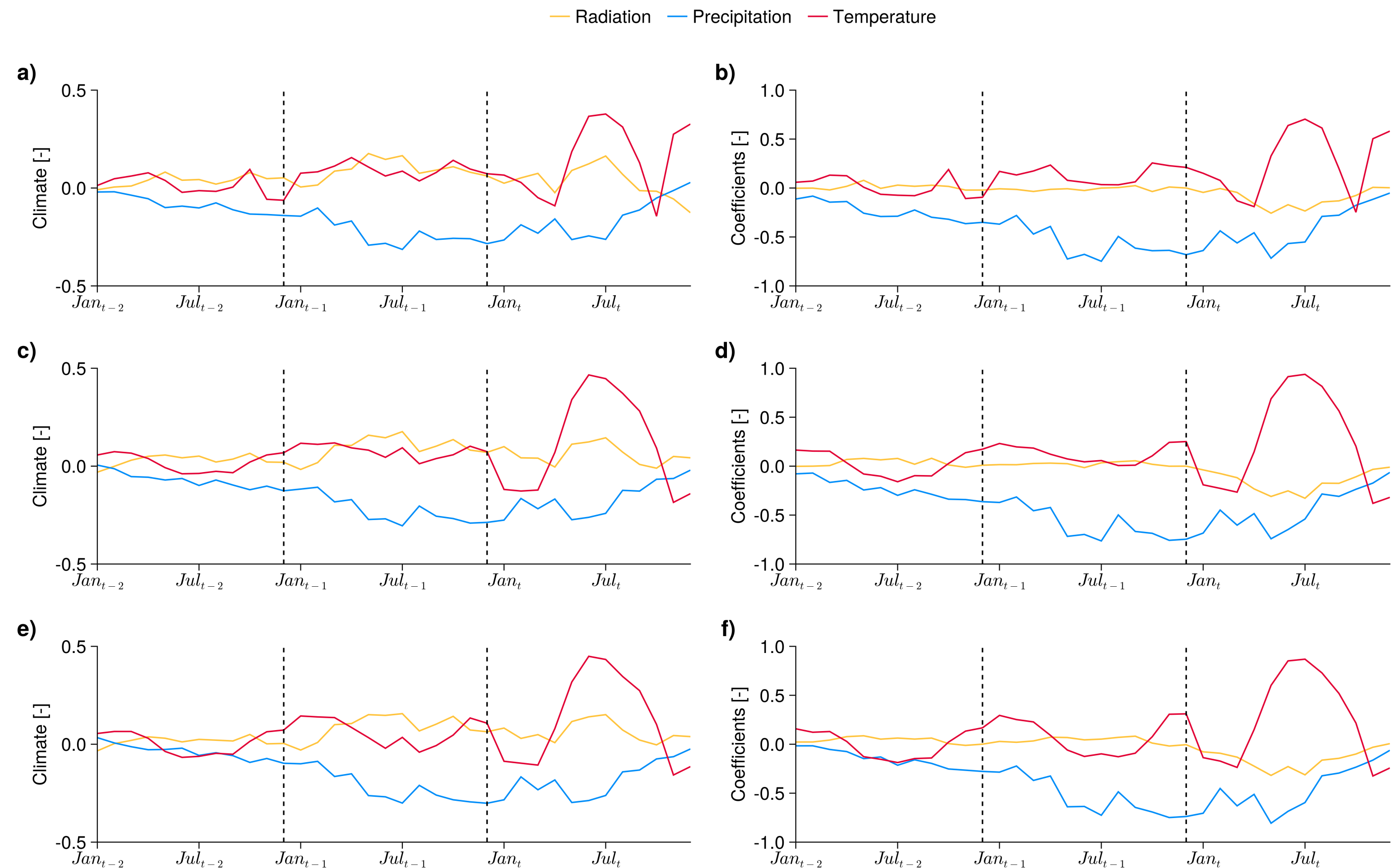
- Predictability
 - Spruce == Pine >> Beech
 - Static variables help

$$CSI = \frac{TP}{TP + FP + FN}$$

Extreme Mortality

Composites & Logistic Regression

- Extreme mortality > 90th percentile
- Drier years with higher temperature and solar radiation
- Pine and Spruce has similar behaviour



Composites are nice

We want to know more!

- Composite and Logistic Regression show one dominant (superimposed) mechanisms for extreme forest mortality
- Pathways leading to high forest mortality

VAE

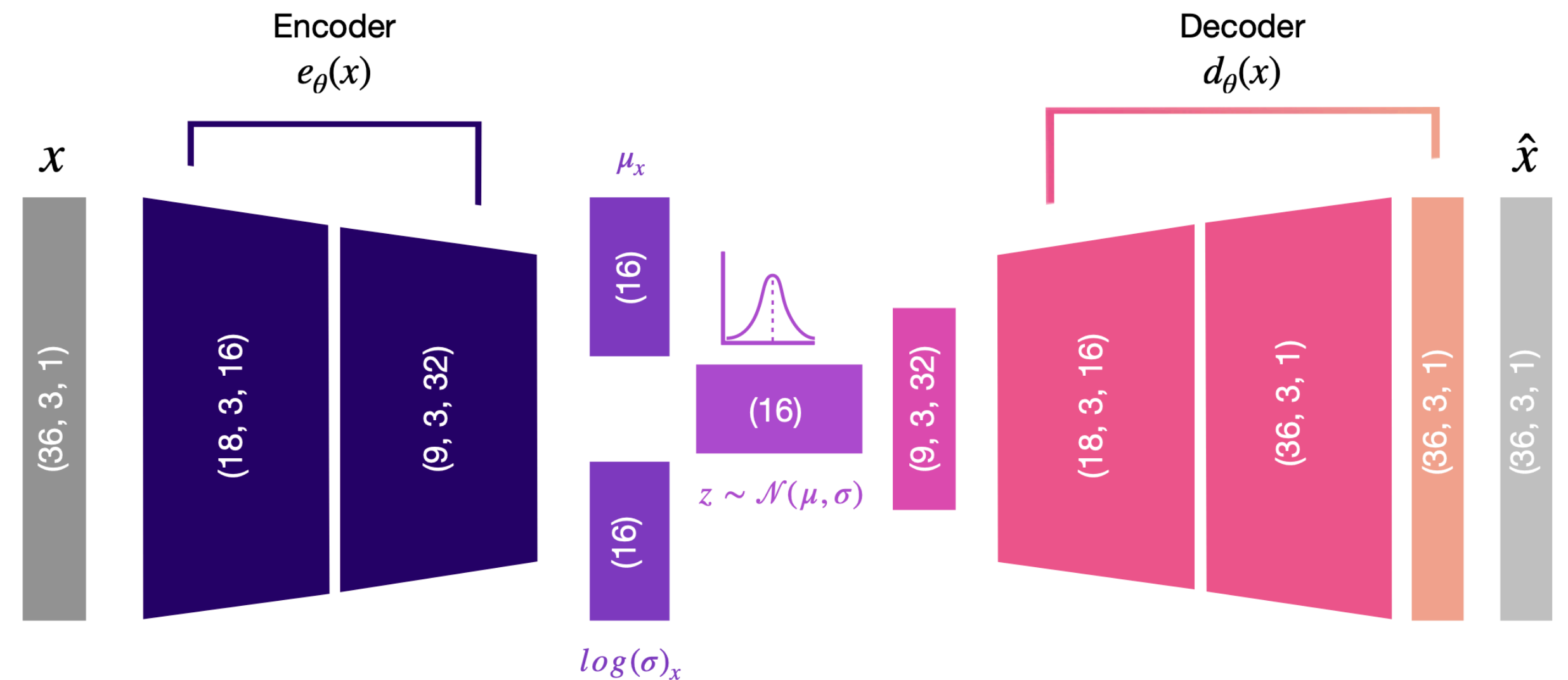
Generative Modelling Approach

- VAE has two parts
 - Encoder
 - Decoder
- Input space is reduced to latent space (Encoder)
- Reconstruction is done again from the latent space (Decoder)
- Latent dimensions are independent and normal (approximately)

VAE

Architecture

- Input
 - 3 years (monthly), radiation, temperature and precipitation
- 16 latent dimensions
- Loss has 2 components



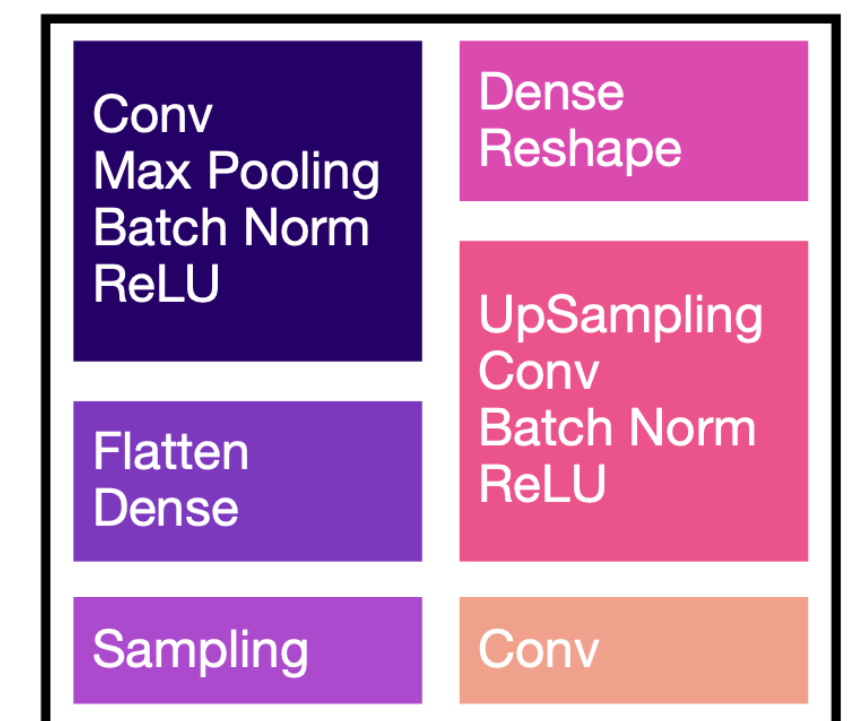
ReconstructionLoss

$$\|x - \hat{x}\|_2 = \|x - d_\phi(z)\|_2 = \|x - d_\phi(\mu_x + \sigma_x \epsilon)\|_2$$

KLDivergence

$$D_{KL}(\mathcal{N}(\mu_x, \sigma_x) \parallel \mathcal{N}(0, I))$$

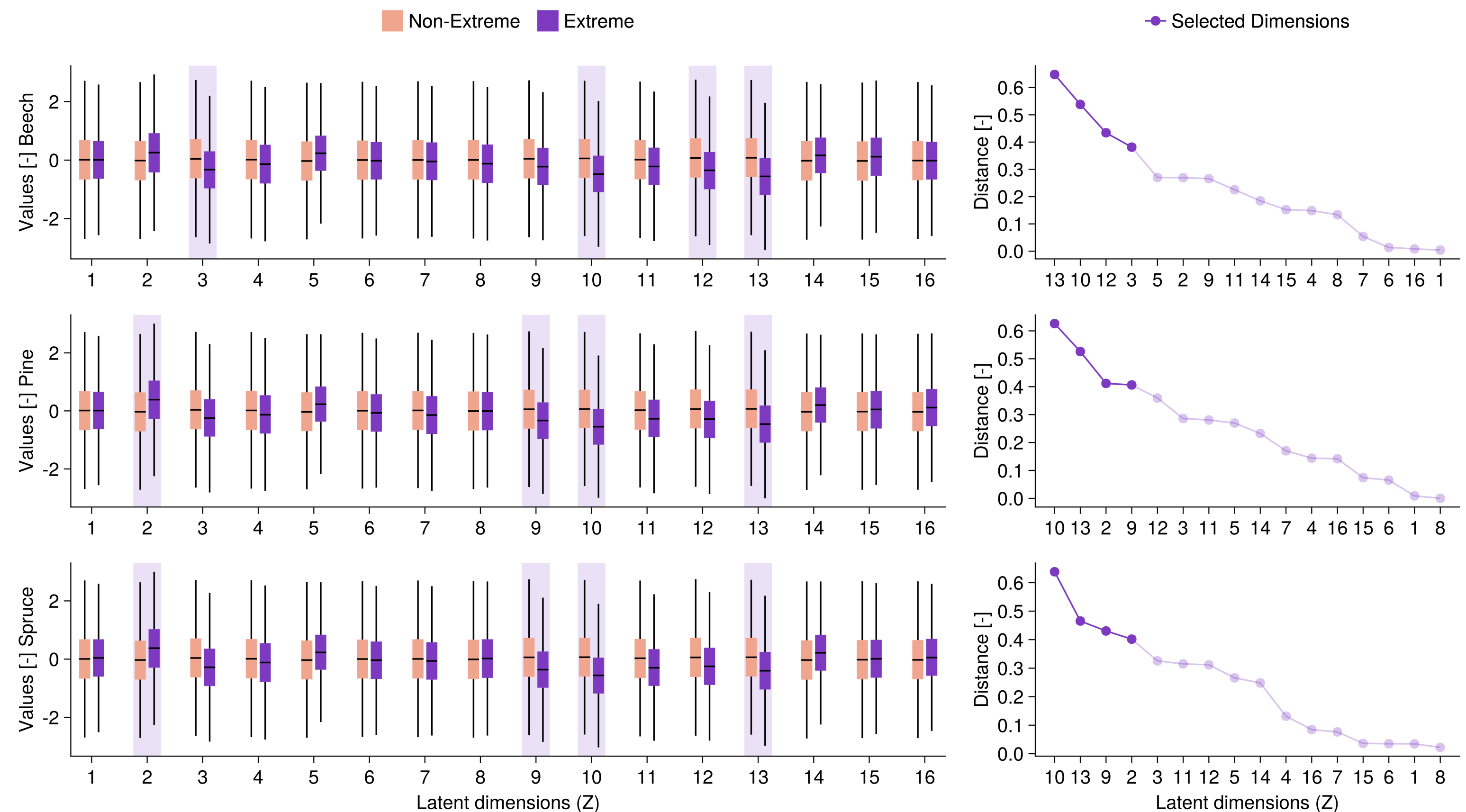
Total Loss = ReconstructionLoss + β x KLDivergence



Results

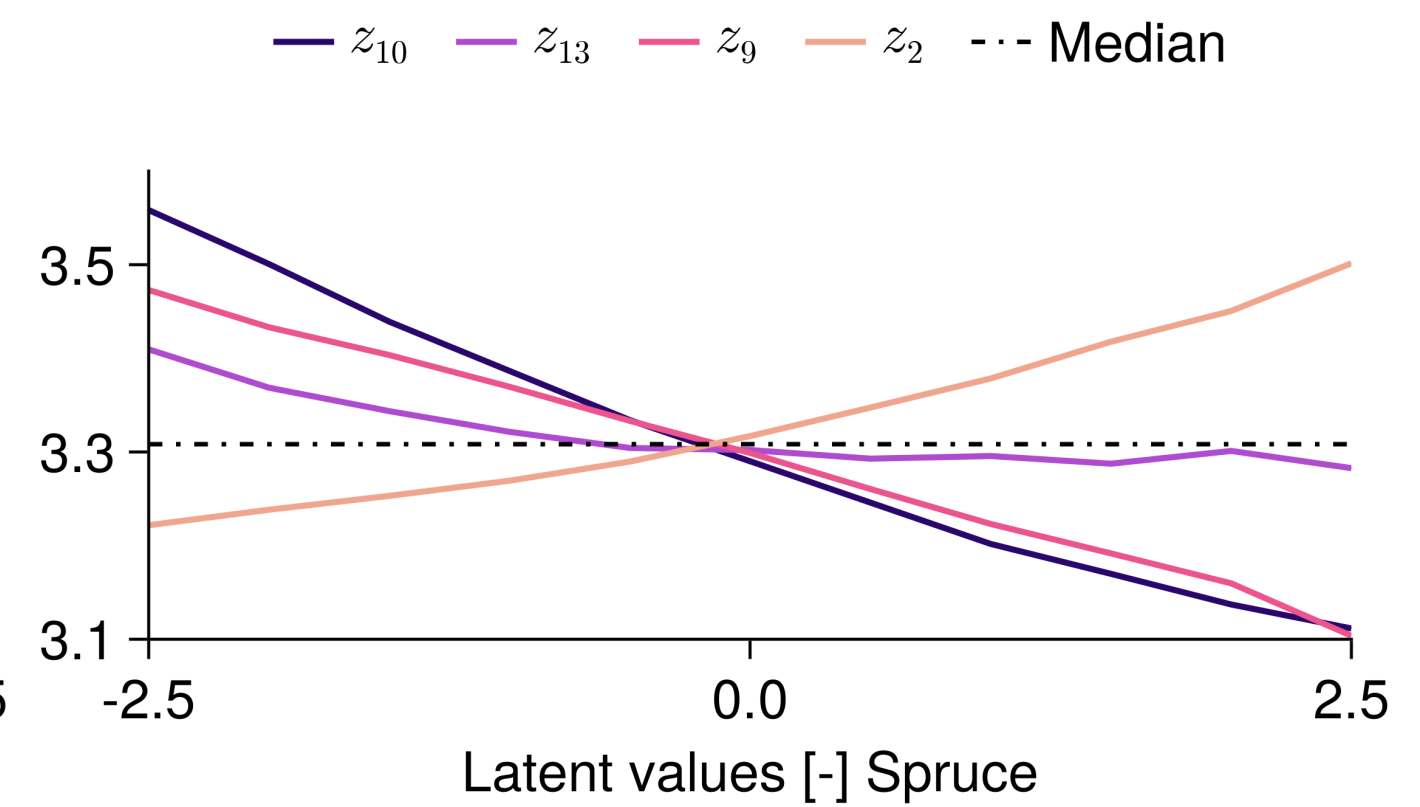
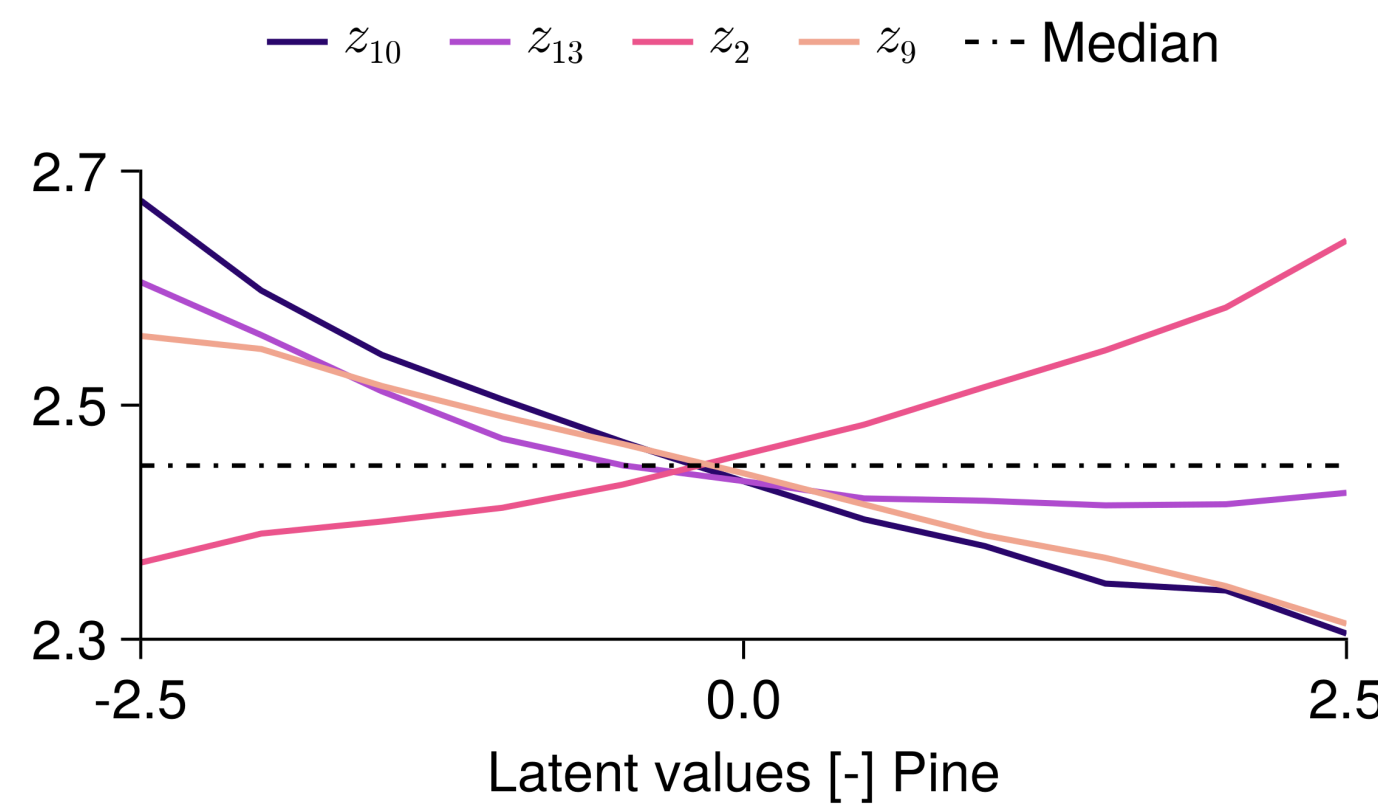
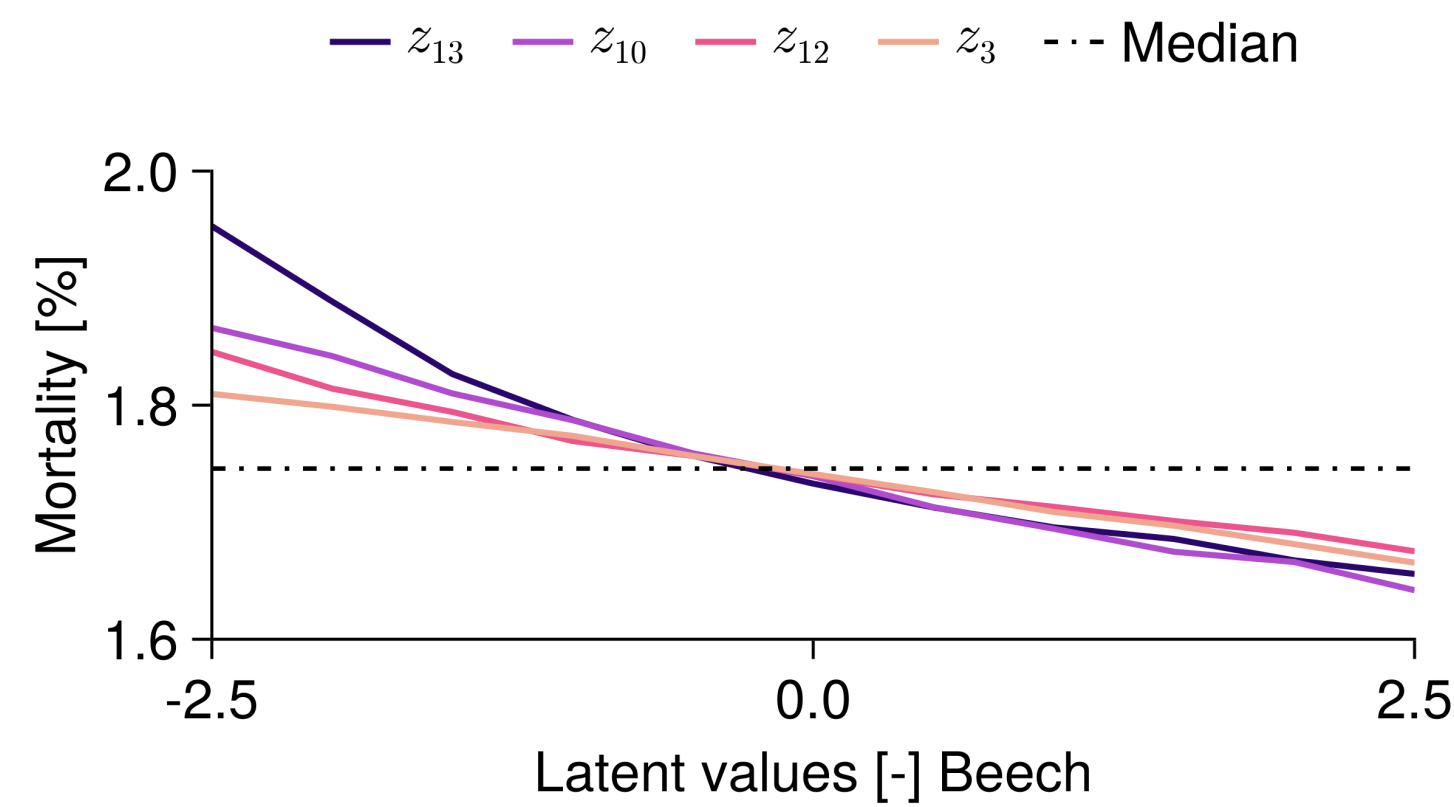
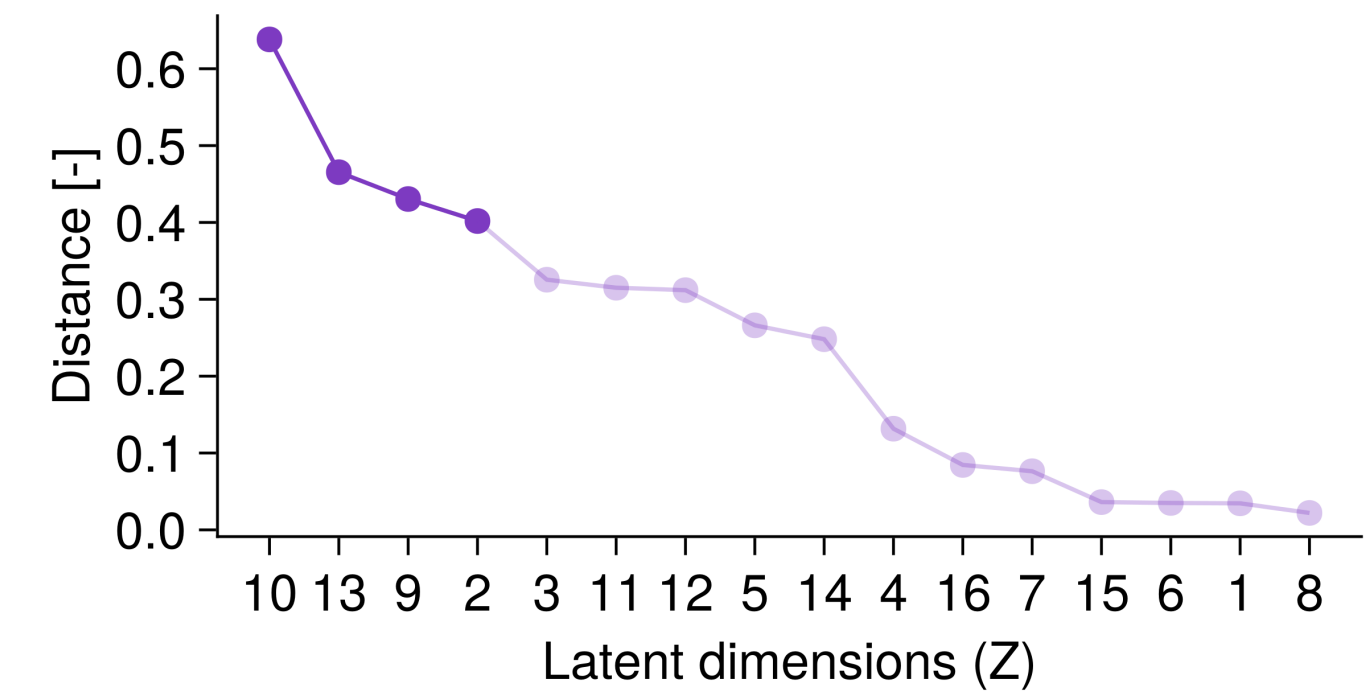
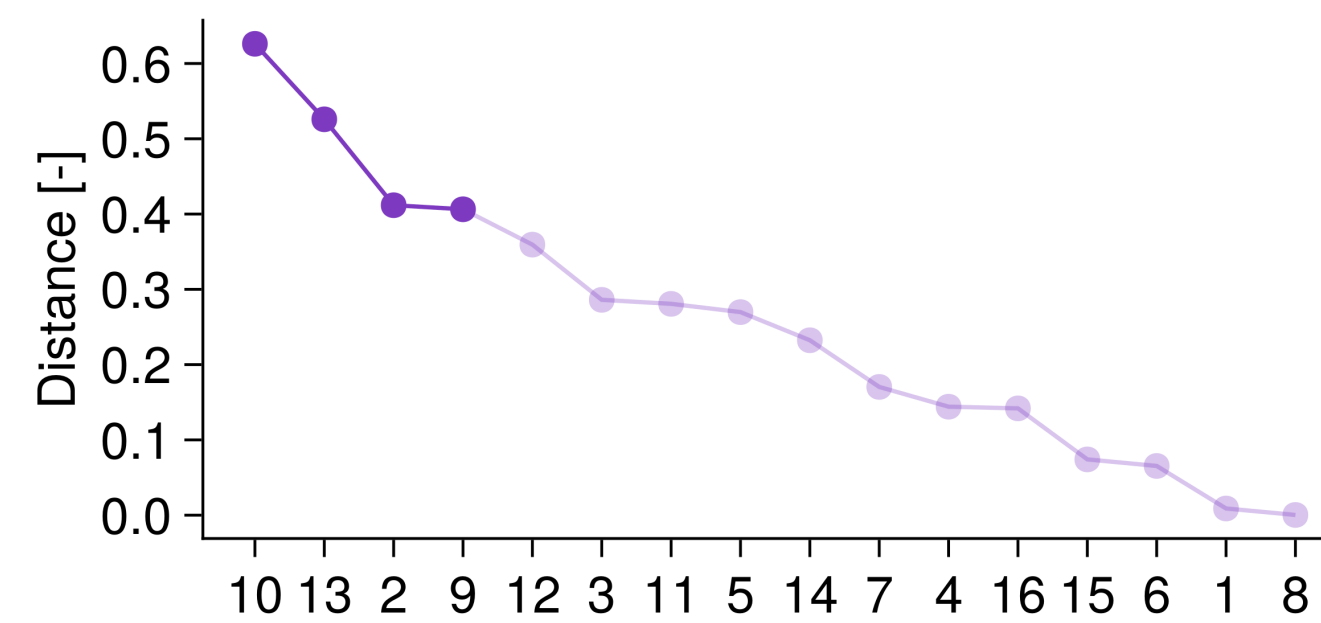
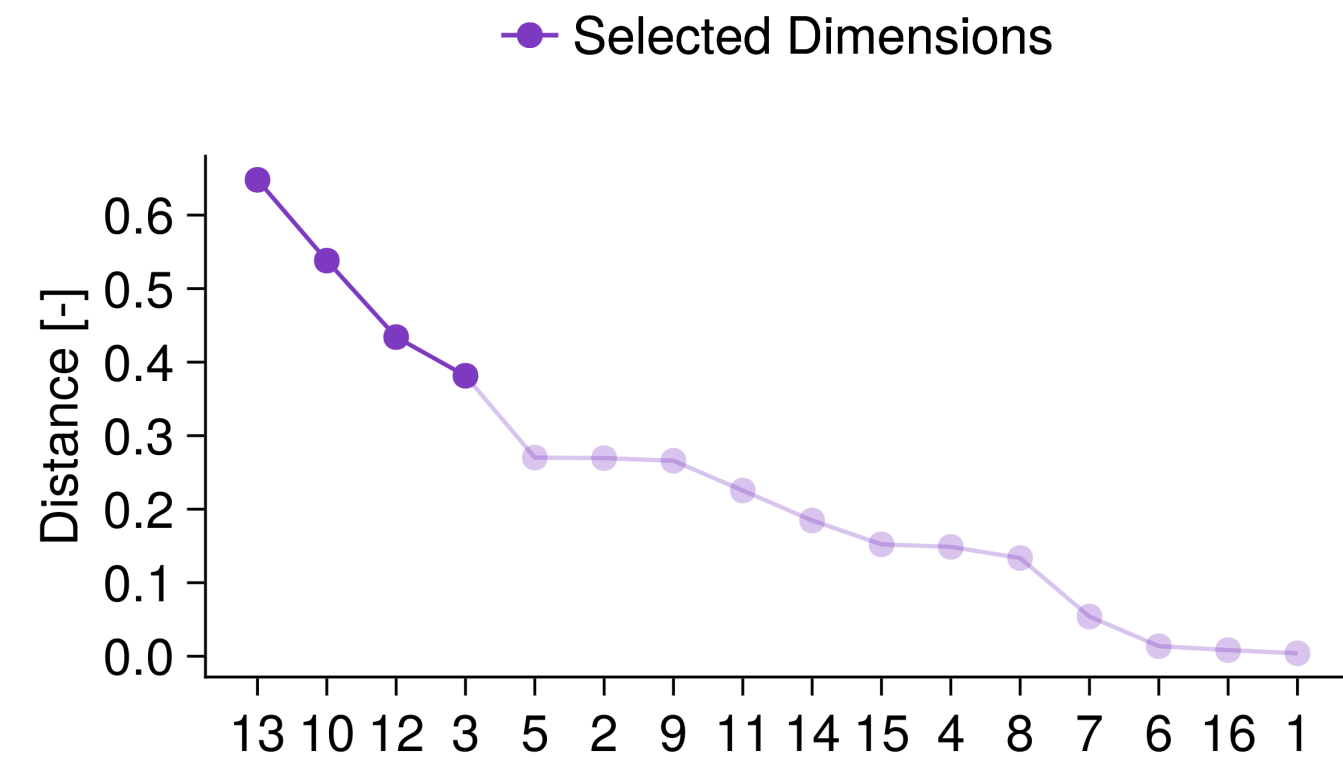
Latent Dimensions

- Extreme vs Non-Extreme latent values
- Difference between latent values
- Pine and Spruce have similar behaviour



Results

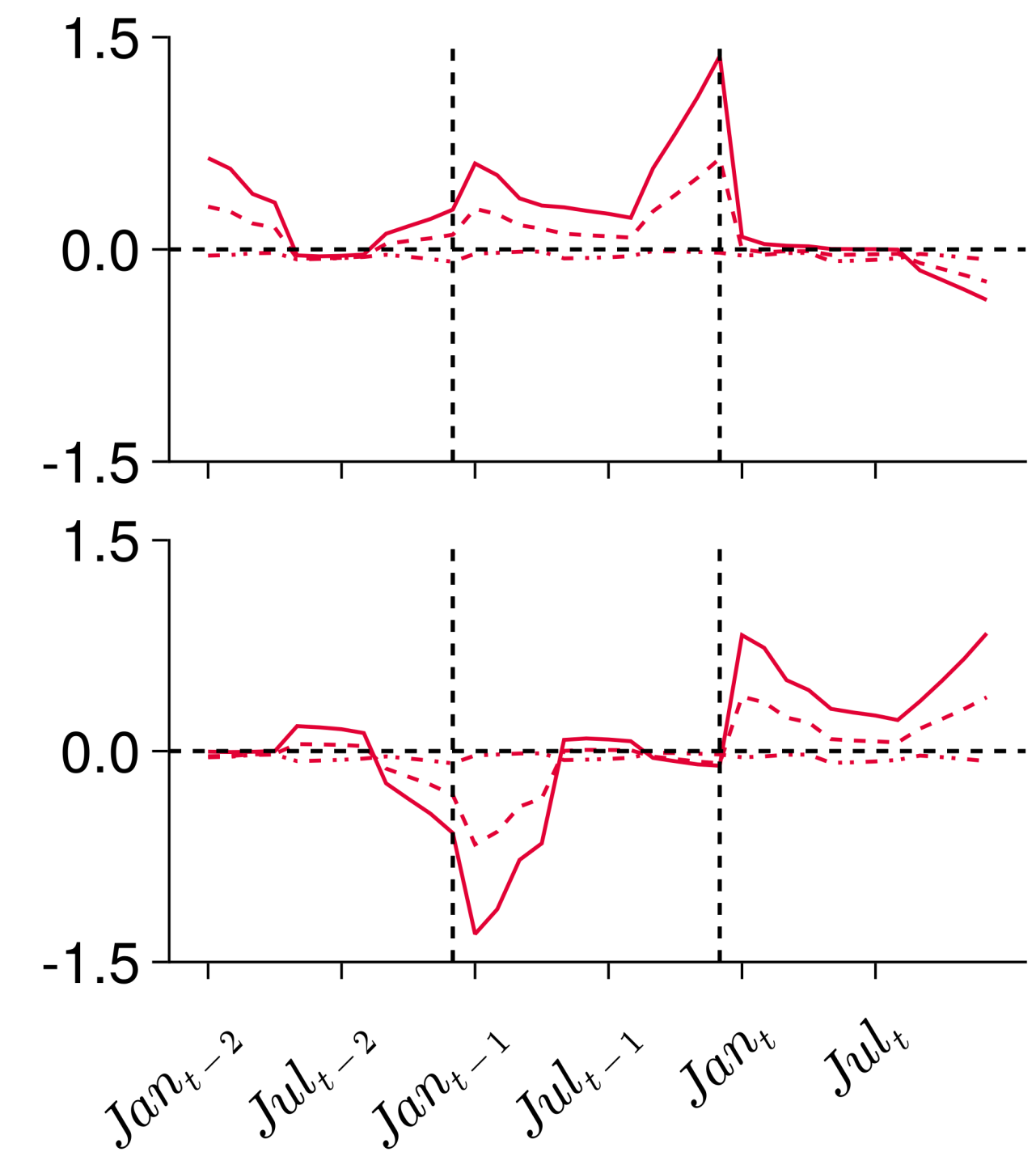
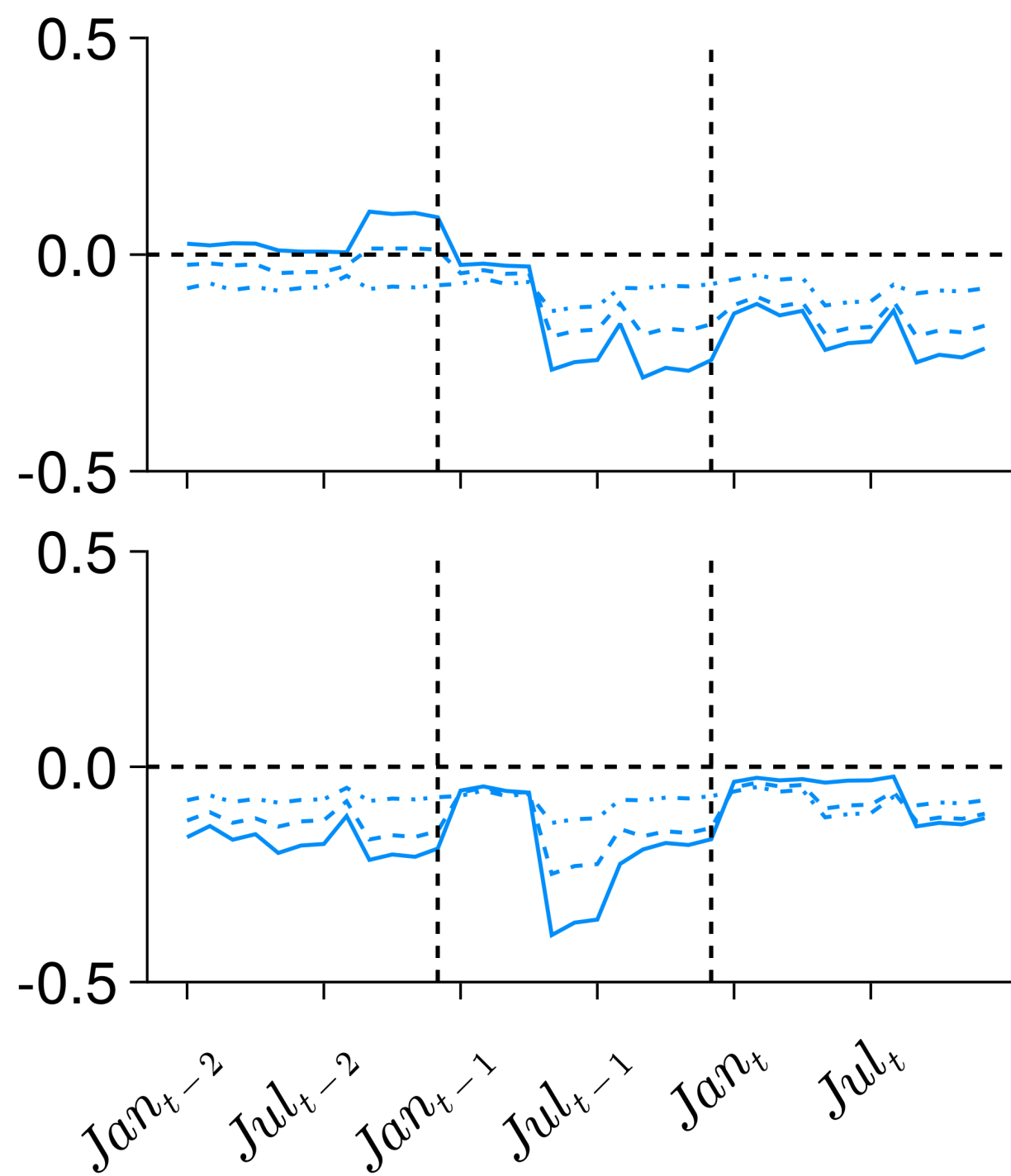
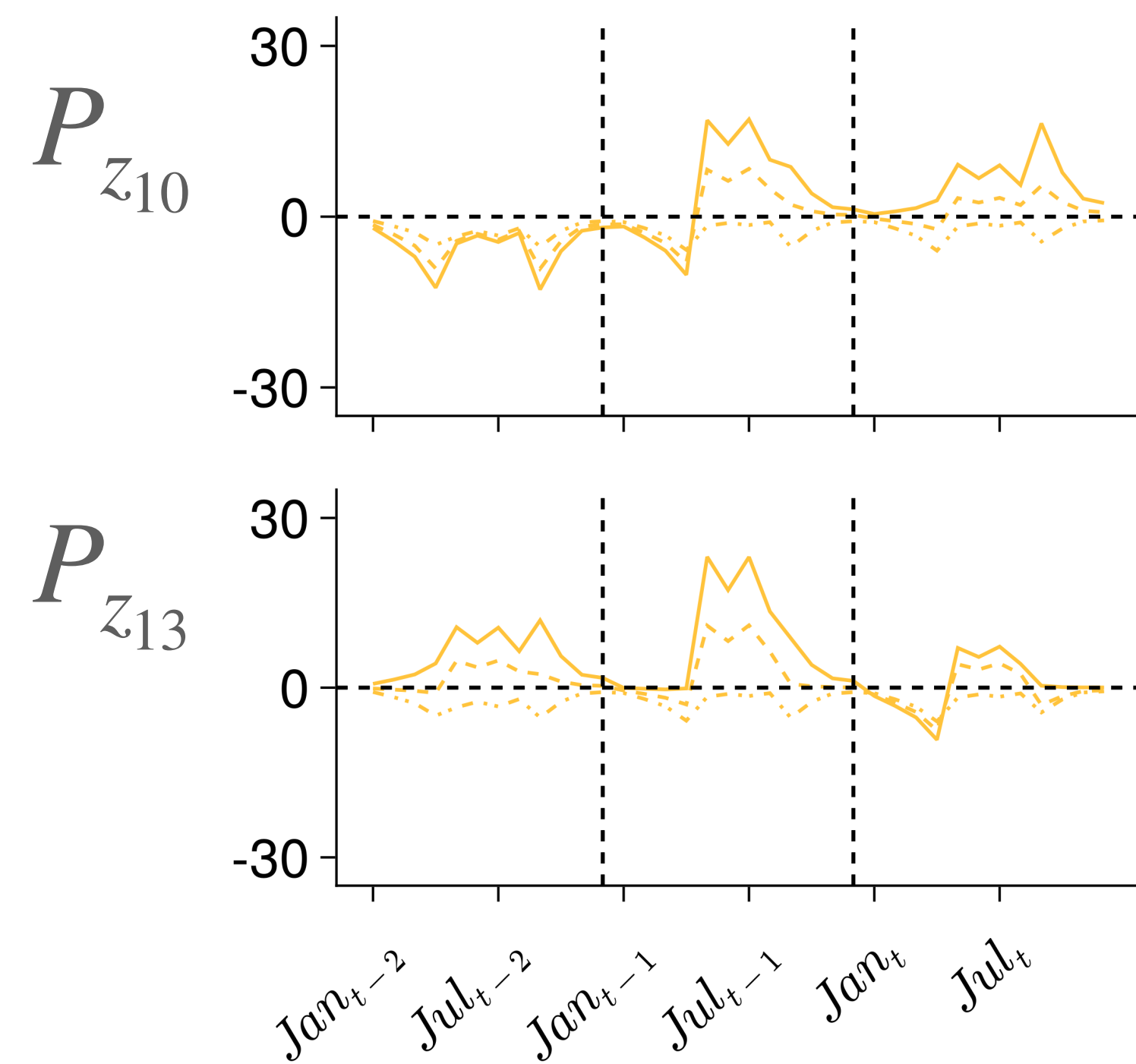
Latent dimensions vs Mortality



Results

All

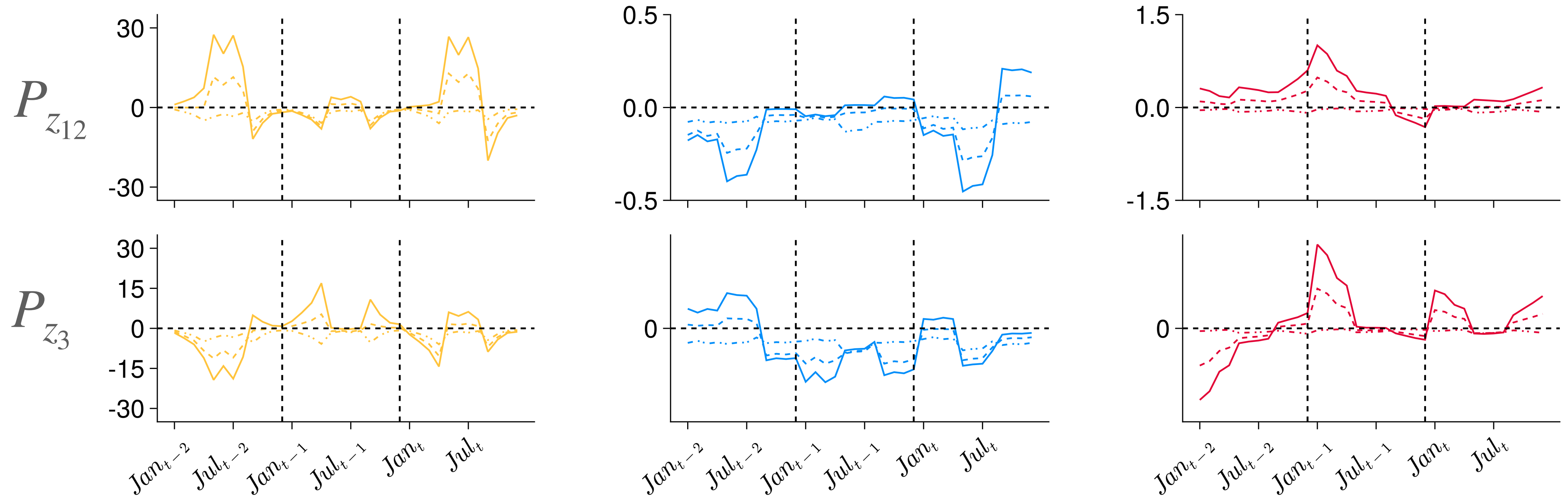
— Radiation — Precipitation — Temperature - - - $Pz_i^{0\sigma}$ - - - $Pz_i^{1\sigma}$ — $Pz_i^{2\sigma}$



Results

Beech

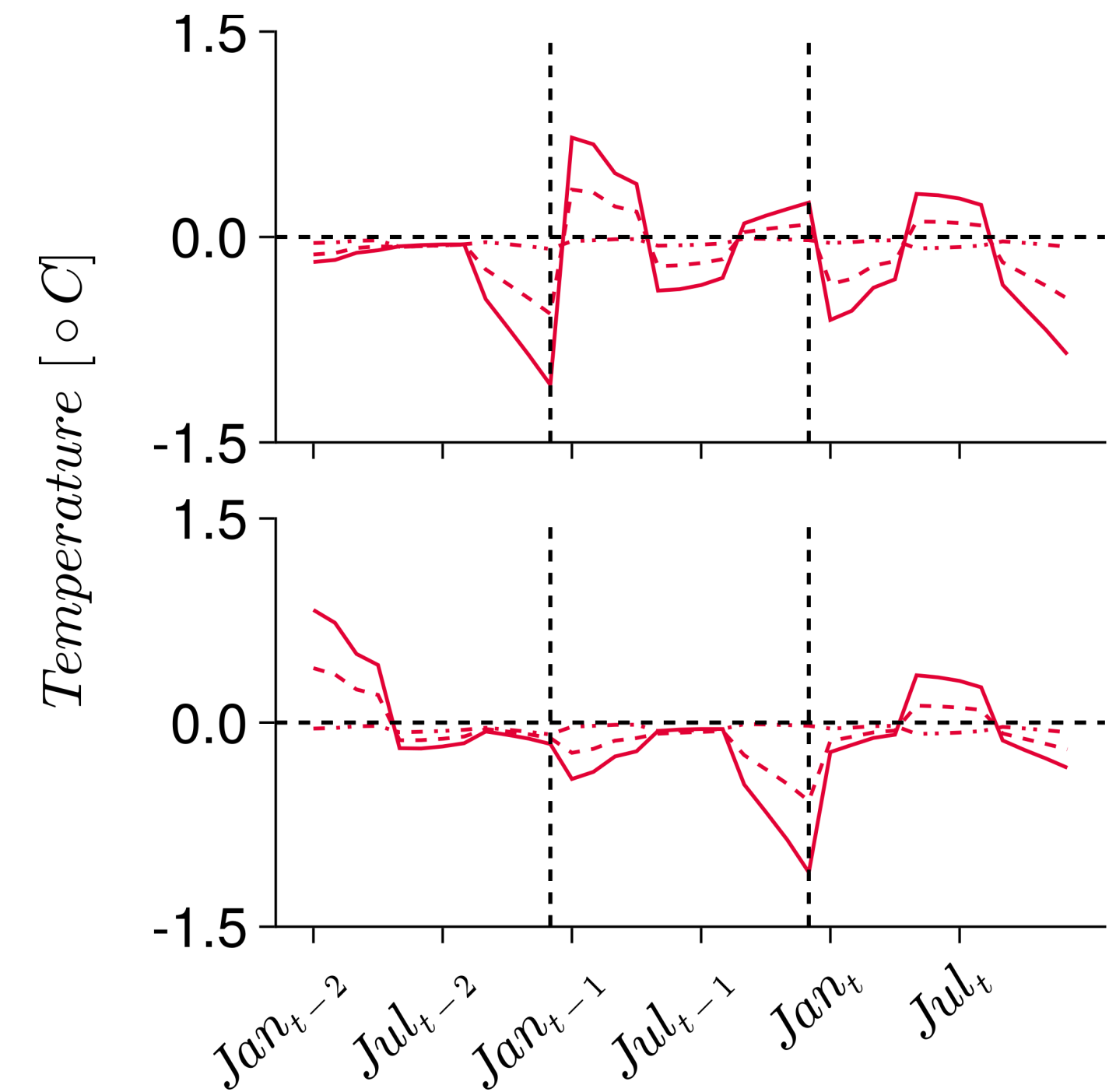
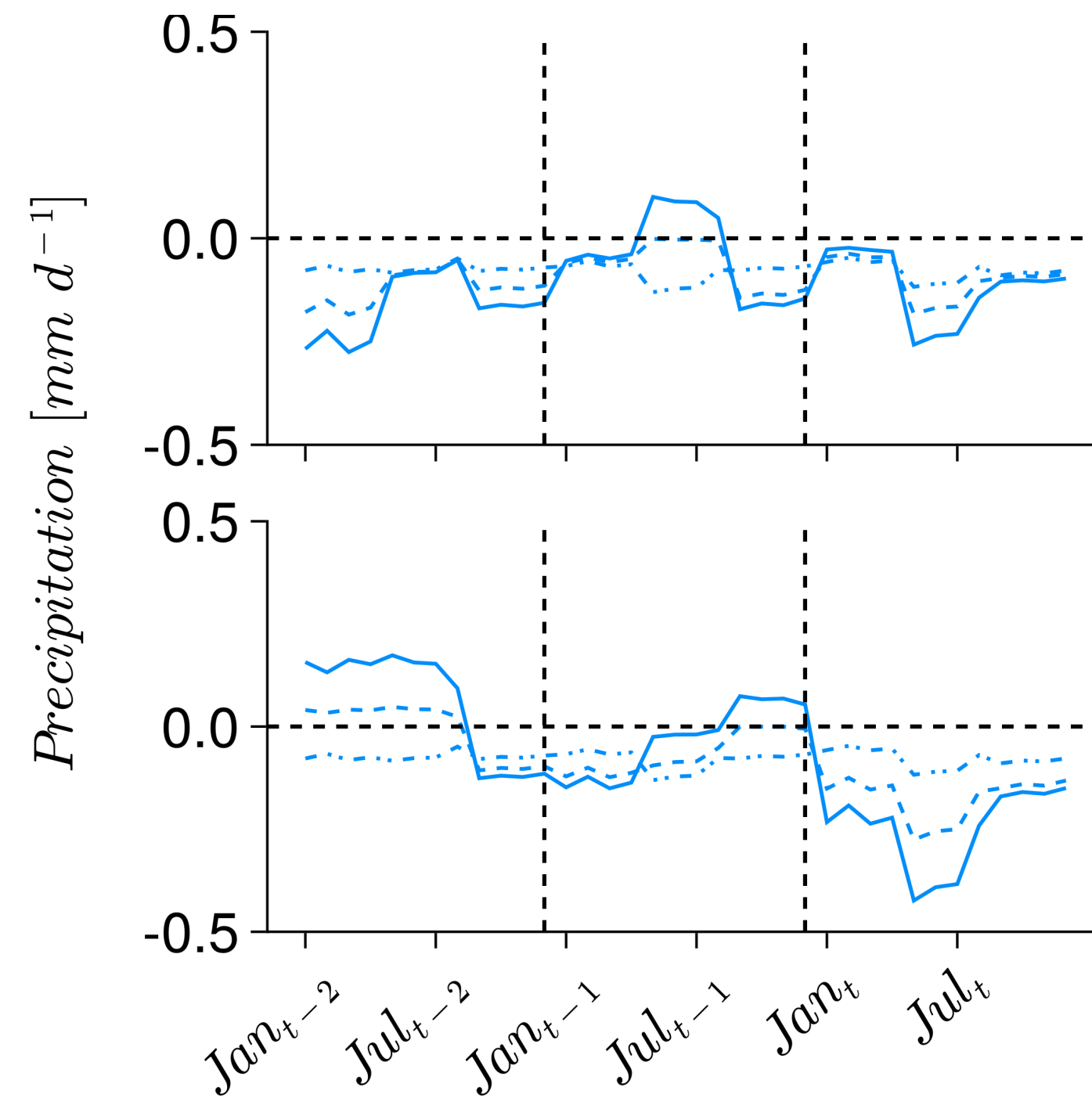
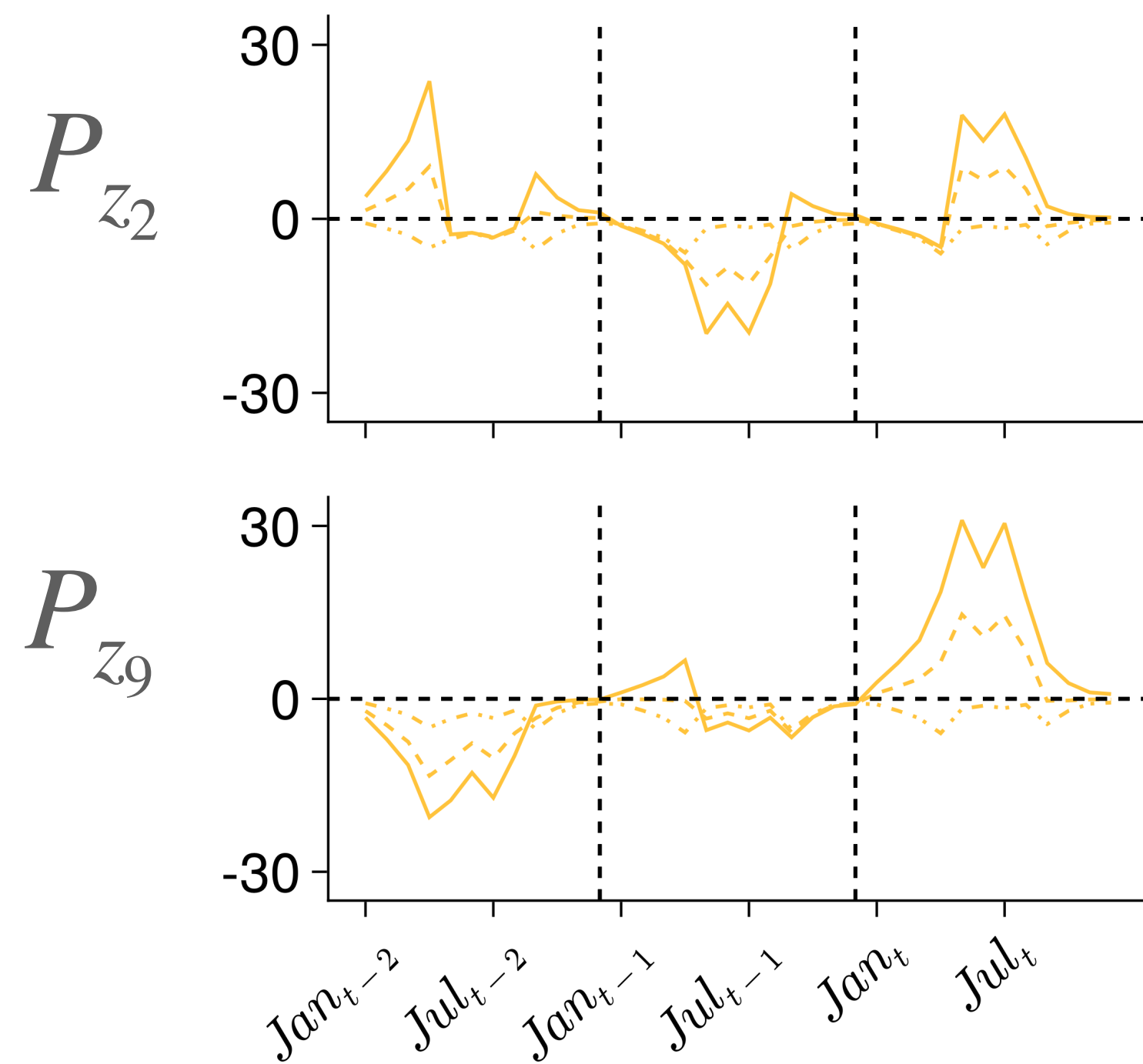
— Radiation — Precipitation — Temperature - - - $Pz_i^{0\sigma}$ - - - $Pz_i^{1\sigma}$ — $Pz_i^{2\sigma}$



Results

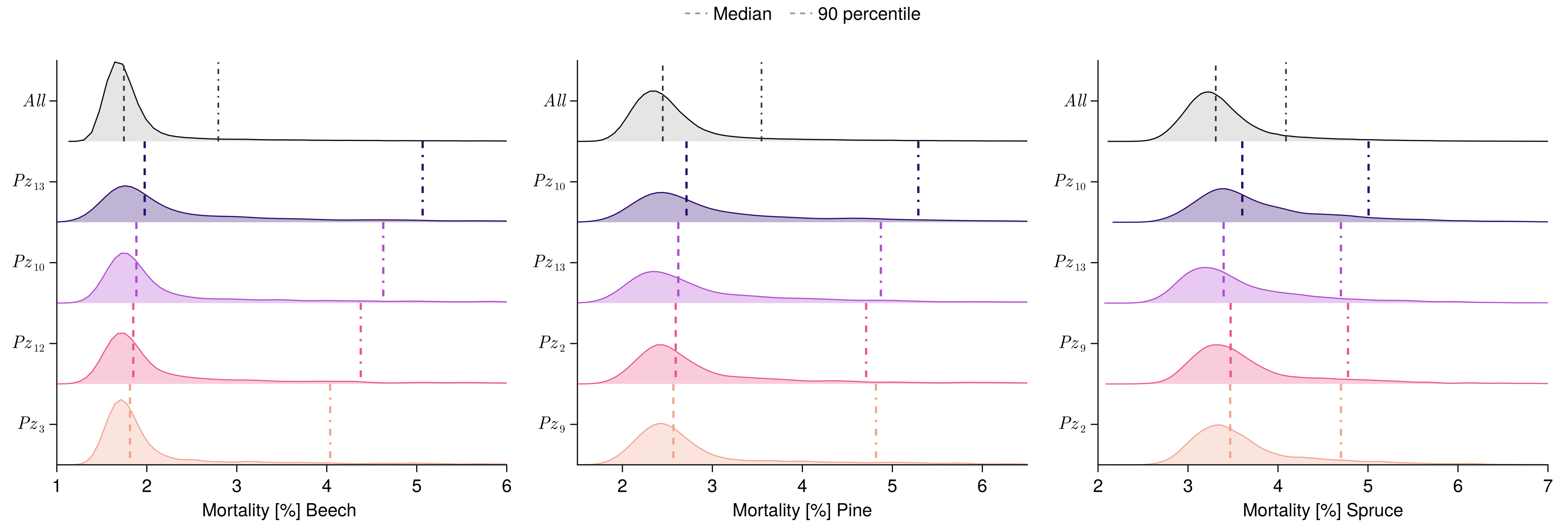
Pine and Spruce

— Radiation — Precipitation — Temperature — $P_{z_i}^{0\sigma}$ - - - $P_{z_i}^{1\sigma}$ — $P_{z_i}^{2\sigma}$



Results

Mortality density



Conclusion

- Forest mortality is a complex phenomenon
- Weather far in past matters for mortality
- State variables matter
- Composite analysis/Logistic regression identifies the dominant average
- VAEs identify climate prototypes of forest mortality
- Identified prototypes are associated with higher forest mortality (on average)

Thank You!