

# STC-ViT: Spatio Temporal Continuous Vision Transformer for Weather Forecasting

Hira Saleem<sup>1</sup>, Prof. Flora Salim<sup>1</sup>, Dr. Cormac Purcell<sup>1,2</sup>

## Introduction

Operational weather forecasting system relies on computationally expensive physics-based models. Recently, transformer-based models have shown remarkable potential in weather forecasting achieving state-of-the-art results. However, transformers are discrete models which limit their ability to learn the continuous spatio-temporal features of the dynamical weather system. We address this issue with STC-ViT, a Spatio-Temporal Continuous Vision Transformer for weather forecasting. STC-ViT incorporates the continuous time Neural ODE layers with multi-head attention mechanism to learn the continuous weather evolution over time.

## Aims

The main aim of this study is to predict extreme events at high resolution. To achieve this aim future studies would look at:

- Climate Emulation
- Weather and climate downscaling
- Explainable Artificial Intelligence

## Methods

We propose STC-ViT which leverages the continuous learning paradigm to effectively learn the complex spatio-temporal changes even from weather data recorded at coarser resolution. The idea is to parameterize the attention mechanism by converting it into a differentiable function. Continuous temporal attention is calculated sample-wise and combined with the patch wise spatial attention to learn the spatio-temporal mapping of weather variables in the embedding space of the vision transformer. Furthermore, we add derivation as a pre-processing step to prepare the discrete data for continuous model and explore the role of normalization in continuous modelling.

## Results

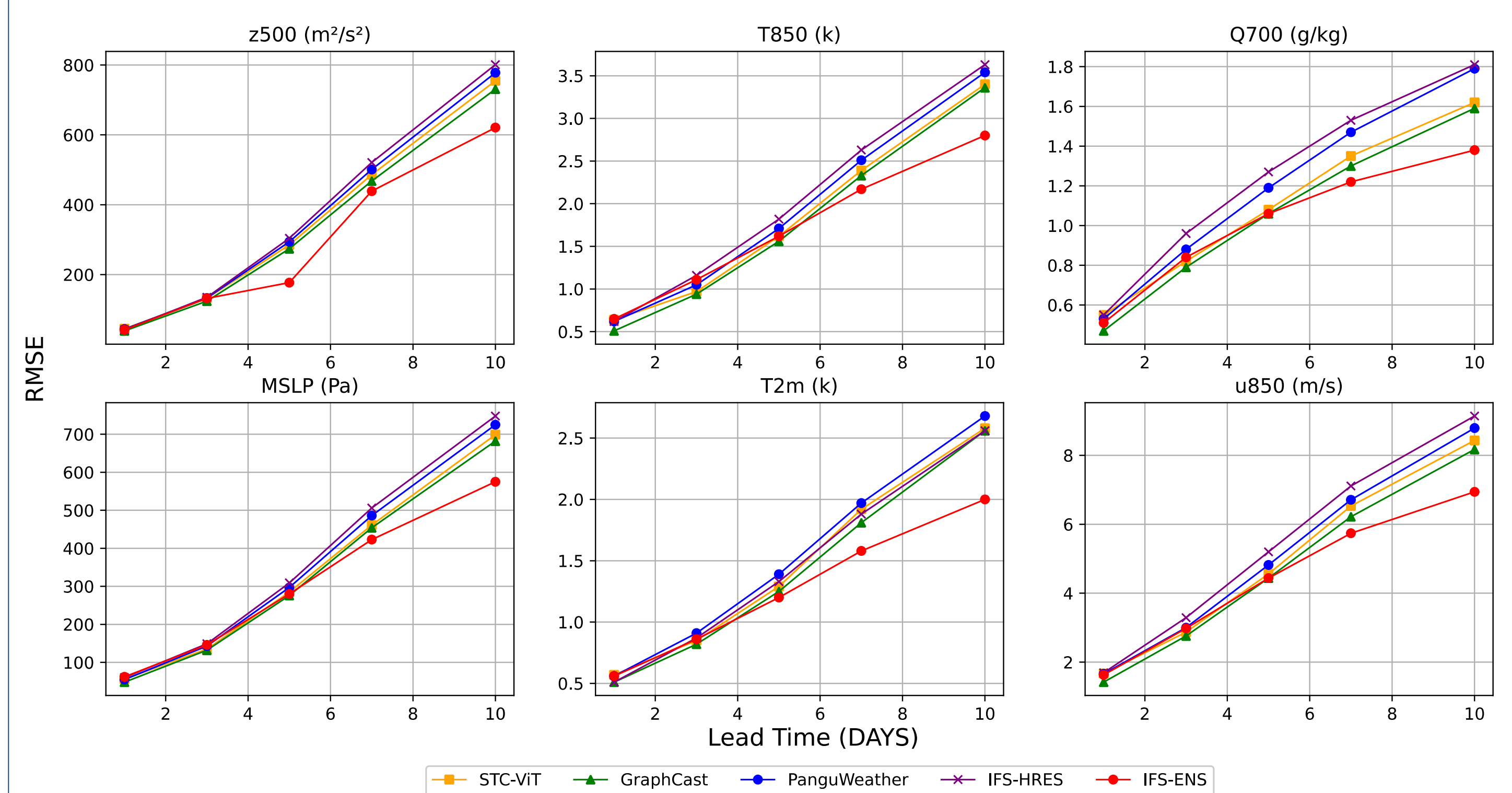


Figure 2: RMSE comparison of STC-ViT trained at 1.40625 degree with GraphCast and PanguWeather trained at 0.25 degree, IFS-ENS at 0.2 degree and IFS-HRES at 0.1 degree resolution data for lead times ranging from 1 to 10 days

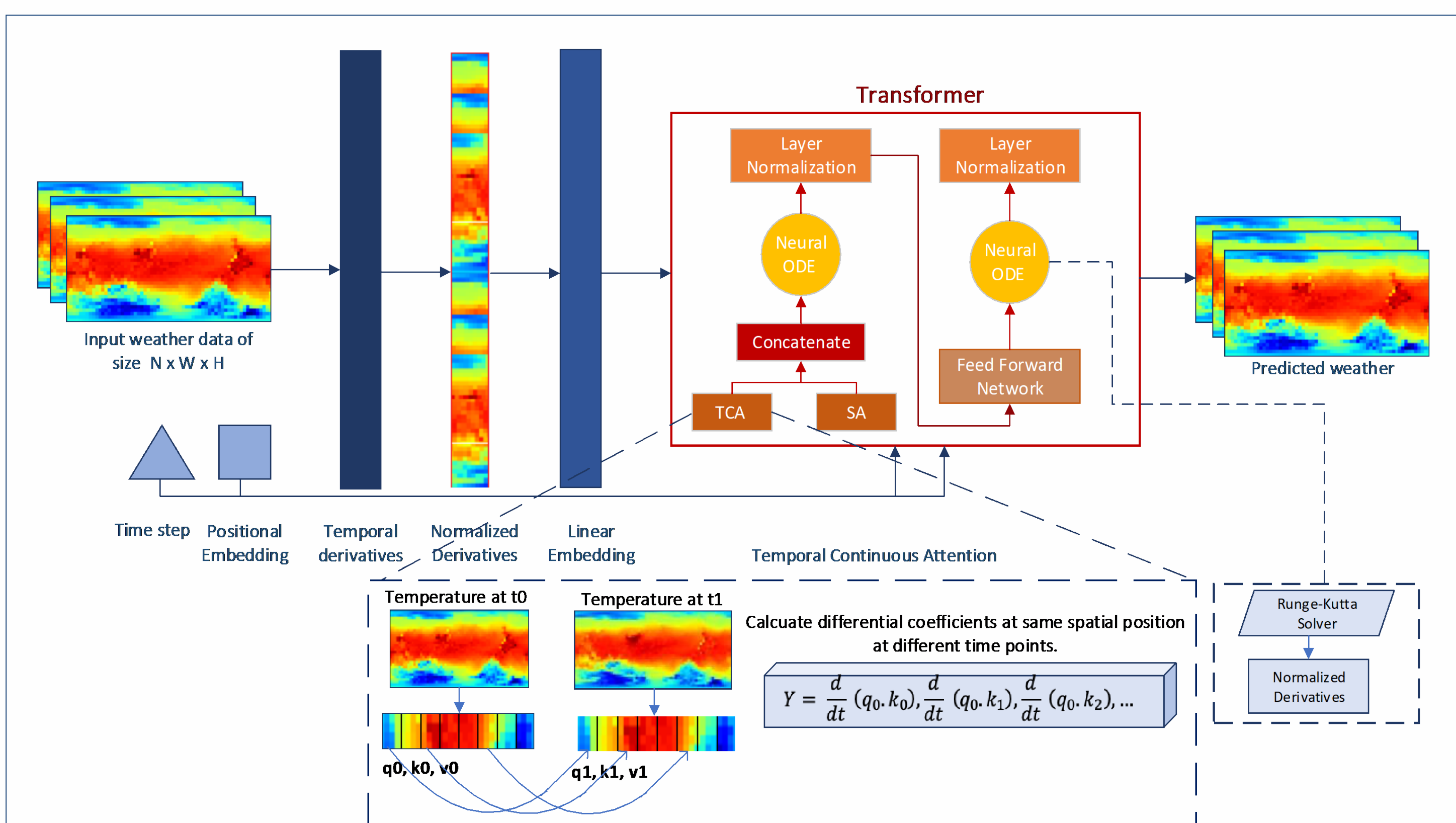


Figure 3: Prediction pipeline of STC-ViT

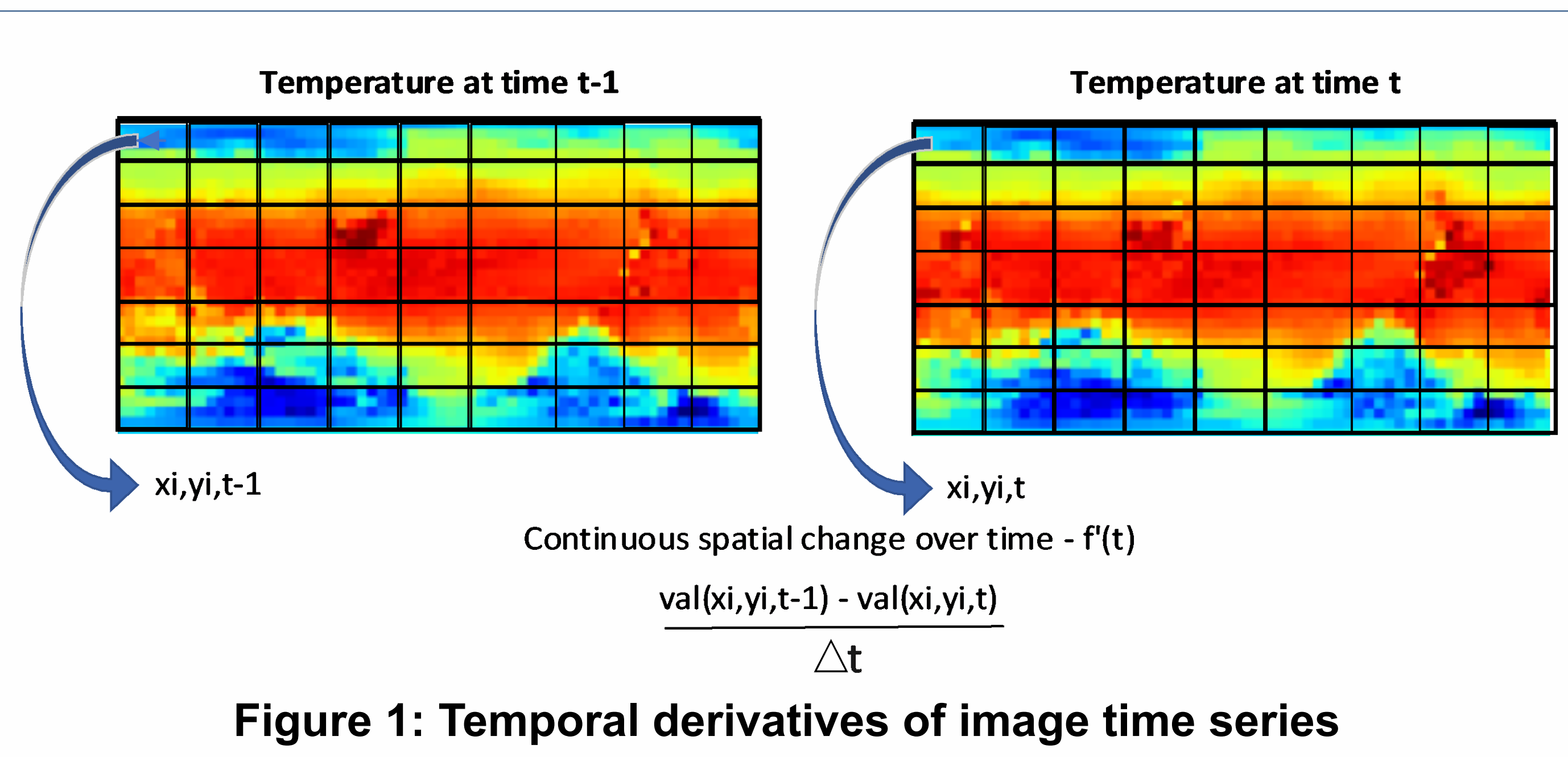


Figure 1: Temporal derivatives of image time series

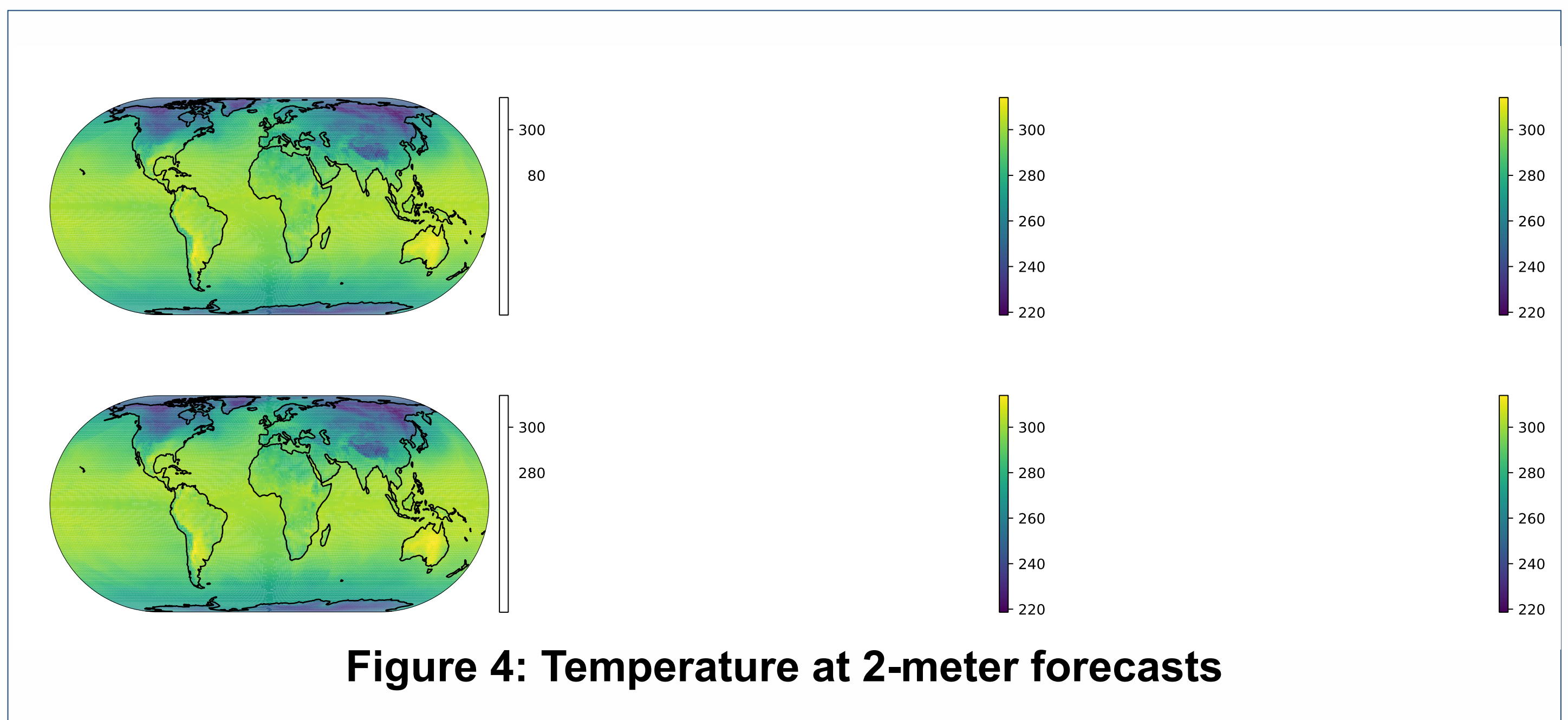


Figure 4: Temperature at 2-meter forecasts

<sup>1</sup>University of New South Wales (UNSW), Sydney, <sup>2</sup>Trillium Technologies