

QGIS plugins for Image Fusion and Image Imputation

by

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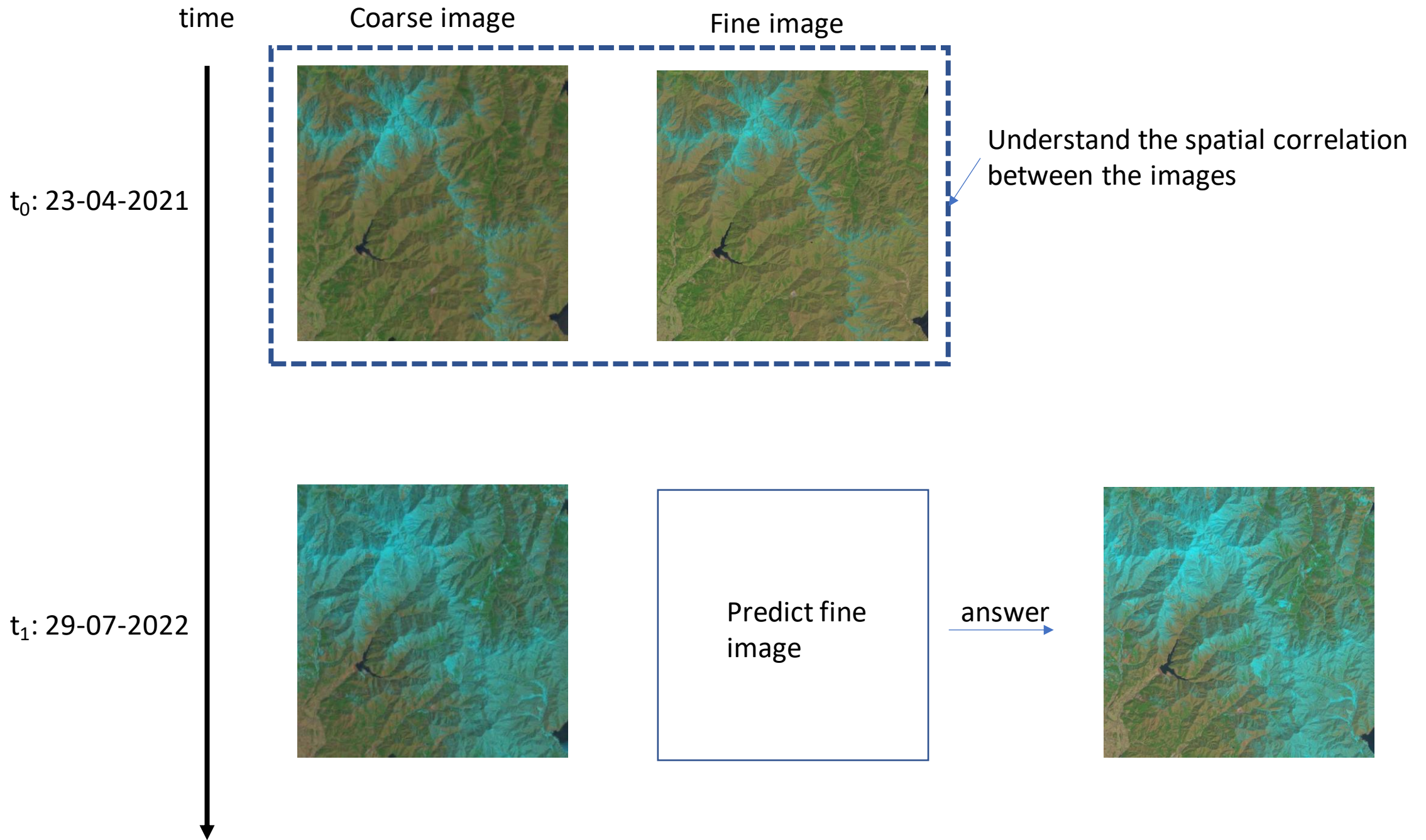
Outline

- Part-1
 - Introduction to Image Fusion
 - Demo of our QGIS image fusion plugin
 - Future work
- Part-2
 - Introduction to Imputation
 - State-of-the-art techniques
 - Demo of our QGIS image imputation plugin
 - Future work

Part 1: Image Fusion

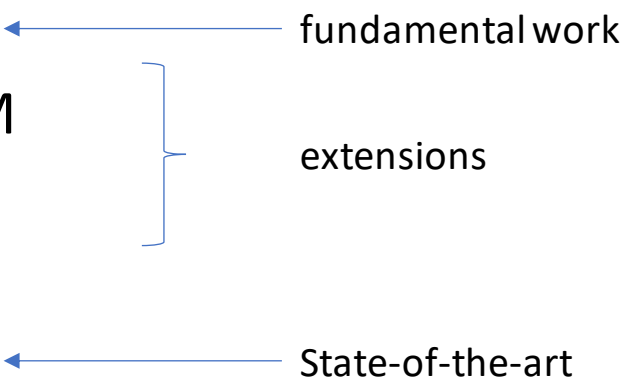
Introduction

- Image fusion:
 1. Understanding the spatial correlation between the low-resolution and high-resolution images of a particular location and
 2. Generate a higher-resolution image for a given low-resolution image taken later.




Related Works

- Existing approaches

- **STARFM** ← fundamental work
 - ESTARFM
 - SPSTFM
 - ...
 - **HISTIF** ← State-of-the-art
- 

Limitations

- 
- geo-registration errors
(preserve shape but loose spatial details)
 - tolerant to geo-registration errors
but suffers from bias

- **Improved HISTIF** (*our algorithm published in IEA-AIE 2022*)

Demo: QGIS Plugin for Image Fusion

Bandi Mountain Data

Future Work

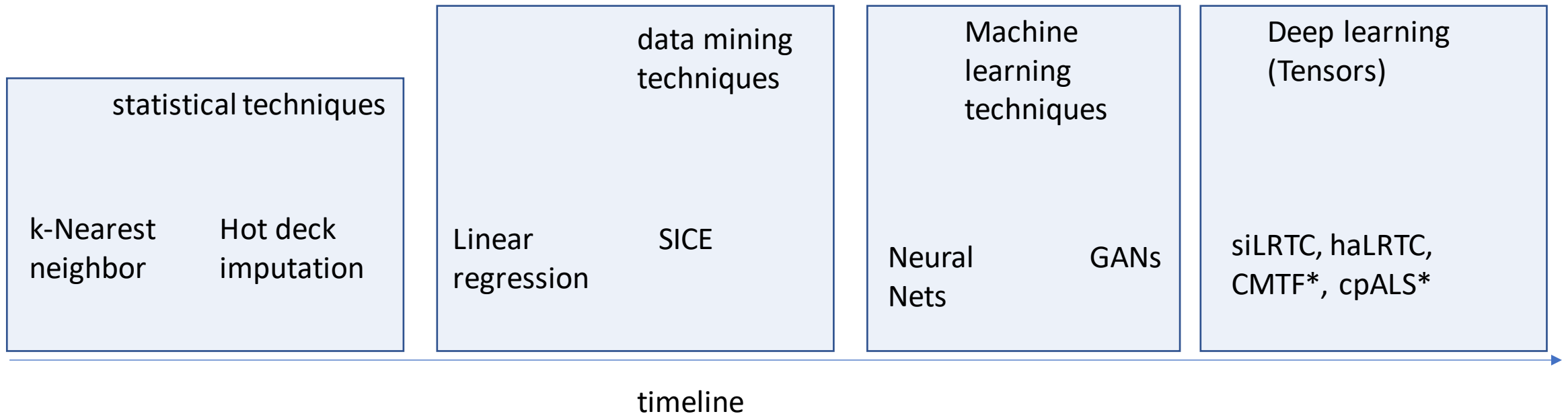
- The IHISITF algorithm is a sequential algorithm relying on a single CPU.
- We are developing parallel and distributed solutions for IHISTIF using multiple CPU cores.
- We are also exploring new Image fusion techniques.
 - Our new technique submitted to DASFAA 2022 improves over IHISITF by 10 to 20%.

Part 2: Image Imputation

Introduction

- Missing pixels is a fundamental problem in image analytics.
- **Imputation** is a technique employed to fill up the missing pixels.
- K-Nearest Neighbor is the widely adopted basic imputation technique.
- However, much better imputation techniques have been recently proposed in the field of Computer Science

State-of-the-art techniques



Demo: QGIS Plugin for Image Imputation

Future Works

- Existing Tensor-based imputation techniques do not include humans in the decision-making process.
- We are developing novel tensor-based imputation techniques by incorporating the “human-in-the-loop” concept.
 - The user will input the missing imagery data.
 - The user will also tell the technique of which data to be considered for model building.
 - Our model will learn the data specified by the user and perform imputation.
 - Preliminary results demonstrate that we can improve the performance by 10%.

The END

Thank you and questions