

Super-Resolution of SOHO/MDI Magnetograms Using SolarCNN

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1. Introduction

We introduce a deep learning method named SolarCNN for enhancing AR line-of-sight (LOS) magnetograms obtained by the Michelson Doppler Imager (MDI) onboard SOHO. The model is trained to super-resolve MDI magnetograms using references from the Helioseismic and Magnetic Imager (HMI) onboard SDO.

In this document, we describe the usage of the trained SolarCNN model to super-resolve the test data.

Since model training requires GPUs, it is omitted here.

2. Prediction Workflow

2.1 Load Model

The SolarCNN model has been saved using the `model.save()` function in TensorFlow. It includes custom loss and metric functions: `mix_loss` and `ssim_metric`, which must be specified when loading the model.

```
1 import tensorflow as tf
2 from model import build_model, mix_loss, ssim_metric
3 try:
4     model = tf.keras.models.load_model("model/model_solarcnn",
5                                         custom_objects={'mix_loss': mix_loss, 'ssim_metric': ssim_metric})
6     print("model loaded")
7 except:
8     print("model loading error")
9
```

✓ 4.4s

model loaded

2.2 Load Input Data

Test data are stored as FITS files in the directory `data/image/test/`. Each file is loaded and sorted by filename.

```

1 from astropy.io import fits
2 import os
3
4 def load_data(path):
5     data_row = [(name, fits.open(os.path.join(path, name))[0].data) for name in os.listdir(path)]
6     data_sort = sorted(data_row, key=lambda x: x[0])
7     data = [item[1] for item in data_sort]
8     name = [item[0] for item in data_sort]
9     return tf.Variable(initial_value=data, dtype=tf.float32), name
10
11 try:
12     test_input, test_name = load_data("data/image/test/")
13     print("data loaded, total sample:", len(test_name))
14 except:
15     print("data loading error")

```

✓ 0.3s

data loaded, total sample: 3

2.3 Perform Super-Resolution

Each input magnetogram is passed to the SolarCNN model for prediction. The output is a super-resolved magnetogram saved in the FITS format with filenames of the form:

enh_mdi_YYYYMMDD_HHMMSS_SolarCNN.fits

Here's the prediction and saving loop:

```

1 output_path = "data/pred/"
2 for i in range(test_input.shape[0]):
3     print("Predict sample "+str(i+1))
4     test_sample = tf.reshape(test_input[i], (-1, 256, 256, 1))
5     prediction = model.predict(test_sample/2000.)
6     pred_name = "enh_mdi_" + test_name[i][4:20] + "SolarCNN.fits"
7     fits_data = fits.PrimaryHDU(prediction[0]*2000.)
8     fits_data.writeto(os.path.join(output_path, pred_name), overwrite=True)
9     print("Finished")

```

✓ 1.9s

```

Predict sample 1
1/1 [=====] - 1s 847ms/step
Predict sample 2
1/1 [=====] - 0s 497ms/step
Predict sample 3
1/1 [=====] - 0s 492ms/step
Finished

```

3. Conclusion

We present a deep learning model, SolarCNN, for enhancing AR LOS magnetograms of SOHO/MDI. This method can be applied to existing MDI data to produce high-resolution magnetic field maps.