

```
In [ ]: import matplotlib.pyplot as plt

In [ ]: station_name = "Vienna"
test_year = 2019
at_iterations = "final"
has_no_time_context = False

In [ ]: from climatereconstructionai import evaluate
import os
import regex as re

def evaluate_wrapper(use_time_context = True, iter = "final", station = station_name, reconstruct_gaps=False):

    snapshot_folder_name = f"snapshots_{station.lower()}"
    if not use_time_context:
        snapshot_folder_name += "_backup"

    # copy "test_args_template.txt" to "test_args_temp.txt"
    # while replacing:

    replace_dir = {
        "STATIONNAME" : station.lower() if not reconstruct_gaps else f"the_gaps_{station.lower()}",
        "ITERATION" : str(iter),
        "SNAPSHOTDIR" : snapshot_folder_name,
    }

    if reconstruct_gaps:
        replace_dir["_leading_trailing"] = ""
        replace_dir["cleaned_"] = "cleaned_era5_for_"

    template_file = "test_args_template_trailing.txt" if use_time_context else "test_args_template_no-time.txt"
    temp_file = "test_args_temp.txt"

    with open(template_file, "r") as f:
        lines = f.readlines()
        with open(temp_file, "w") as f2:
            for line in lines:
                for key in replace_dir:
                    line = re.sub(key, replace_dir[key], line)
                f2.write(line)

            # save and close
            f2.close()

    evaluate(temp_file)

    os.remove(temp_file)
```

```
In [ ]: import xarray as xr
from utils import DataSet, DatasetPlotter
import numpy as np
import os

test_folder_path = "/work/bm1159/XCES/xces-work/k203179/data/test"
train_folder_path = "/work/bm1159/XCES/xces-work/k203179/data/train"
```

```
output_file_path = "outputs/output_output.nc"
era5_file = test_folder_path + f"/era5_for_{station_name.lower()}_leading_trailing.nc"
era5_full_file = f"/work/bm1159/XCES/xces-work/k203179/data_sets/era5_for_{station_name.lower()}.nc"
era5_gaps_file = f"/work/bm1159/XCES/xces-work/k203179/data_sets/era5_gaps_for_{station_name.lower()}.nc"

# get measurements values

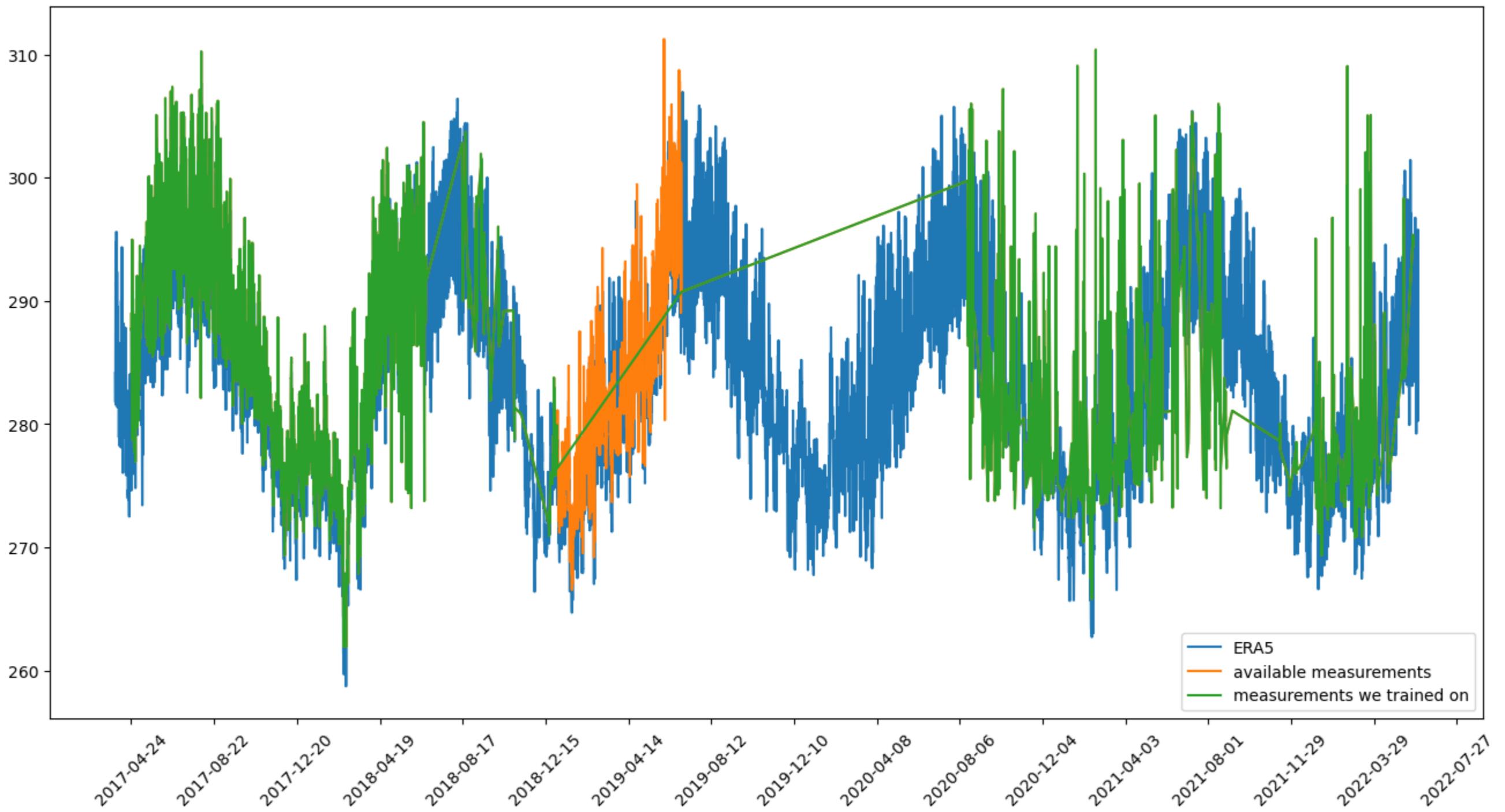
measurements_data = xr.open_dataset(test_folder_path + f"/{station_name.lower()}.nc")
era5_data = xr.open_dataset(era5_file)
measurements_data_trained = xr.open_dataset(train_folder_path + f"/expected_{station_name.lower()}.nc")

era5_full_data = xr.open_dataset(era5_full_file)

plt.plot(era5_full_data["time"], era5_full_data["tas"].mean(axis=(1,2)), label = "ERA5")
plt.plot(measurements_data["time"], measurements_data["tas"].mean(axis=(1,2)), label = "available measurements")
plt.plot(measurements_data_trained["time"], measurements_data_trained["tas"].mean(axis=(1,2)), label = "measurements we trained on")
plt.legend()

# rotate x-axis labels, and show them every 6 months
plt.xticks(rotation=45)
plt.gca().xaxis.set_major_locator(plt.MultipleLocator(120))

# figure size A4 landscape
plt.gcf().set_size_inches(16, 8)
plt.show()
```



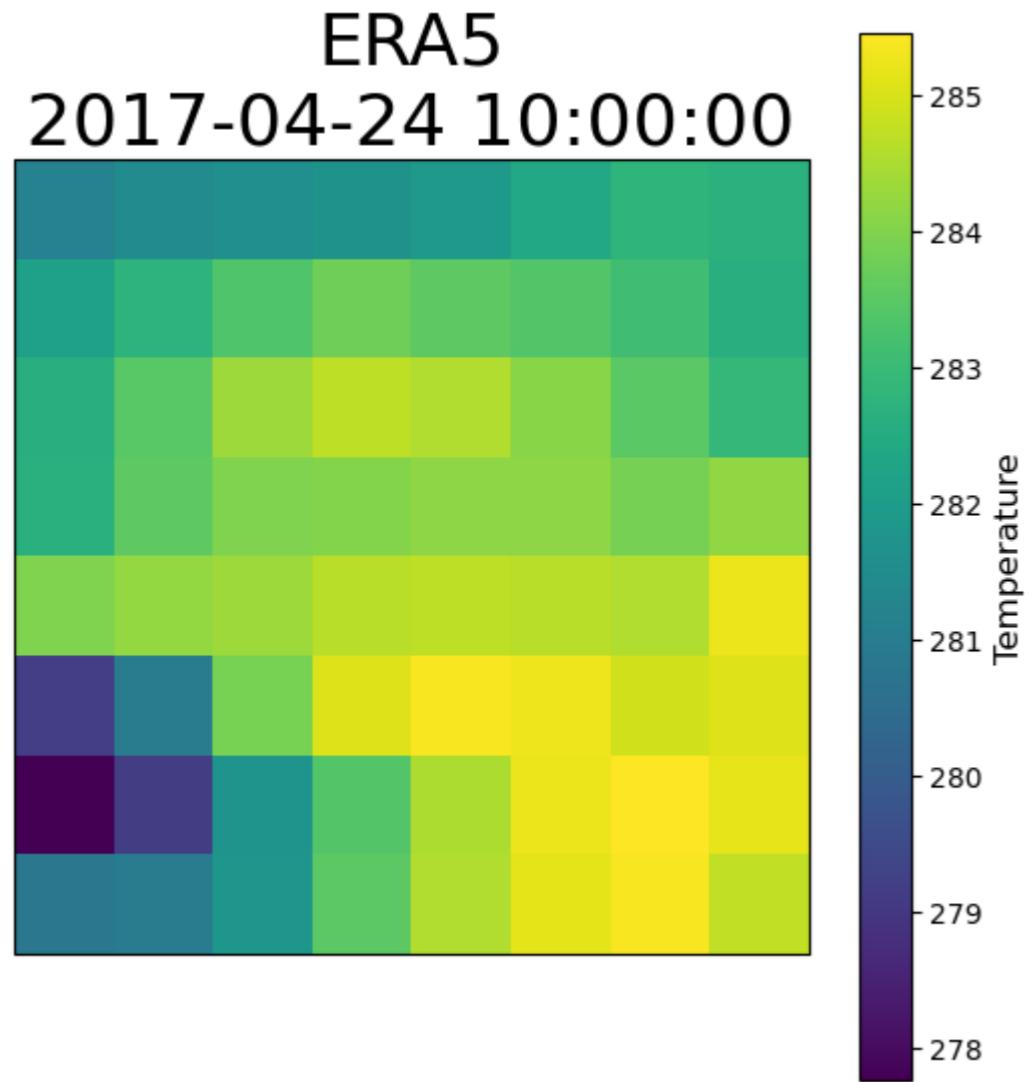
Evaluate with no timecontext

```
In [ ]: evaluate_wrapper(use_time_context = False, iter = at_iterations, station = station_name)
output_no_time_ds = DataSet(output_file_path, name = "Reconstructed")
output_no_time_data = xr.open_dataset(output_file_path)
```

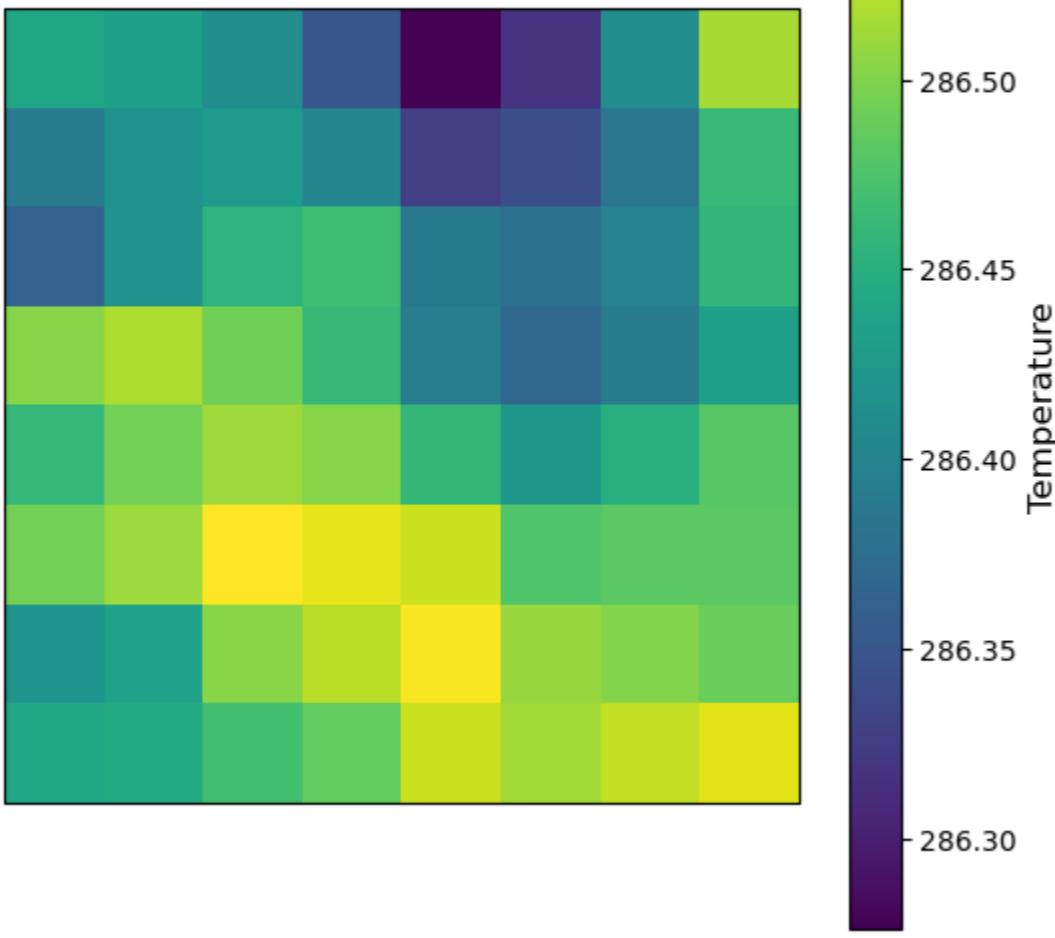
```
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/climatereconstructionai/utils/normalizer.py:10: RuntimeWarning: Mean of empty slice
    img_mean.append(np.nanmean(np.array(img_data[i])))
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/numpy/lib/nanfunctions.py:1879: RuntimeWarning: Degrees of freedom <= 0 for slice.
    var = nanvar(a, axis=axis, dtype=dtype, out=out, ddof=ddof,
100%|██████████| 1/1 [00:07<00:00,  7.39s/it]
```

```
In [ ]: plotter1 = DatasetPlotter(DataSet(era5_file, name = "ERA5"))
plotter1.plot()
plotter2 = DatasetPlotter(output_no_time_ds)
```

```
plotter2.time_index_list = plotter1.time_index_list  
plotter2.plot()
```



Reconstructed 2017-04-24 10:00:00



Evaluate with time context

```
In [ ]: if has_no_time_context:  
    output_with_time_data = output_no_time_data  
else:  
    evaluate_wrapper(use_time_context = True, iter = at_iterations, station = station_name)  
    output_with_time_ds = DataSet(output_file_path, name="Reconstructed with time context")  
    output_with_time_data = xr.open_dataset(output_file_path)  
  
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/climatereconstructionai/utils/normalizer.py:10: RuntimeWarning: Mean of empty slice  
    img_mean.append(np.nanmean(np.array(img_data[i])))  
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/numpy/lib/nanfunctions.py:1879: RuntimeWarning: Degrees of freedom <= 0 for slice.  
    var = nanvar(a, axis=axis, dtype=dtype, out=out, ddof=ddof,  
100%|██████████| 1/1 [00:11<00:00, 11.39s/it]
```

```
In [ ]: # get coordinates from measurements nc file  
import numpy as np  
  
station_lon, station_lat = measurements_data.lon.values[0], measurements_data.lat.values[0]  
print(f"station is at {station_lon}, {station_lat}")  
  
# get nearest coordinates in era5  
def get_left_right_nearest_elem_in_sorted_array(array, value):  
    length = len(array)  
    left = len(list(filter(lambda x: x <= value, array))) - 1
```

```

right = length - len(list(filter(lambda x: x >= value, array)))
nearest = min(left, right, key=lambda x: abs(array[x] - value))
return left, right, nearest

test_array = [1, 2, 3, 4, 5, 6, 7, 8]
test_search = 5.51

print(f"searching for {test_search} in {test_array}")
left_idx, right_idx, nearest_idx = get_left_right_nearest_elem_in_sorted_array(test_array, test_search)
print(f"idx left to {test_search} is {left_idx}, idx right to {test_search} is {right_idx}, nearest idx is {nearest_idx}")
print(f"mid crop: {test_array[left_idx:right_idx+1]}")

```

station is at 16.3609, 48.2303
searching for 5.51 in [1, 2, 3, 4, 5, 6, 7, 8]
idx left to 5.51 is 4, idx right to 5.51 is 5, nearest idx is 5
mid crop: [5, 6]

```

In [ ]: def era_vs_reconstructed_comparision_to_df(era5_data, reconstructed_with_time_data, reconstructed_no_time_data, measurements_data):

    lon_left_idx, lon_right_idx, lon_nearest_idx = get_left_right_nearest_elem_in_sorted_array(era5_data.lon.values, station_lon % 360)
    lat_left_idx, lat_right_idx, lat_nearest_idx = get_left_right_nearest_elem_in_sorted_array(era5_data.lat.values, station_lat)

    era5_nearest_values = era5_data.variables["tas"][:, lon_nearest_idx, lat_nearest_idx]

    reconstructed_with_time_data_values = reconstructed_with_time_data.variables["tas"].stack(grid=['lat', 'lon']).values
    reconstructed_no_time_data_values = reconstructed_no_time_data.variables["tas"].stack(grid=['lat', 'lon']).values

    measurements_data_values = measurements_data.variables["tas"][...].mean(axis=(1,2))

    # timeaxis
    time = era5_data.variables["time"][:]

    import pandas as pd

    # create dataframe with all values
    df = pd.DataFrame()

    df["time"] = time

    # index should be time
    df.set_index("time", inplace=True)

    df["era5_nearest"] = era5_nearest_values
    df["reconstructed_median"] = [np.median(x) for x in reconstructed_no_time_data_values]
    df["reconstructed_with_time_context_median"] = [np.median(x) for x in reconstructed_with_time_data_values]
    df["measurements"] = measurements_data_values

    return df

```

Generate Dataframe

- makes resampling easier

```

In [ ]: hourly_df = era_vs_reconstructed_comparision_to_df(era5_data, output_with_time_data, output_no_time_data, measurements_data)

# print a section of the df

```

```

start_date = test_year.__str__() + "-01-01"
end_date = test_year.__str__() + "-12-31"

hourly_df = hourly_df[start_date:end_date]

```

Implement plotting method of dataframe

```

In [ ]: def plot_n_steps_of_df(df, as_delta, n=None, title=None, plot_trailing_time=False):

    plot_trailing_time = plot_trailing_time and not has_no_time_context

    time = df.index.values
    if n is None:
        n = len(df)

    # random slice of n consecutive datapoints
    import random
    slice_start = random.randint(0, len(time) - n)
    time_slice = slice(slice_start, slice_start + n)

    time = time[time_slice]

    # era5_mid_values = df["era5_mid"].values -273.15
    era5_nearest_values = df["era5_nearest"].values - 273.15
    reconstructed_median_values = df["reconstructed_median"].values - 273.15
    reconstructed_median_with_time_context_values = df["reconstructed_with_time_context_median"].values - 273.15

    measurements_values = df["measurements"].values - 273.15

    rmse_reconstructed = np.sqrt(np.sum((reconstructed_median_values[time_slice] - measurements_values[time_slice])**2) / len(time))
    rmse_reconstructed_with_time_context = np.sqrt(np.sum((reconstructed_median_with_time_context_values[time_slice] - measurements_values[time_slice])**2) / len(time))
    rmse_era5_nearest = np.sqrt(np.sum((era5_nearest_values[time_slice] - measurements_values[time_slice])**2) / len(time))

    correlation_reconstructed = np.corrcoef(reconstructed_median_values[time_slice], measurements_values[time_slice])[0,1]
    correlation_reconstructed_with_time_context = np.corrcoef(reconstructed_median_with_time_context_values[time_slice], measurements_values[time_slice])[0,1]
    correlation_era5_nearest = np.corrcoef(era5_nearest_values[time_slice], measurements_values[time_slice])[0,1]

    if as_delta:
        # era5_mid_values = era5_mid_values - measurements_values
        era5_nearest_values = era5_nearest_values - measurements_values
        reconstructed_median_values = reconstructed_median_values - measurements_values
        reconstructed_median_with_time_context_values = reconstructed_median_with_time_context_values - measurements_values
        measurements_values = measurements_values - measurements_values

        # y-axis title, temperature difference
        plt.ylabel("Delta calculated by subtracting measurement data [C°]")
    else:
        plt.ylabel("Temperature at surface [C°]")

    plt.plot(time, era5_nearest_values[time_slice], label="ERA5 nearest point", color="red")
    # plt.plot(time, era5_mid_values[time_slice], label="ERA5 nearest 4 points")

    plt.plot(time, reconstructed_median_values[time_slice], label="Reconstructed for validation", color="blue")
    if plot_trailing_time:

```

```

plt.plot(time, reconstructed_median_with_time_context_values[time_slice], label="Reconstructed for val. using also ERA5 step before & after", color="green")

plt.plot(time, measurements_values[time_slice], label="Measurements", color="black")

# x-axis labels 90 degrees
plt.xticks(rotation=45)

# title
if title is not None:
    plt.title(title)

# font size of legend
plt.rcParams.update({'font.size': 10})

# font size of axis labels
plt.rcParams.update({'axes.labelsize': 12})

# font size of title
plt.rcParams.update({'axes.titlesize': 26})

plt.legend()
# position legend below chart to the right
plt.legend(bbox_to_anchor=(1, 1.15), loc='upper right', borderaxespad=0.)

# text below diagram with RMSE and Correlation in fontsize 10
plt.text(0.1, 0.95, f"RMSE reconstructed: {rmse_reconstructed:.2f} C°\n" +
         (f"RMSE reconstr. (time context): {rmse_reconstructed_with_time_context:.2f} C°\n" if plot_trailing_time else "") +
         f"RMSE ERA5 nearest point: {rmse_era5_nearest:.2f} C°",
         fontsize=10, transform=plt.gcf().transFigure)

plt.text(0.3, 0.95, f"Correlation reconstructed: {correlation_reconstructed:.3f}\n" +
         (f"Correlation reconstr. (time context): {correlation_reconstructed_with_time_context:.3f}\n" if plot_trailing_time else "") +
         f"Correlation ERA5 nearest point: {correlation_era5_nearest:.3f}",
         fontsize = 10, transform=plt.gcf().transFigure)

# figure size A4 landscape
plt.gcf().set_size_inches(16, 8)

plt.show()

```

Hourly Analysis

...over the full year

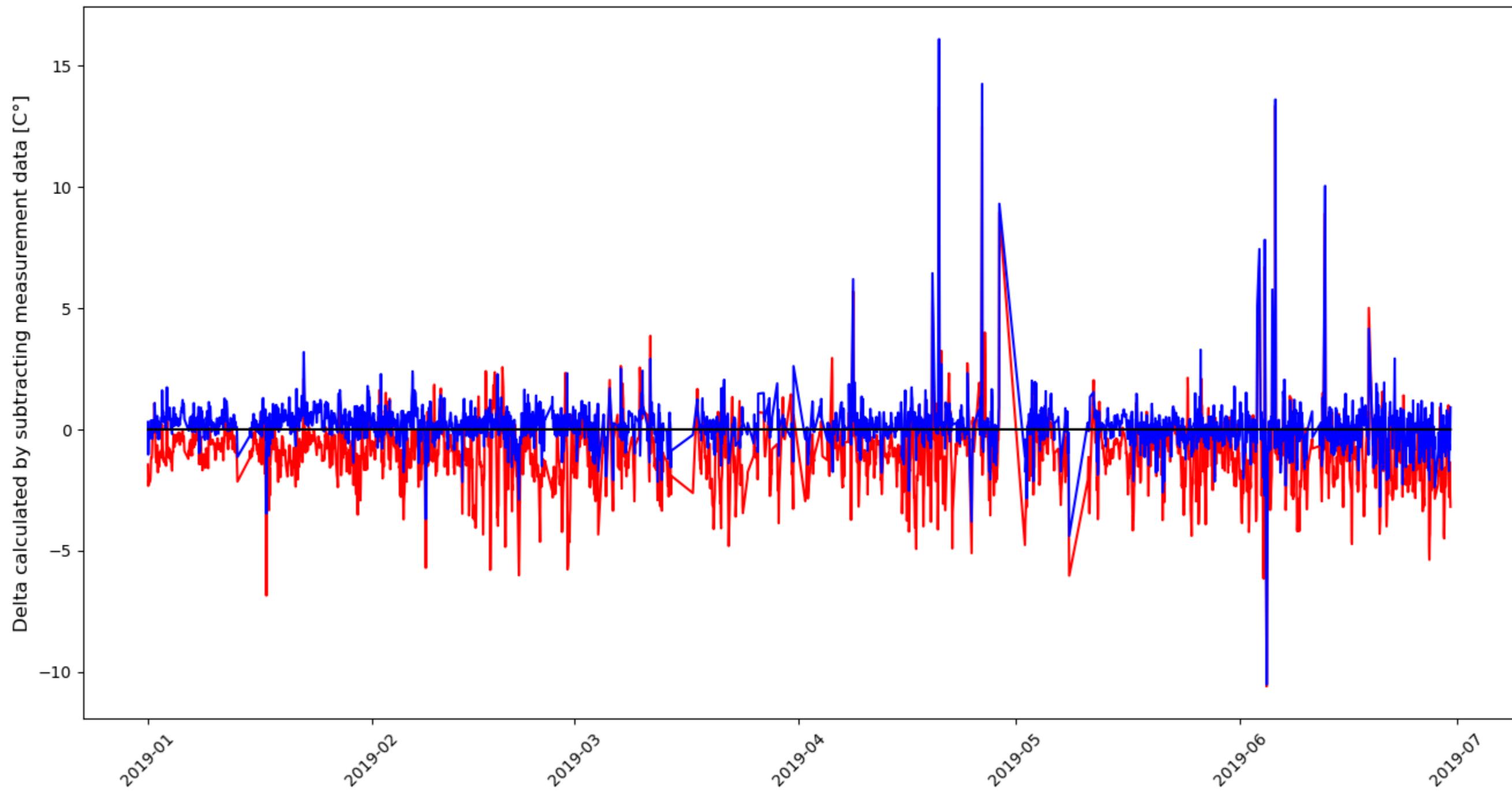
```
In [ ]: title = f"{station_name} - {test_year}"
plot_n_steps_of_df(hourly_df, as_delta=True, title=title)
plot_n_steps_of_df(hourly_df, as_delta=False, title=title)
```

RMSE reconstructed: 0.91 C°
RMSE ERA5 nearest point: 1.65 C°

Correlation reconstructed: 0.994
Correlation ERA5 nearest point: 0.988

- ERA5 nearest point
- Reconstructed for validation
- Measurements

Vienna - 2019

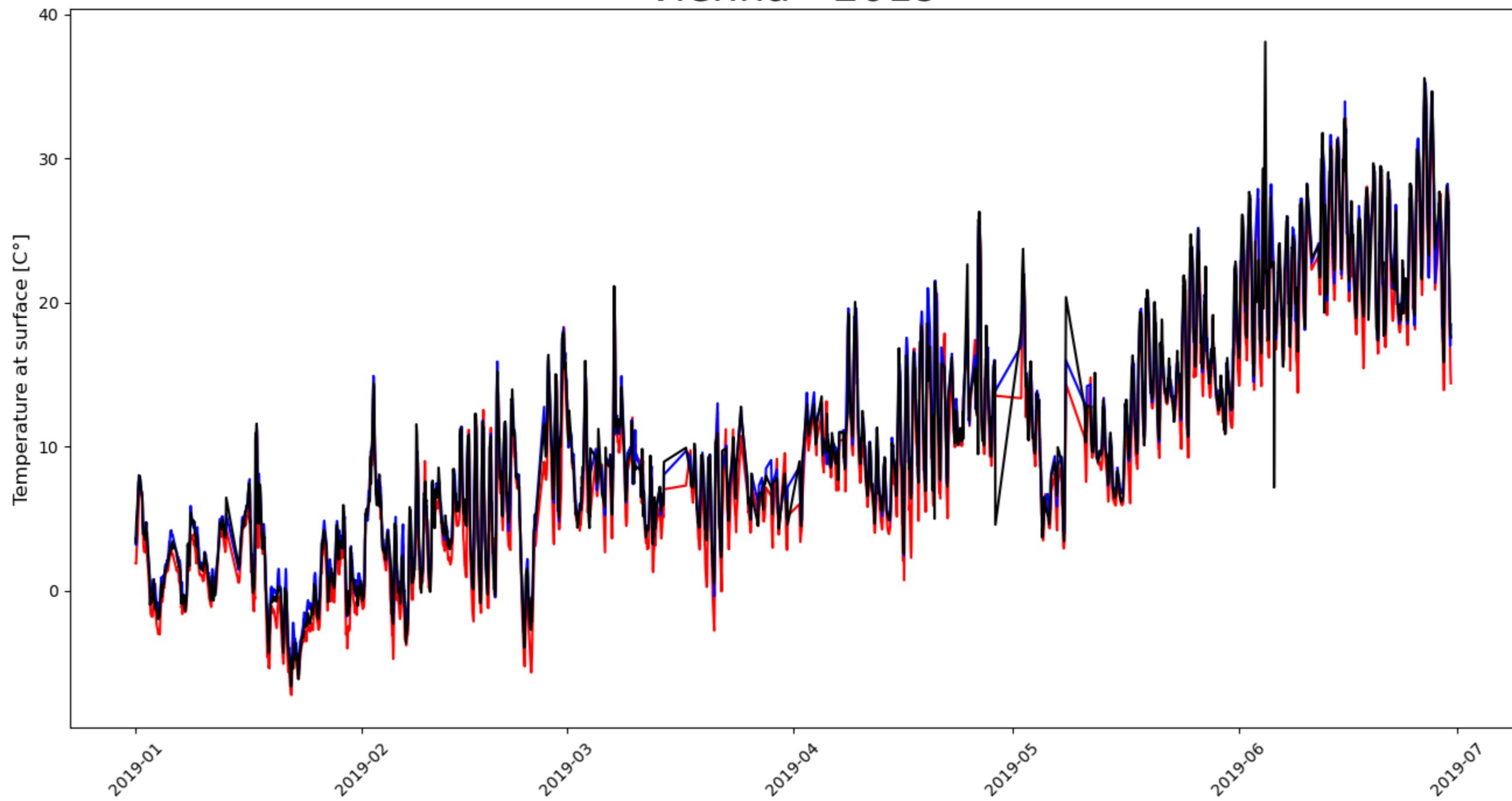


RMSE reconstructed: 0.91 C°
RMSE ERA5 nearest point: 1.65 C°

Correlation reconstructed: 0.994
Correlation ERA5 nearest point: 0.988

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna - 2019



...over random weeks

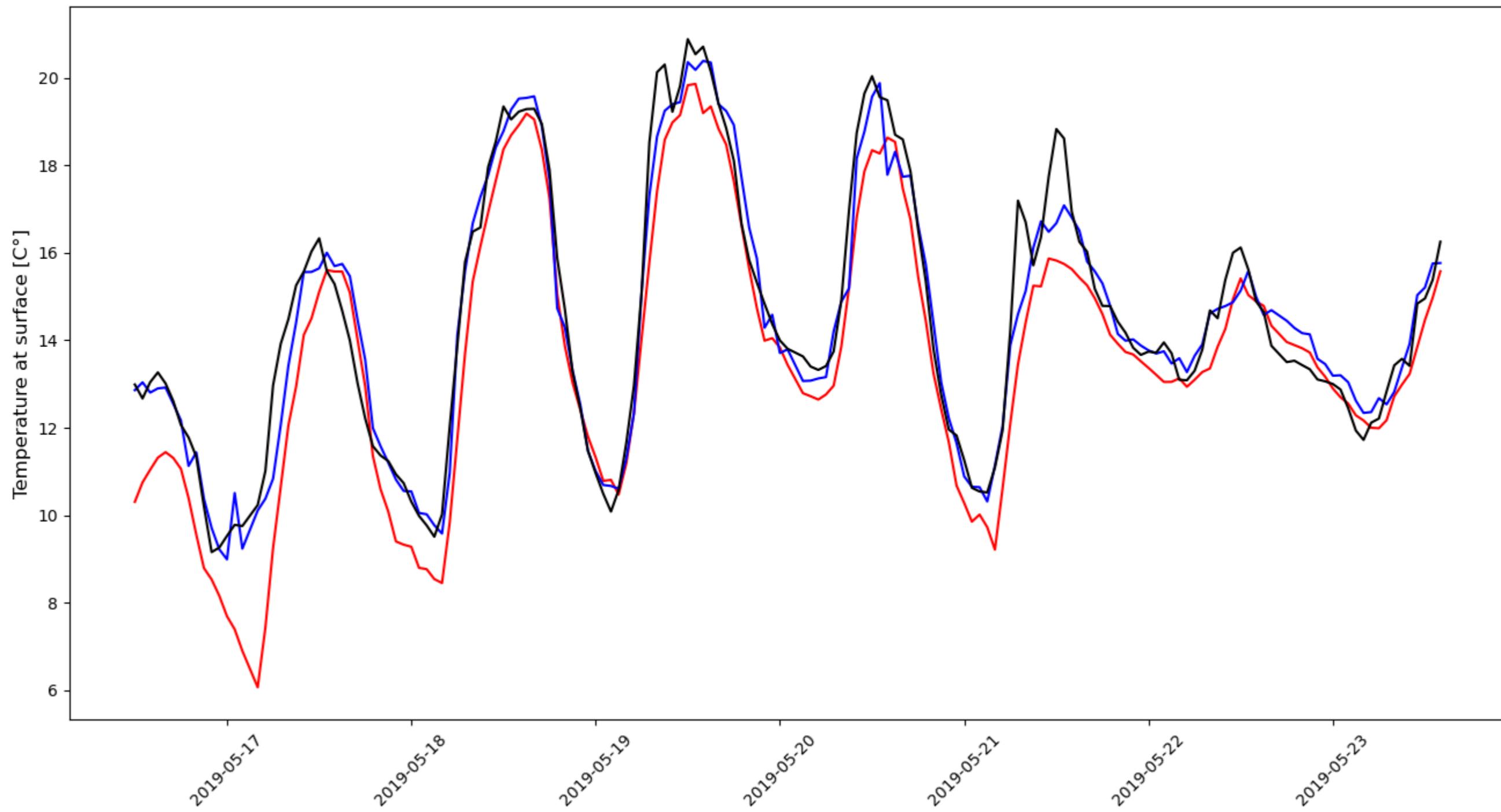
```
In [ ]: n = 168
if n == 168:
    title = f"{station_name} {test_year} - over a random week"
else:
    title = f"{station_name} {test_year} - {n} random consecutive hourly steps"
for _ in range(3):
    plot_n_steps_of_df(hourly_df, as_delta=False, n=n, title=title)
```

RMSE reconstructed: 0.66 C°
RMSE ERA5 nearest point: 1.29 C°

Correlation reconstructed: 0.975
Correlation ERA5 nearest point: 0.949

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - over a random week

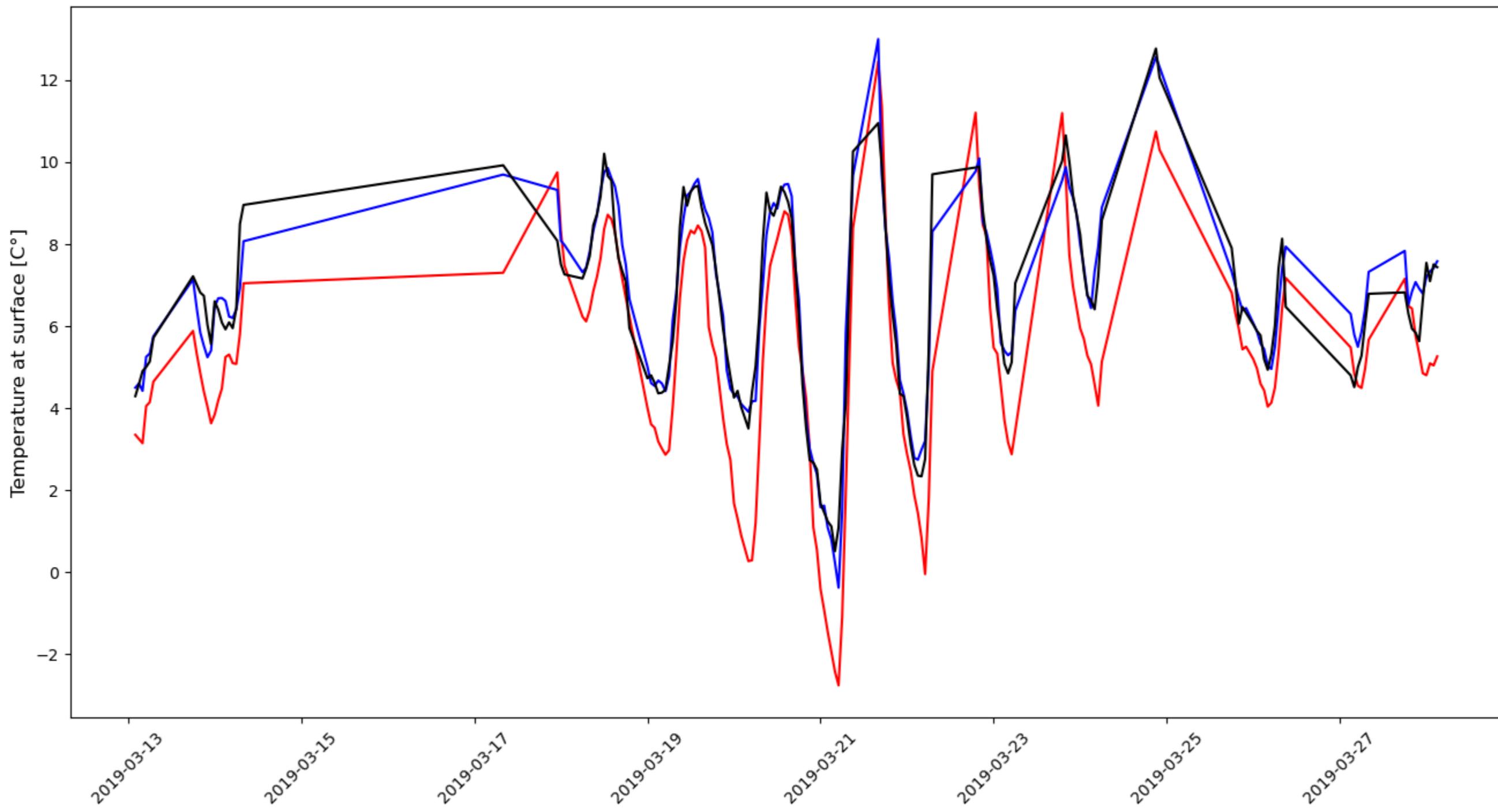


RMSE reconstructed: 0.58 C°
RMSE ERA5 nearest point: 1.78 C°

Correlation reconstructed: 0.969
Correlation ERA5 nearest point: 0.910

- ERA5 nearest point
- Reconstructed for validation
- Measurements

Vienna 2019 - over a random week

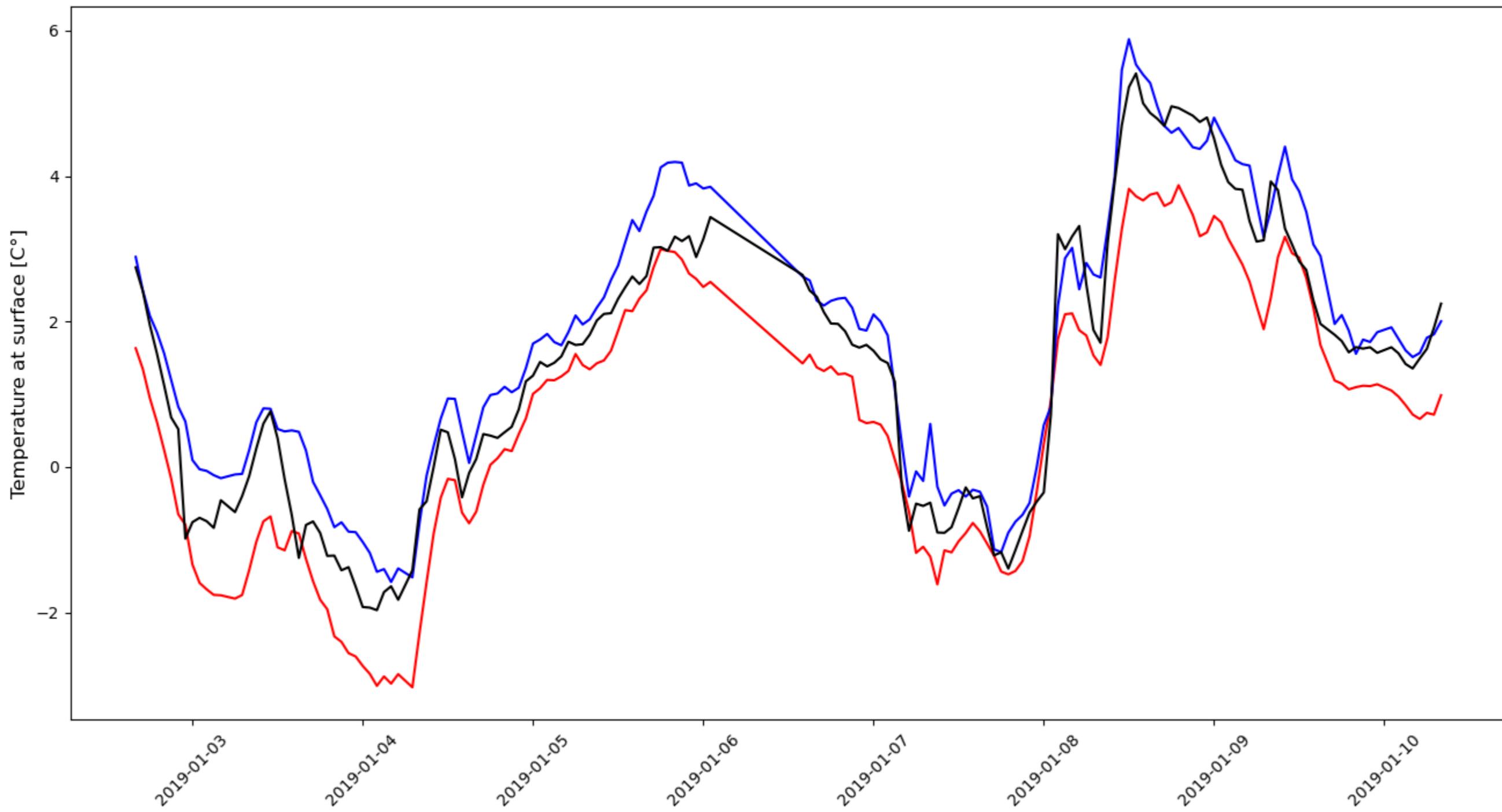


RMSE reconstructed: 0.55 C°
RMSE ERA5 nearest point: 0.85 C°

Correlation reconstructed: 0.978
Correlation ERA5 nearest point: 0.969

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - over a random week



Resample Data from hourly to daily or monthly

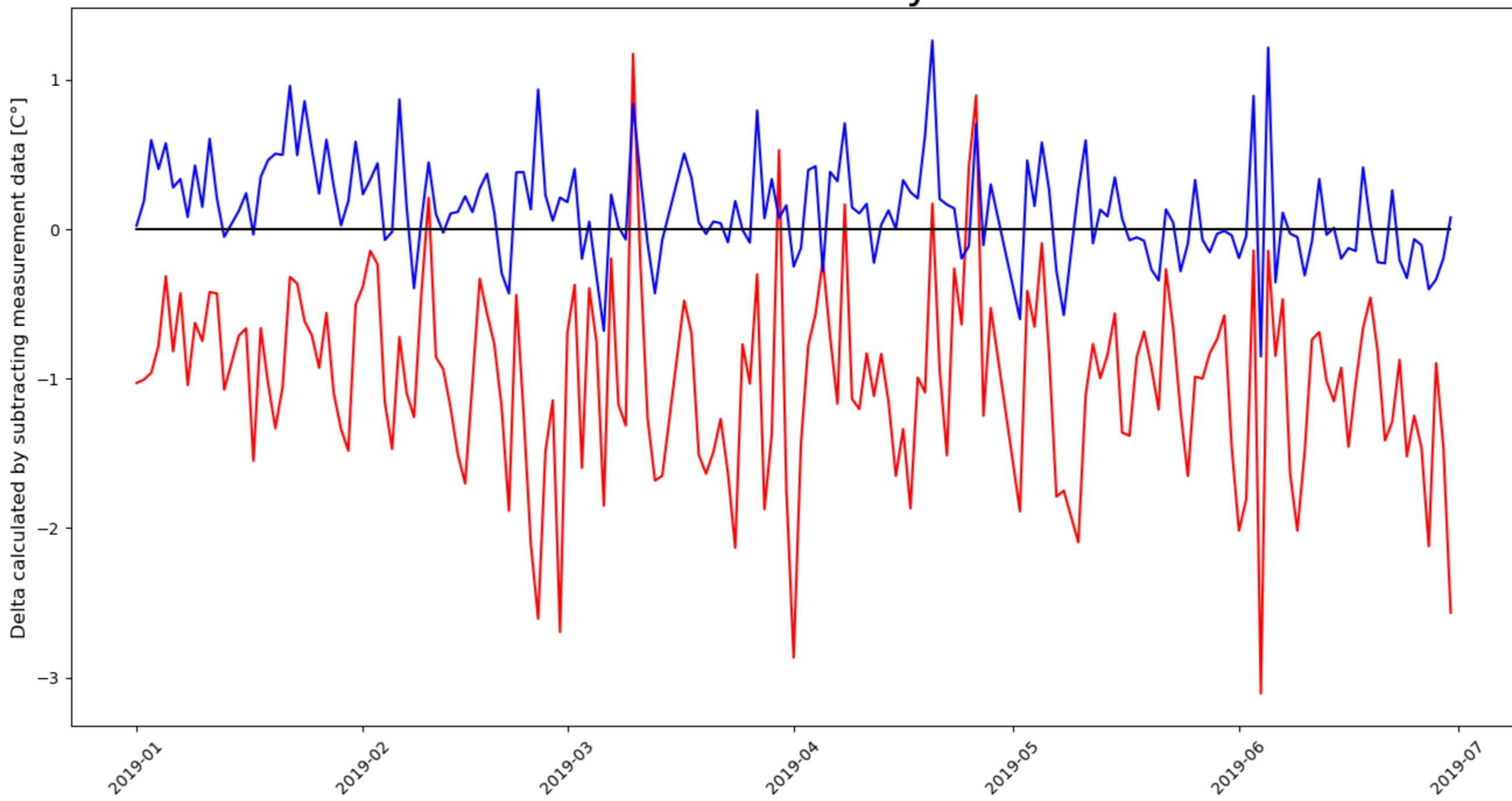
```
In [ ]: # resample to daily mean
daily_df = hourly_df.resample("D").mean()
# drop nans
daily_df = daily_df.dropna()
title = f"{station_name} {test_year} - daily mean"
plot_n_steps_of_df(daily_df, as_delta=True, title=title)
plot_n_steps_of_df(daily_df, as_delta=False, title=title)
```

RMSE reconstructed: 0.37 C°
RMSE ERA5 nearest point: 1.20 C°

Correlation reconstructed: 0.999
Correlation ERA5 nearest point: 0.997

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - daily mean

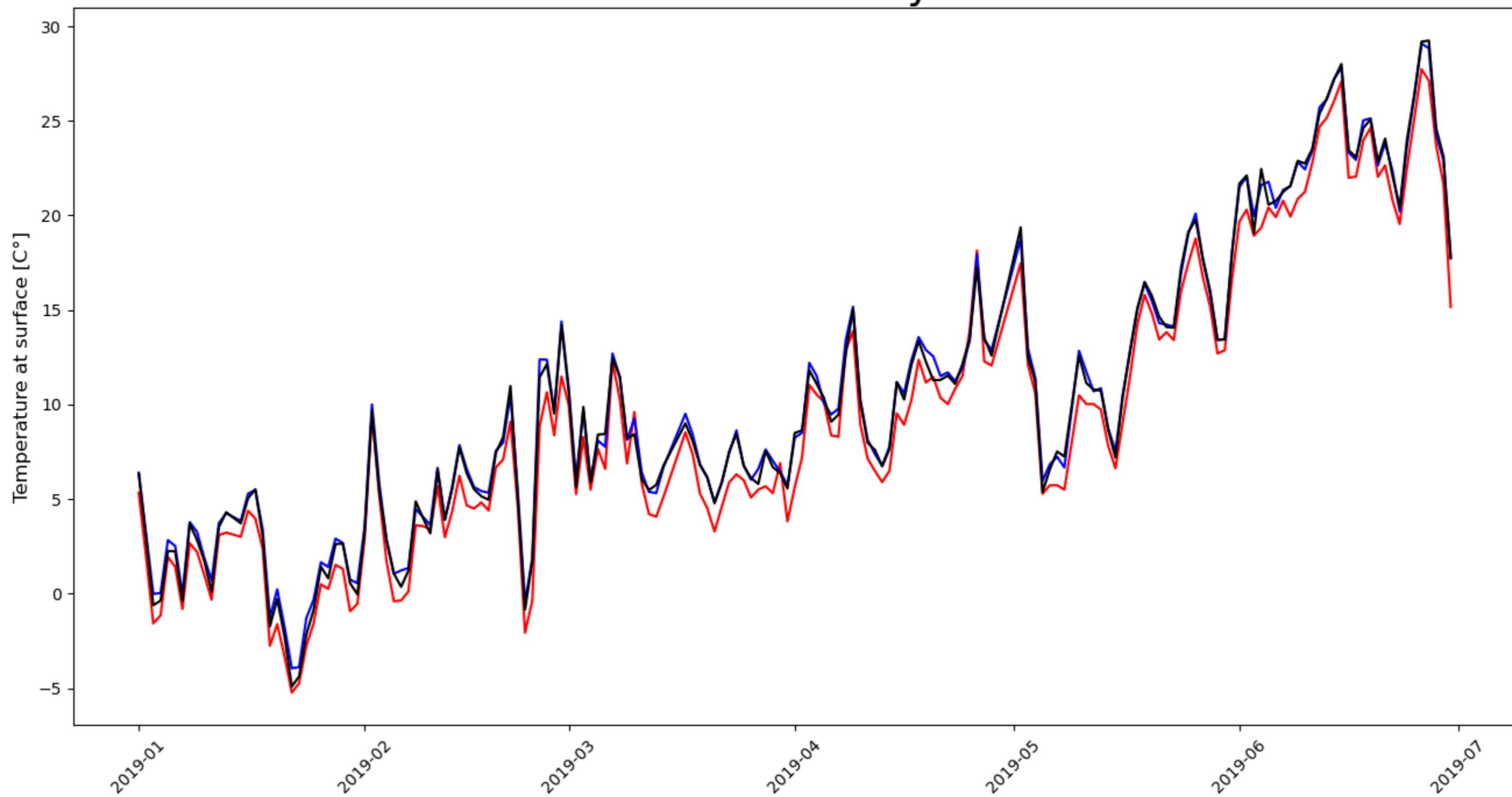


RMSE reconstructed: 0.37 C°
RMSE ERA5 nearest point: 1.20 C°

Correlation reconstructed: 0.999
Correlation ERA5 nearest point: 0.997

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - daily mean



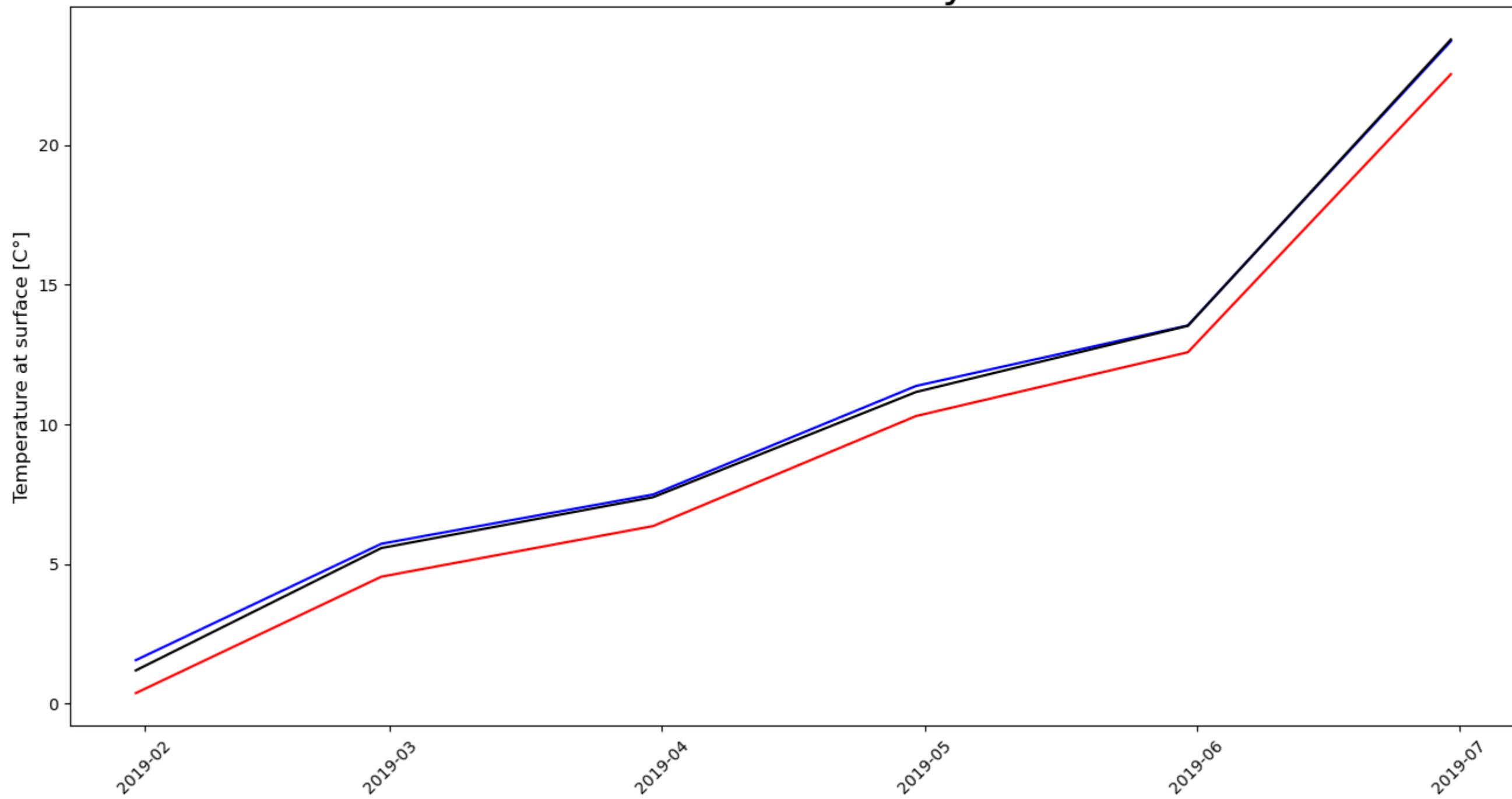
```
In [ ]: # resample rows to monthly mean
df = hourly_df.resample('M').mean()
title = f'{station_name} {test_year} - monthly mean'
plot_n_steps_of_df(df, as_delta = False, title=title)
```

RMSE reconstructed: 0.19 C°
RMSE ERA5 nearest point: 1.00 C°

Correlation reconstructed: 1.000
Correlation ERA5 nearest point: 1.000

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - monthly mean



Average Course of a day

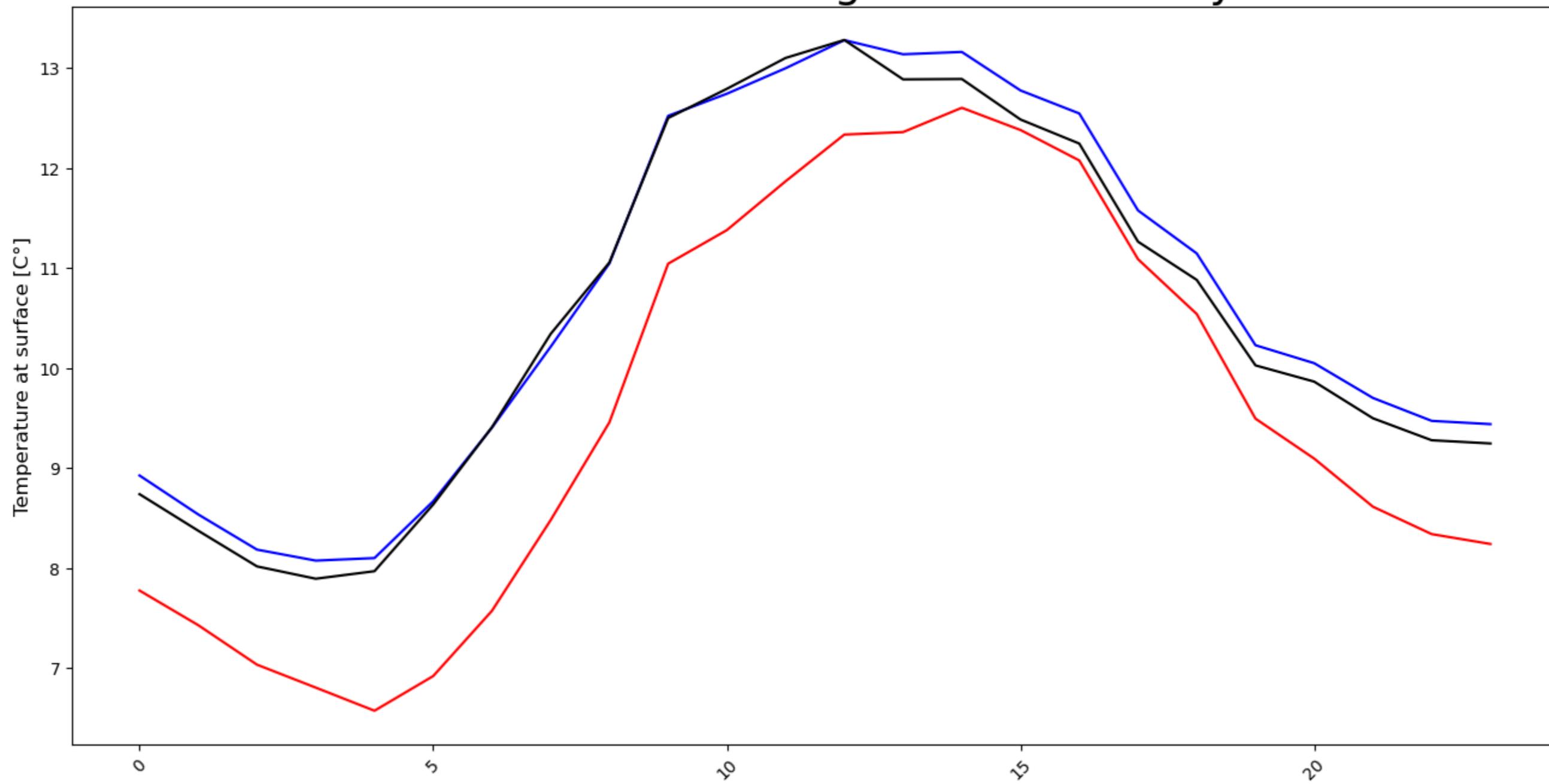
```
In [ ]: # calculate the mean of each 24 hours over the whole year  
  
day_course_df = hourly_df.groupby(hourly_df.index.hour).mean()  
title = f'{station_name} {test_year} - average course of a day'  
plot_n_steps_of_df(day_course_df, as_delta = False, title=title)
```

RMSE reconstructed: 0.19 C°
RMSE ERA5 nearest point: 1.10 C°

Correlation reconstructed: 0.997
Correlation ERA5 nearest point: 0.970

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - average course of a day



Average Course of a month

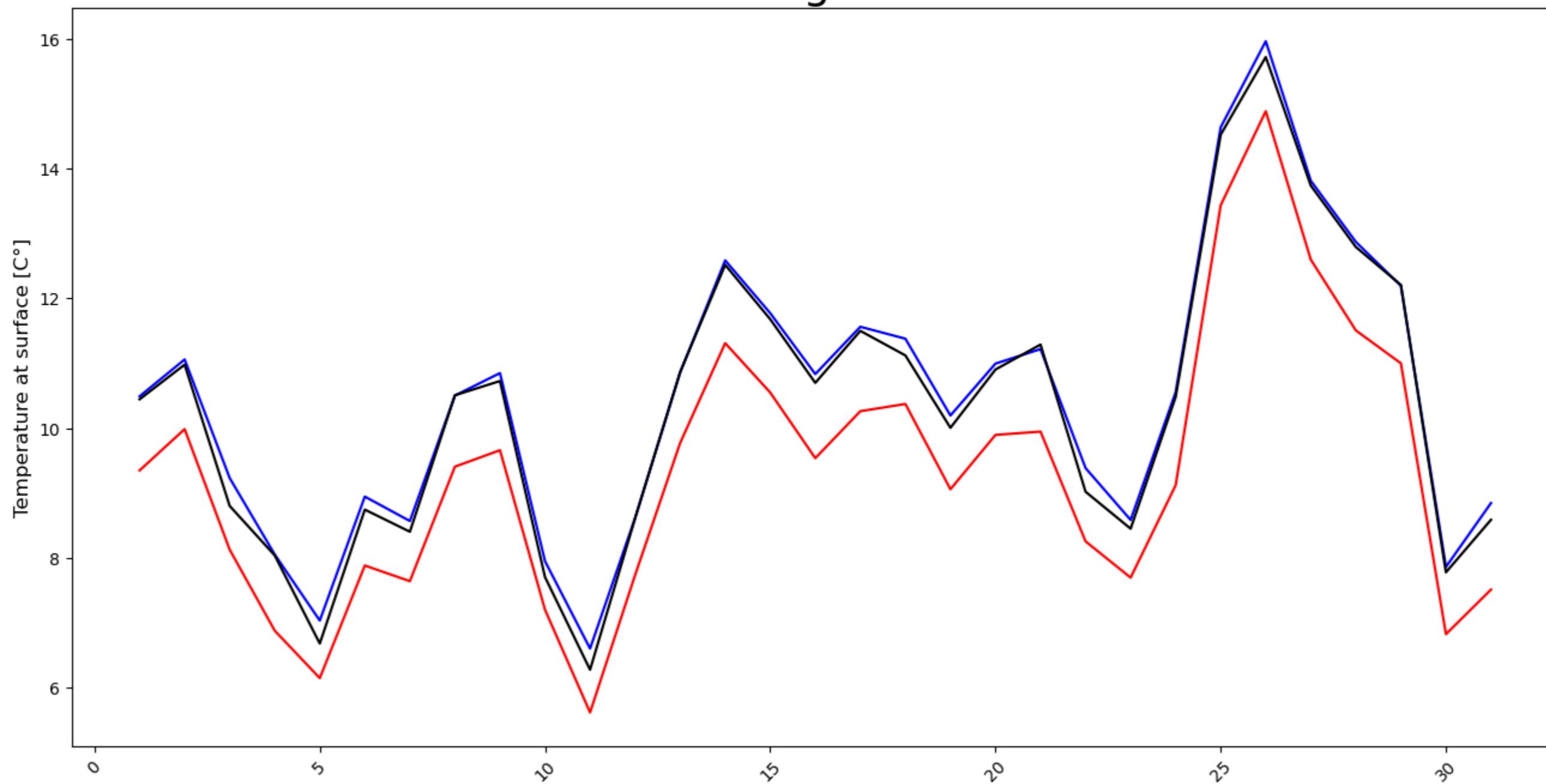
```
In [ ]: # calculate the mean of each day of a month over the whole year
month_course_df = hourly_df.groupby(hourly_df.index.day).mean()
title = f'{station_name} {test_year} - average course of a month'
plot_n_steps_of_df(month_course_df, as_delta = False, title=title)
```

RMSE reconstructed: 0.18 C°
RMSE ERA5 nearest point: 1.01 C°

Correlation reconstructed: 0.999
Correlation ERA5 nearest point: 0.996

ERA5 nearest point
Reconstructed for validation
Measurements

Vienna 2019 - average course of a month



Evaluate for the actually missing time steps

```
In [ ]: evaluate_wrapper(use_time_context = False, iter = at_iterations, station = station_name, reconstruct_gaps=True)
output_filled_gaps_ds = DataSet(output_file_path, name="Reconstructed gaps")
output_actual_gaps_data = xr.open_dataset(output_file_path)
```

```
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/climatereconstructionai/utils/normalizer.py:10: RuntimeWarning: Mean of empty slice
    img_mean.append(np.nanmean(np.array(img_data[i])))
/home/k/k203179/.conda/envs/crai/lib/python3.10/site-packages/numpy/lib/nanfunctions.py:1879: RuntimeWarning: Degrees of freedom <= 0 for slice.
    var = nanvar(a, axis=axis, dtype=dtype, out=out, ddof=ddof,
100%|██████████| 1/1 [00:15<00:00, 15.09s/it]
```

```
In [ ]: def create_filled_in_gaps_df(full_era5_data, filled_gaps_data, measurements_data):  
    import pandas as pd  
  
    def single_df(time, tas, name):  
        df = pd.DataFrame()  
        df["time"] = time  
        df.set_index("time", inplace=True)  
        df[name] = tas  
        return df  
  
    lon_left_idx, lon_right_idx, lon_nearest_idx = get_left_right_nearest_elem_in_sorted_array(era5_data.lon.values, station_lon % 360)  
    lat_left_idx, lat_right_idx, lat_nearest_idx = get_left_right_nearest_elem_in_sorted_array(era5_data.lat.values, station_lat)  
  
    era5_nearest_values = full_era5_data.variables["tas"][:, lon_nearest_idx, lat_nearest_idx]  
    reconstructed_values = filled_gaps_data.variables["tas"].stack(grid=['lat', 'lon']).values  
    reconstructed_values = [np.median(x) for x in reconstructed_values]  
  
    measurements_values = measurements_data.variables["tas"].stack(grid=['lat', 'lon']).values  
    measurements_values = [np.median(x) for x in measurements_values]  
  
    era_df = single_df(full_era5_data.variables["time"], era5_nearest_values, "era5_nearest")  
    reconstructed_df = single_df(filled_gaps_data.variables["time"], reconstructed_values, "filled_in_gaps")  
    measurements_df = single_df(measurements_data.variables["time"], measurements_values, "measurements")  
  
    # merge on time  
    merged = pd.concat([era_df, reconstructed_df, measurements_df], axis=1)  
    # drop rows where reconstructed and measurements are both nan simultaneously  
    return merged.dropna(subset=["filled_in_gaps", "measurements"], how="all")
```

```
In [ ]: hourly_filled_gaps_df = create_filled_in_gaps_df(era5_full_data, output_actual_gaps_data, measurements_data)  
hourly_filled_gaps_df
```

Out[]:

era5_nearest filled_in_gaps measurements

time

2017-04-24 07:00:00	281.133545	280.147217	NaN
2017-04-24 08:00:00	282.382568	281.665955	NaN
2017-04-24 09:00:00	283.316010	283.075714	NaN
2017-04-24 10:00:00	284.647461	NaN	287.723913
2017-04-24 11:00:00	285.885742	285.778595	NaN
...
2022-05-24 12:00:00	294.967529	NaN	295.348077
2022-05-24 13:00:00	294.676453	NaN	294.840909
2022-05-24 14:00:00	294.638306	NaN	294.950000
2022-05-24 15:00:00	294.390717	NaN	294.563333
2022-05-24 16:00:00	293.362244	NaN	294.381707

44554 rows × 3 columns

In []: `def plot_n_steps_of_filled_in_df(df, n=None, title=None, show_in_steps=False):`

```

    time = df.index.values

    era5_nearest_values = df["era5_nearest"].values - 273.15
    reconstructed_median_values = df["filled_in_gaps"].values - 273.15

    measurements_values = df["measurements"].values - 273.15

    import random

    if n is None:
        n = len(df)

    break_loop = 0
    we_have_measurements_and_reconstructed_values_in_this_timeframe = False
    while not we_have_measurements_and_reconstructed_values_in_this_timeframe and break_loop < 1000:

        # random slice of n consecutive datapoints

        slice_start = random.randint(0, len(time) - n)
        time_slice = slice(slice_start, slice_start + n)

        if n == len(df):
            we_have_measurements_and_reconstructed_values_in_this_timeframe = True
        else:
            break_loop += 1
            # check in time slice if we have measurements and reconstructed values
            we_have_measurements_and_reconstructed_values_in_this_timeframe = not (np.all(np.isnan(measurements_values[time_slice])) or np.all(np.isnan(reconstructed_median_v

    time = time[time_slice]

    make_plots_with_era5 = [False, True, True] if show_in_steps else [True]
    make_plots_with_filled = [False, False, True] if show_in_steps else [True]
```

```
for show_era, show_filled in zip(make_plots_with_era5, make_plots_with_filled):

    plt.ylabel("Temperature at surface [C°]")

    # x-axis labels 90 degrees
    plt.xticks(rotation=45)

    # title
    if title is not None:
        plt.title(title)

    # font size of legend
    plt.rcParams.update({'font.size': 10})

    # font size of axis labels
    plt.rcParams.update({'axes.labelsize': 12})

    # font size of title
    plt.rcParams.update({'axes.titlesize': 26})

    # figure size A4 landscape
    plt.gcf().set_size_inches(16, 8)

    if show_era:
        plt.plot(time, era5_nearest_values[time_slice], label="ERA5 nearest point", color="grey", linestyle="--" if n <= 168 else "-")

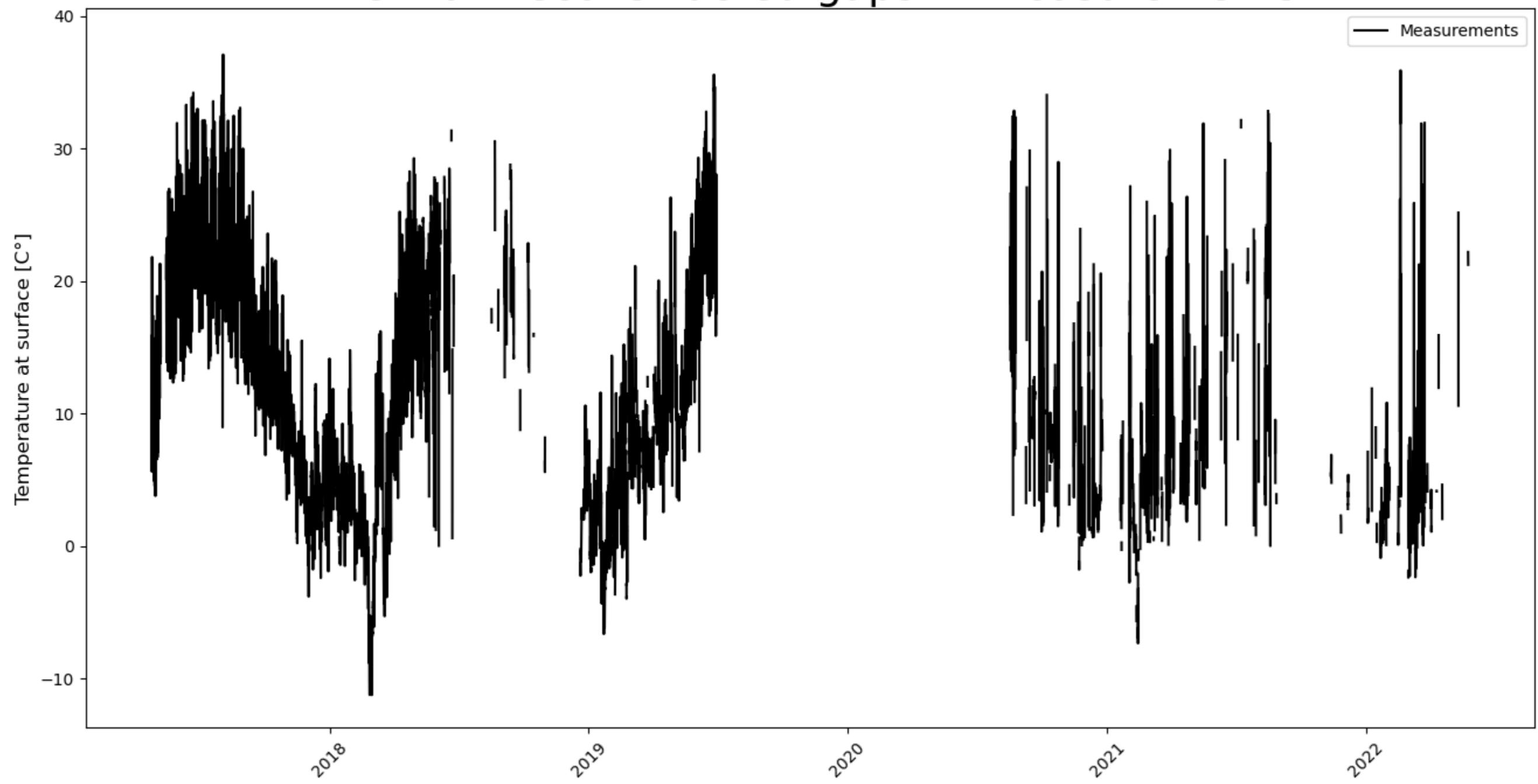
    plt.plot(time, measurements_values[time_slice], label="Measurements", color="black", linestyle="-", marker="o" if n <= 168 else "")

    if show_filled:
        plt.plot(time, reconstructed_median_values[time_slice], label="Reconstructed", color="#d13297", linestyle="-", marker="o" if n <= 168 else "")

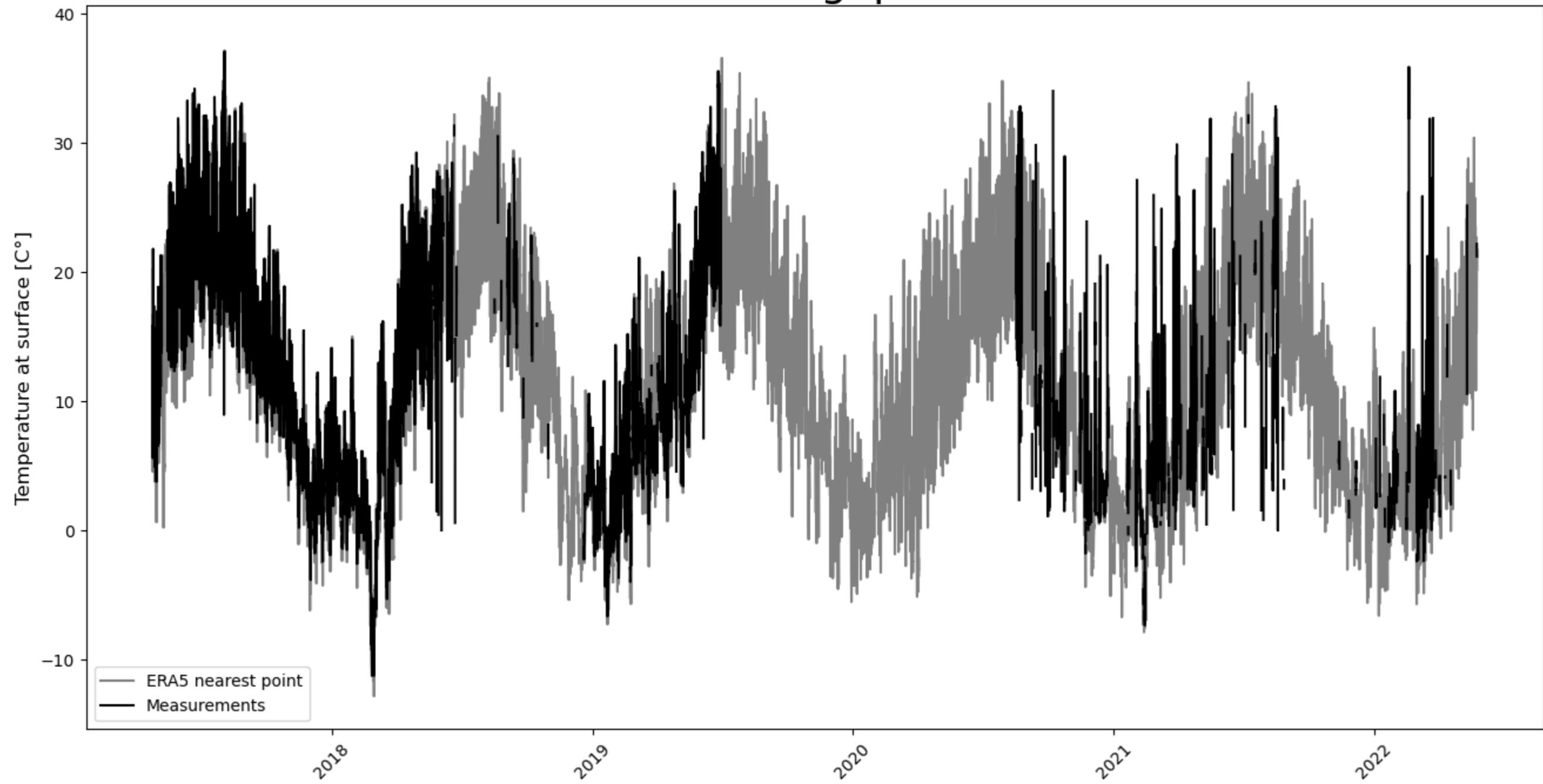
    plt.legend()
    plt.show()
```

```
In [ ]: plot_n_steps_of_filled_in_df(hourly_filled_gaps_df, title=f"{station_name} – reconstructed gaps in measurements", show_in_steps=True)
```

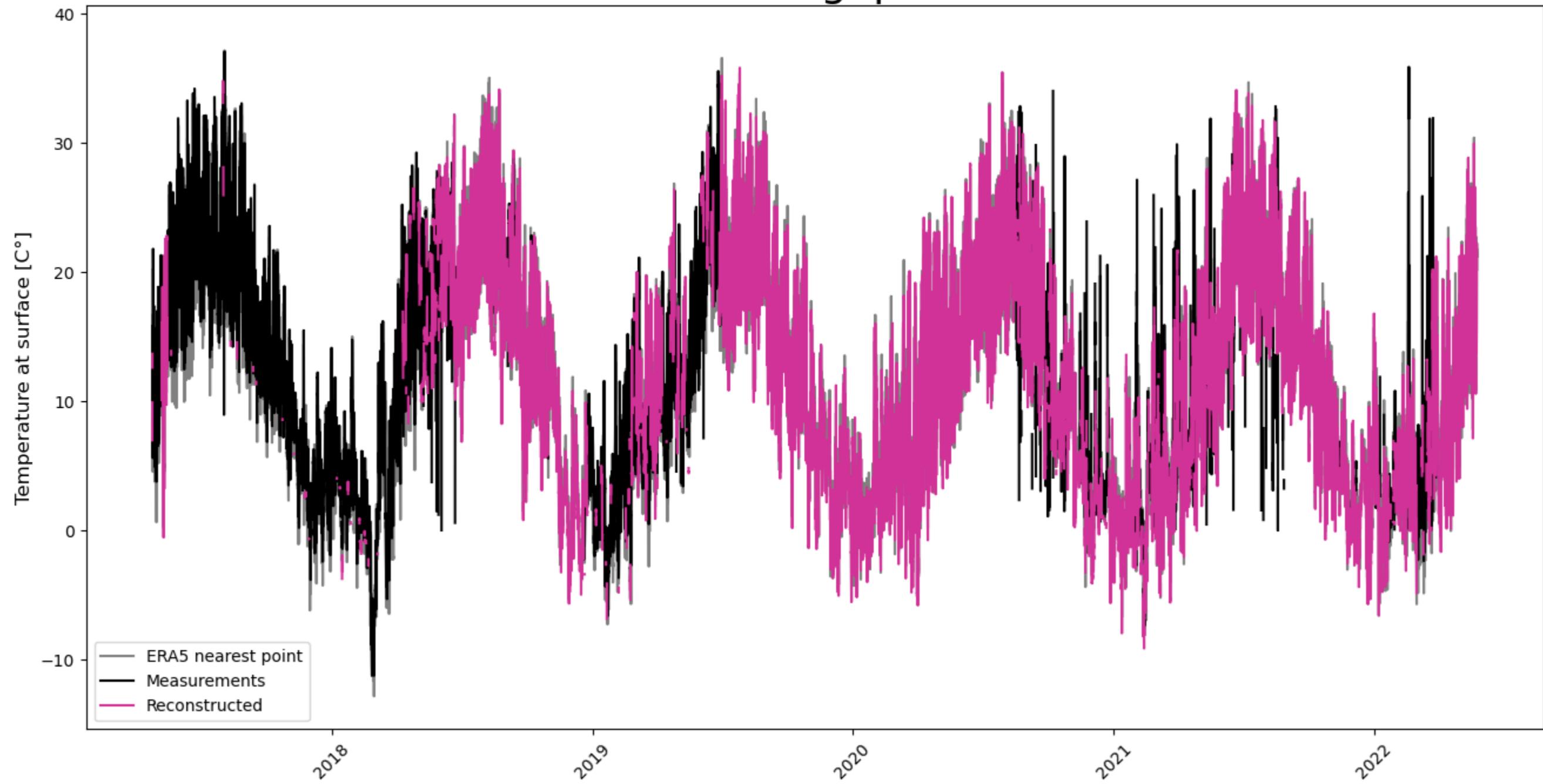
Vienna - reconstructed gaps in measurements



Vienna - reconstructed gaps in measurements



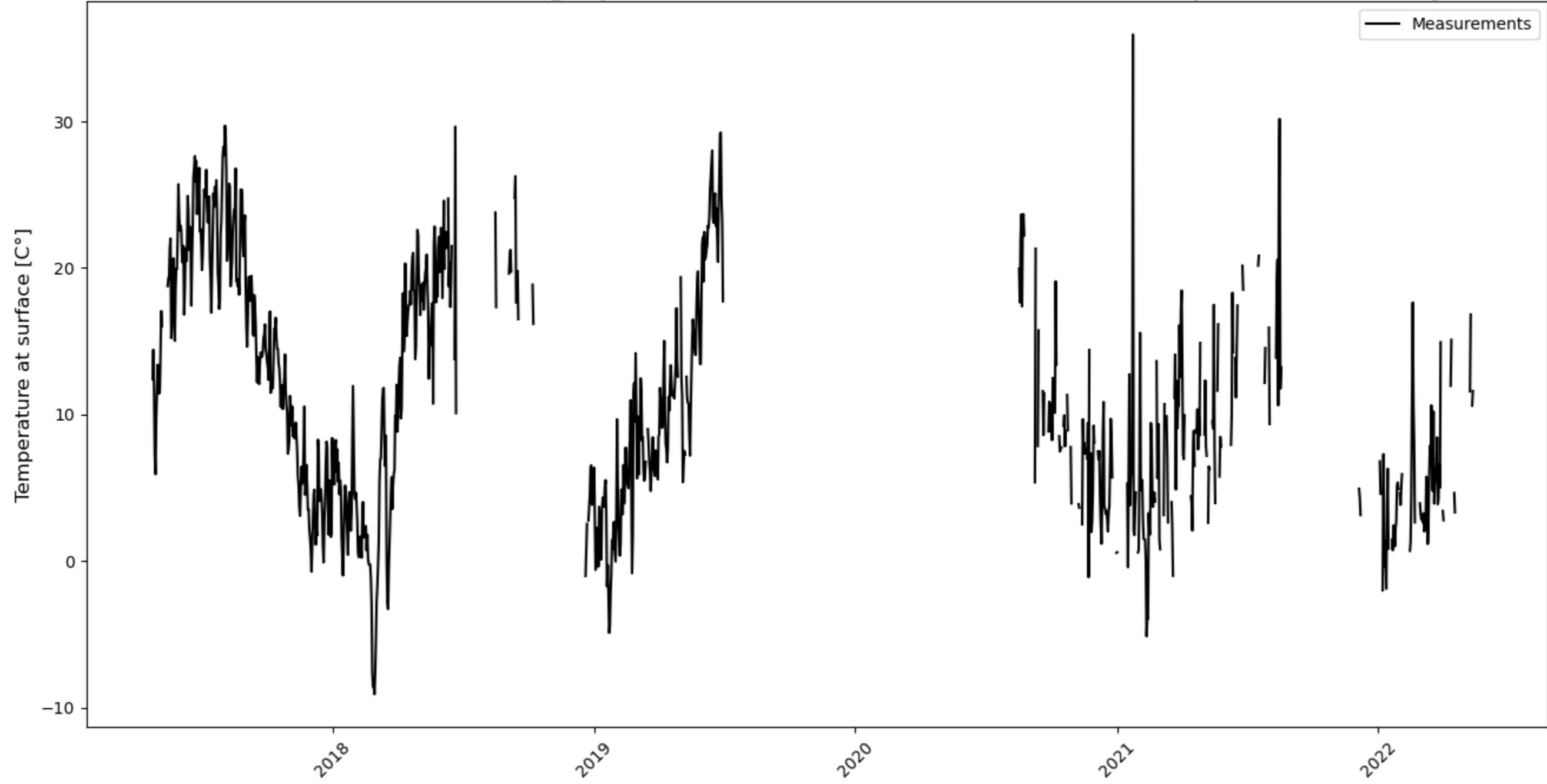
Vienna - reconstructed gaps in measurements



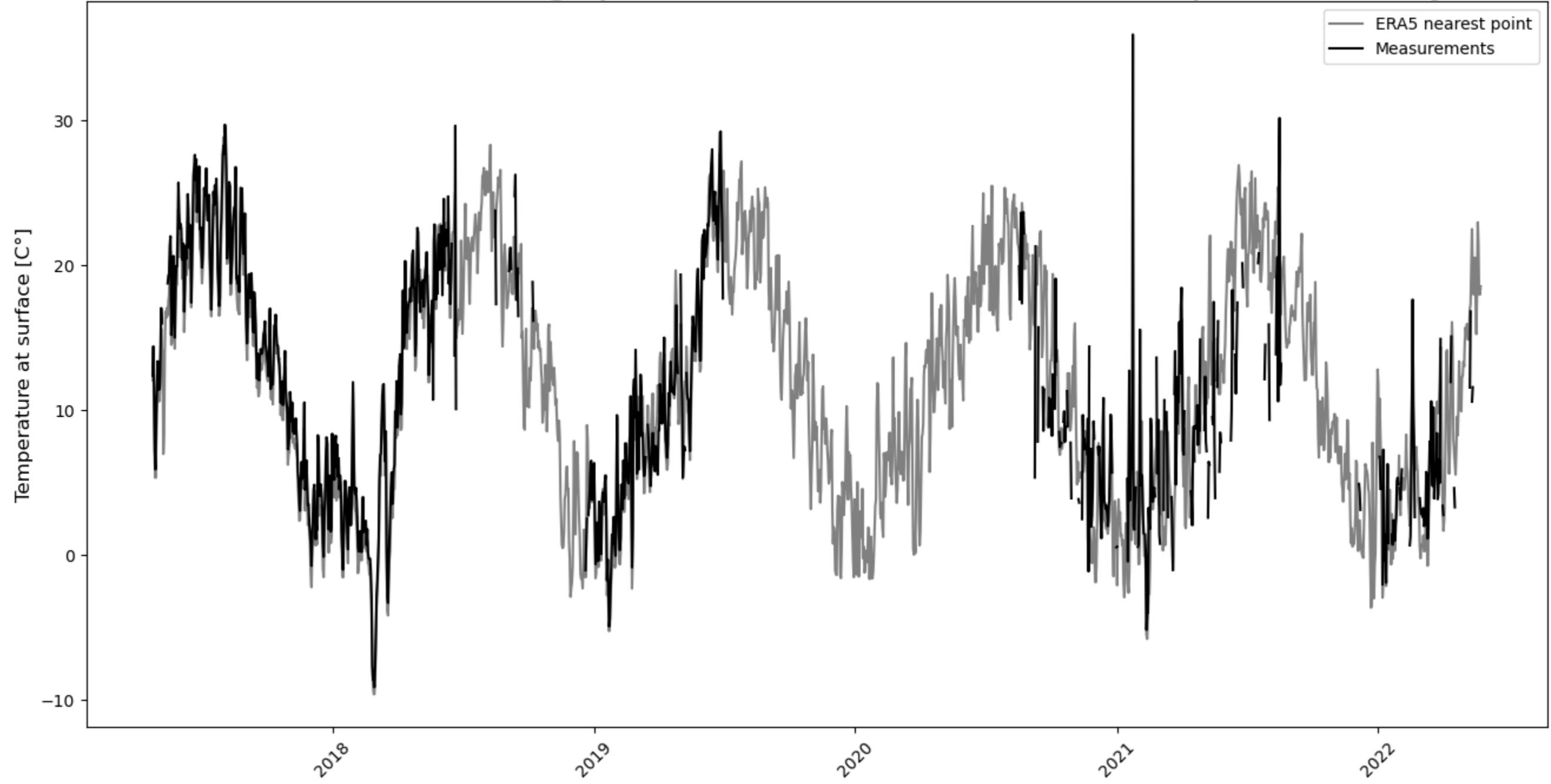
Resampled to Daily Mean

```
In [ ]: daily_filled_gaps_df = hourly_filled_gaps_df.resample("D").mean()  
plot_n_steps_of_filled_in_df(daily_filled_gaps_df, title=f"{station_name} - reconstructed gaps in measurements, resampled to daily mean", show_in_steps=True)
```

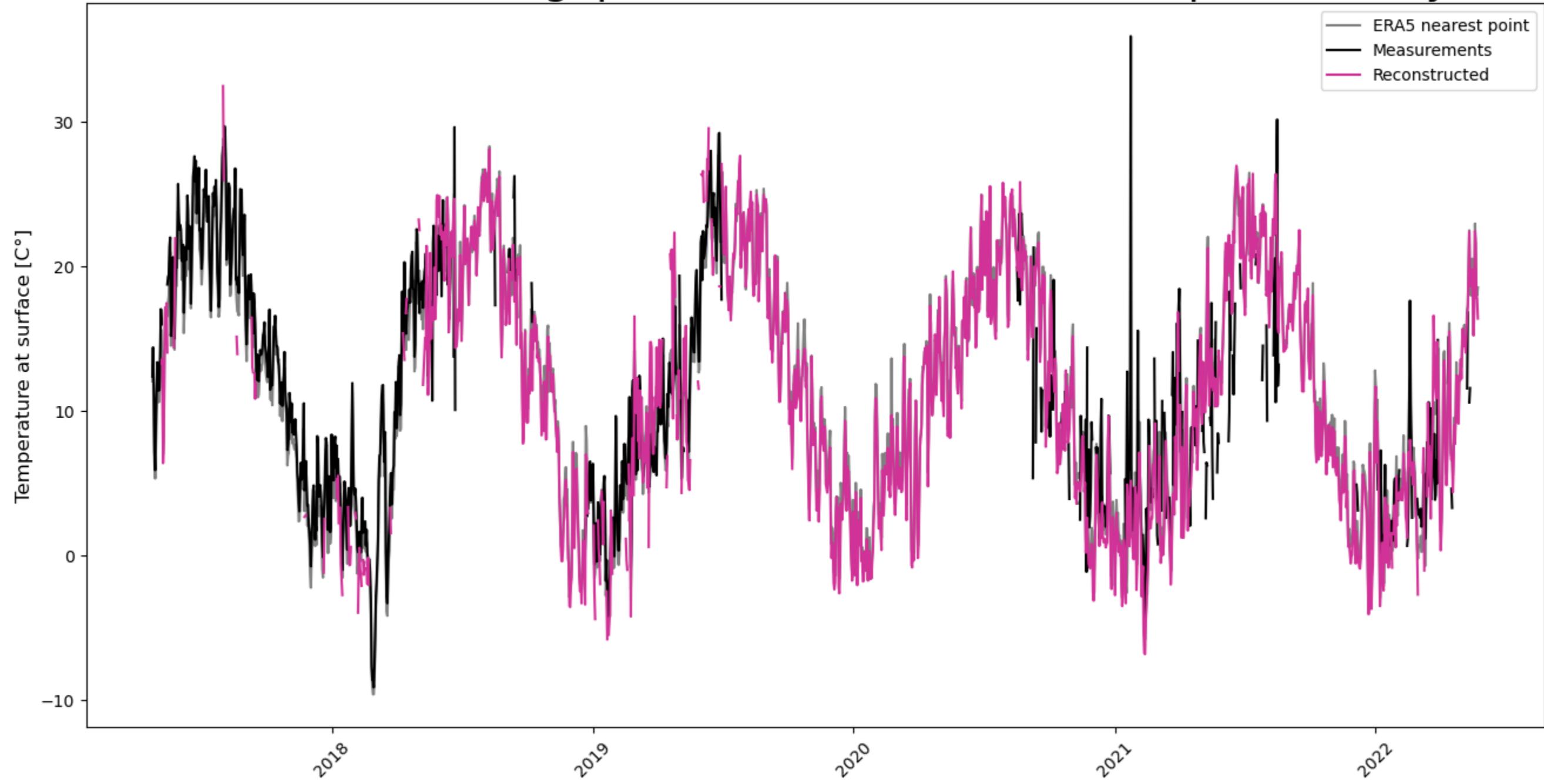
Vienna - reconstructed gaps in measurements, resampled to daily mean



Vienna - reconstructed gaps in measurements, resampled to daily mean



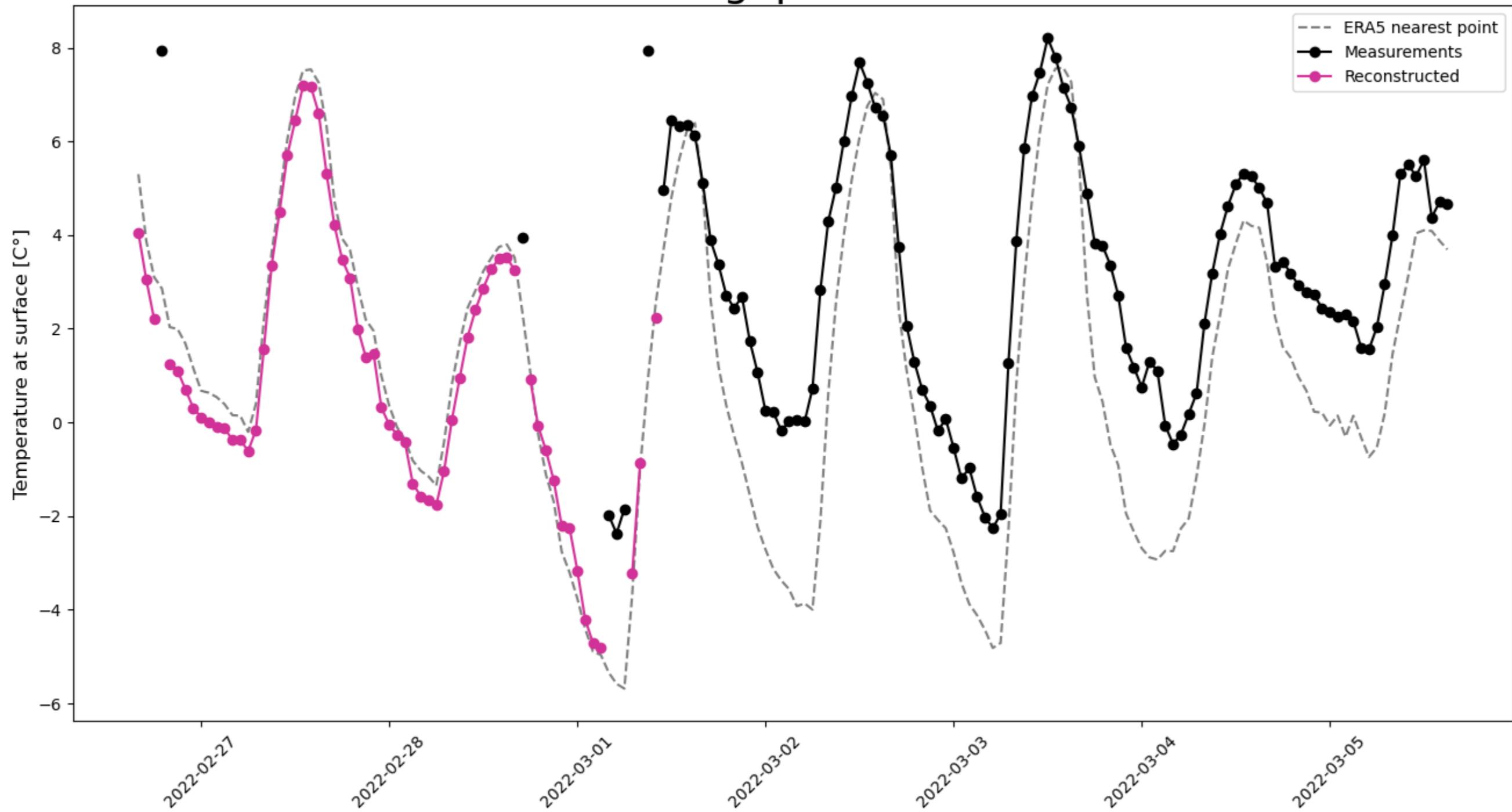
Vienna - reconstructed gaps in measurements, resampled to daily mean



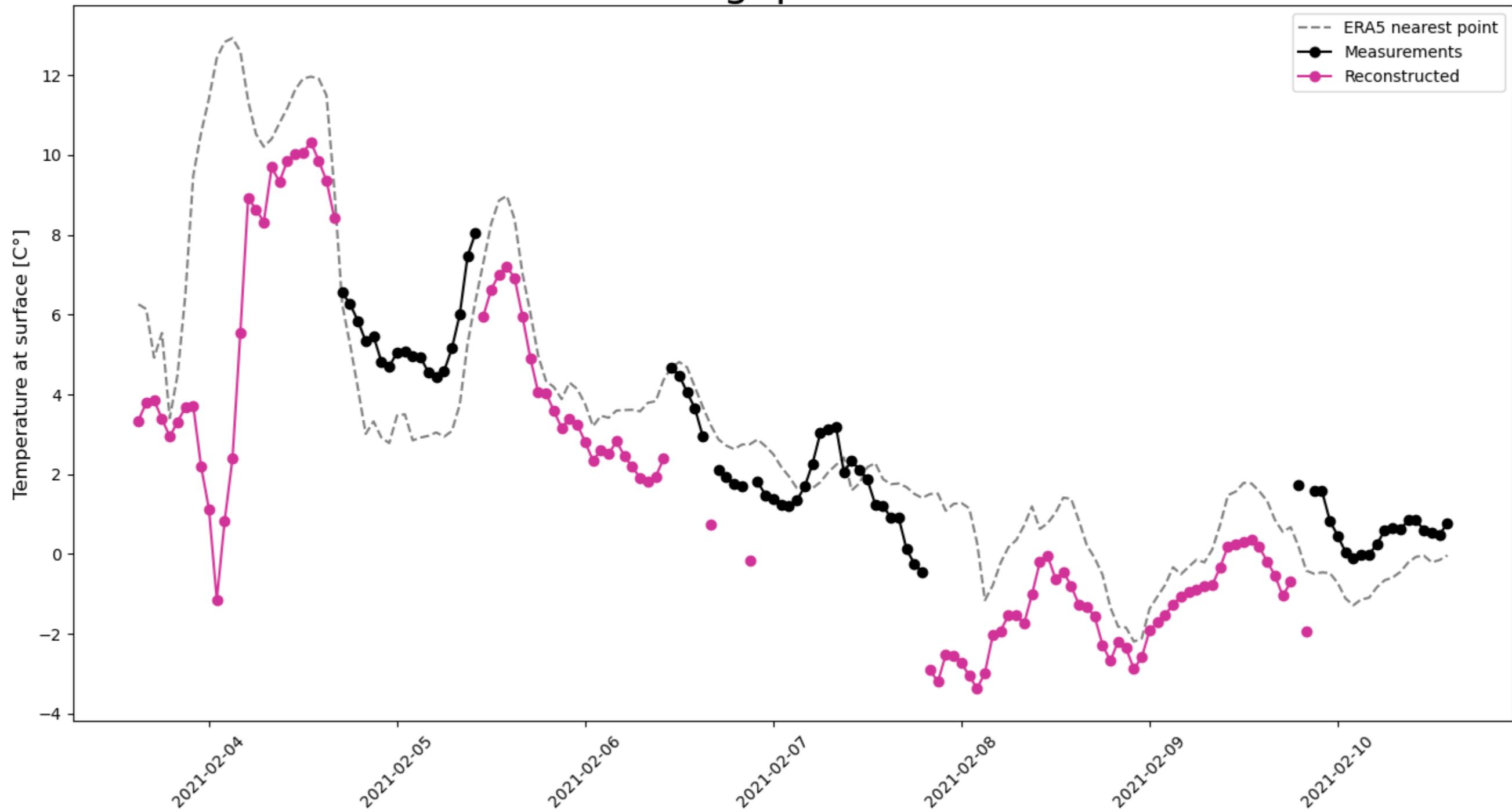
Plot random weeks

```
In [ ]: for i in range(10):
    plot_n_steps_of_filled_in_df(hourly_filled_gaps_df, n=168, title=f"{station_name} - reconstructed gaps - view of a random week")
```

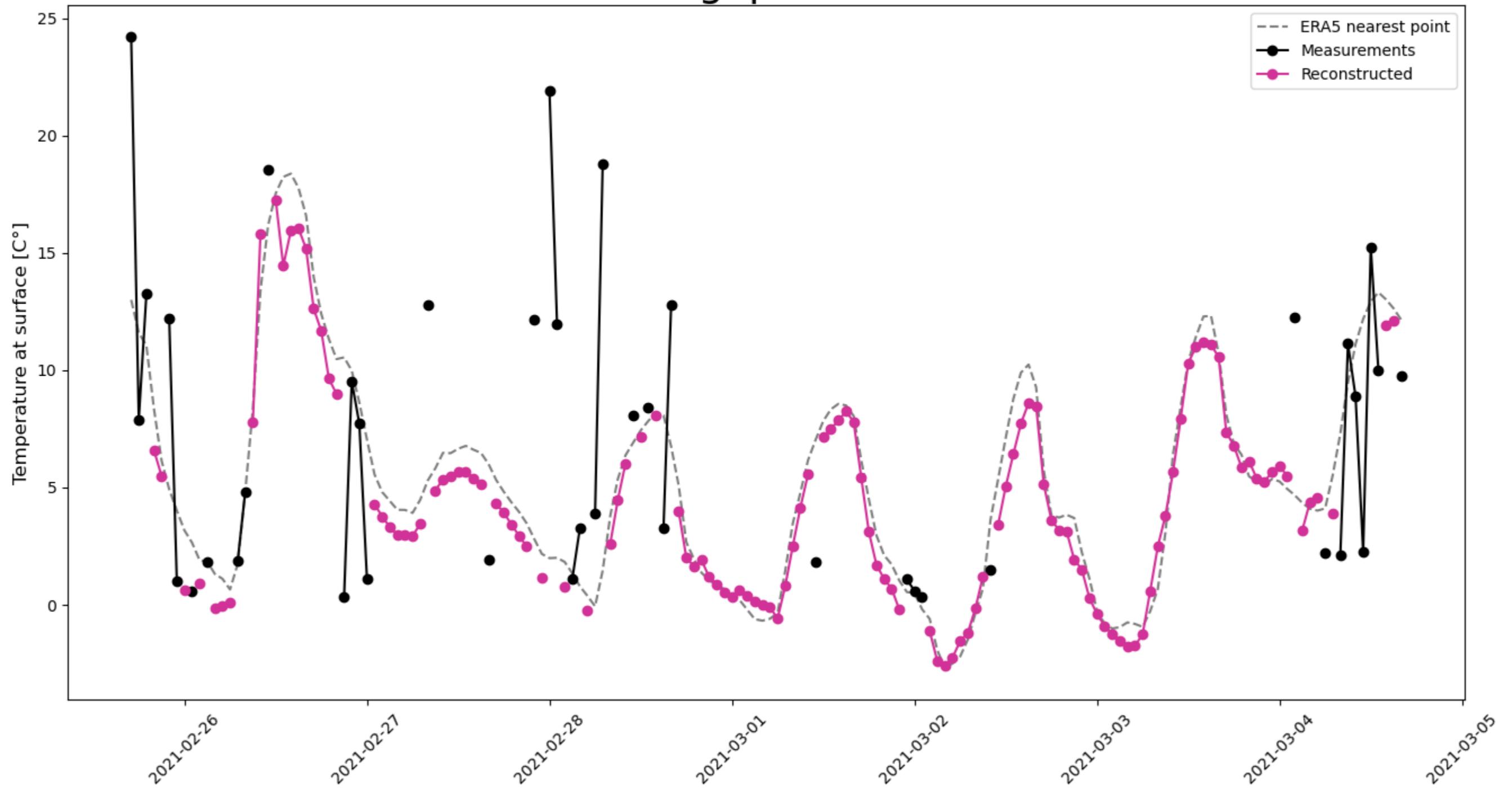
Vienna - reconstructed gaps - view of a random week



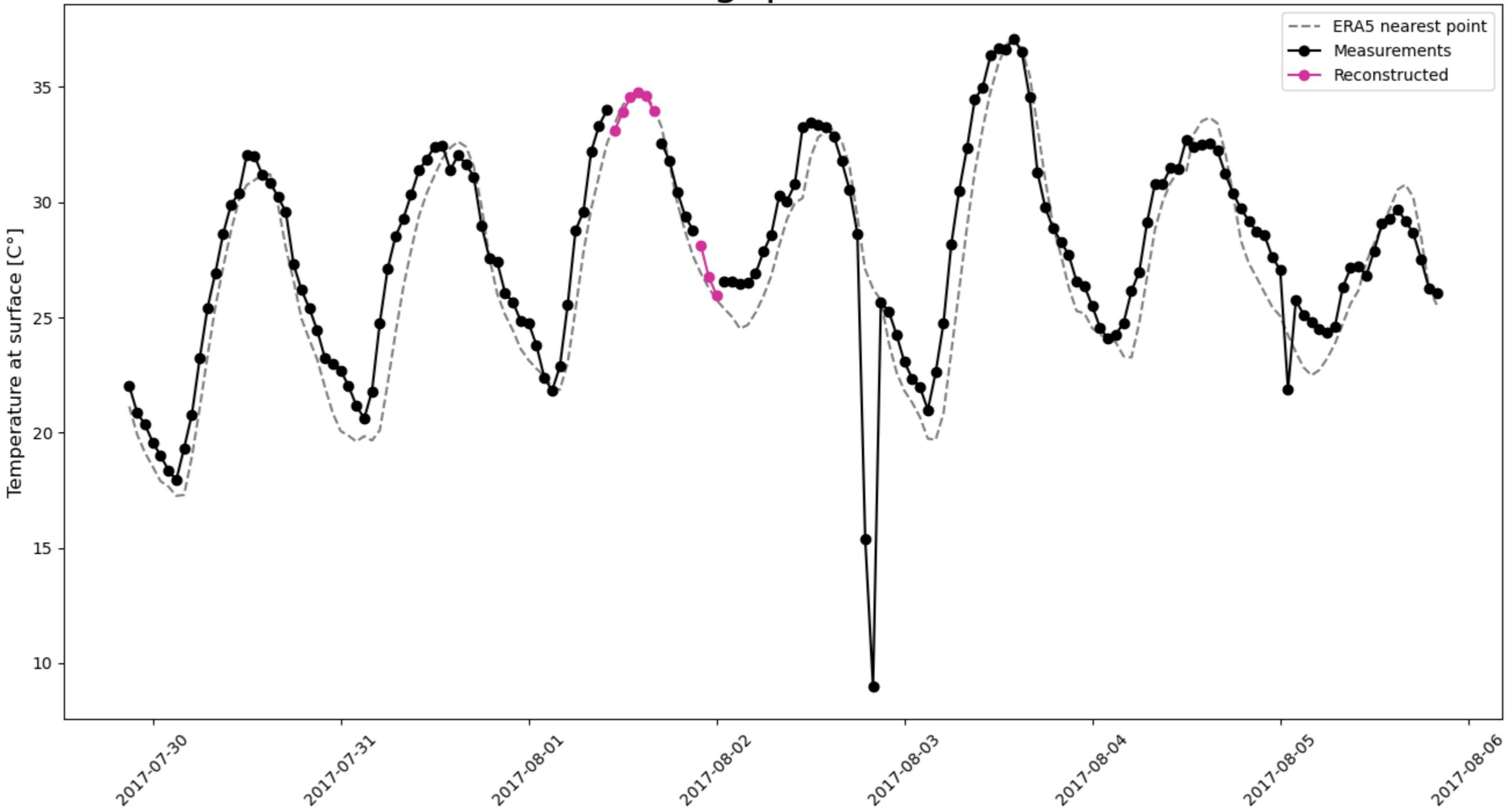
Vienna - reconstructed gaps - view of a random week



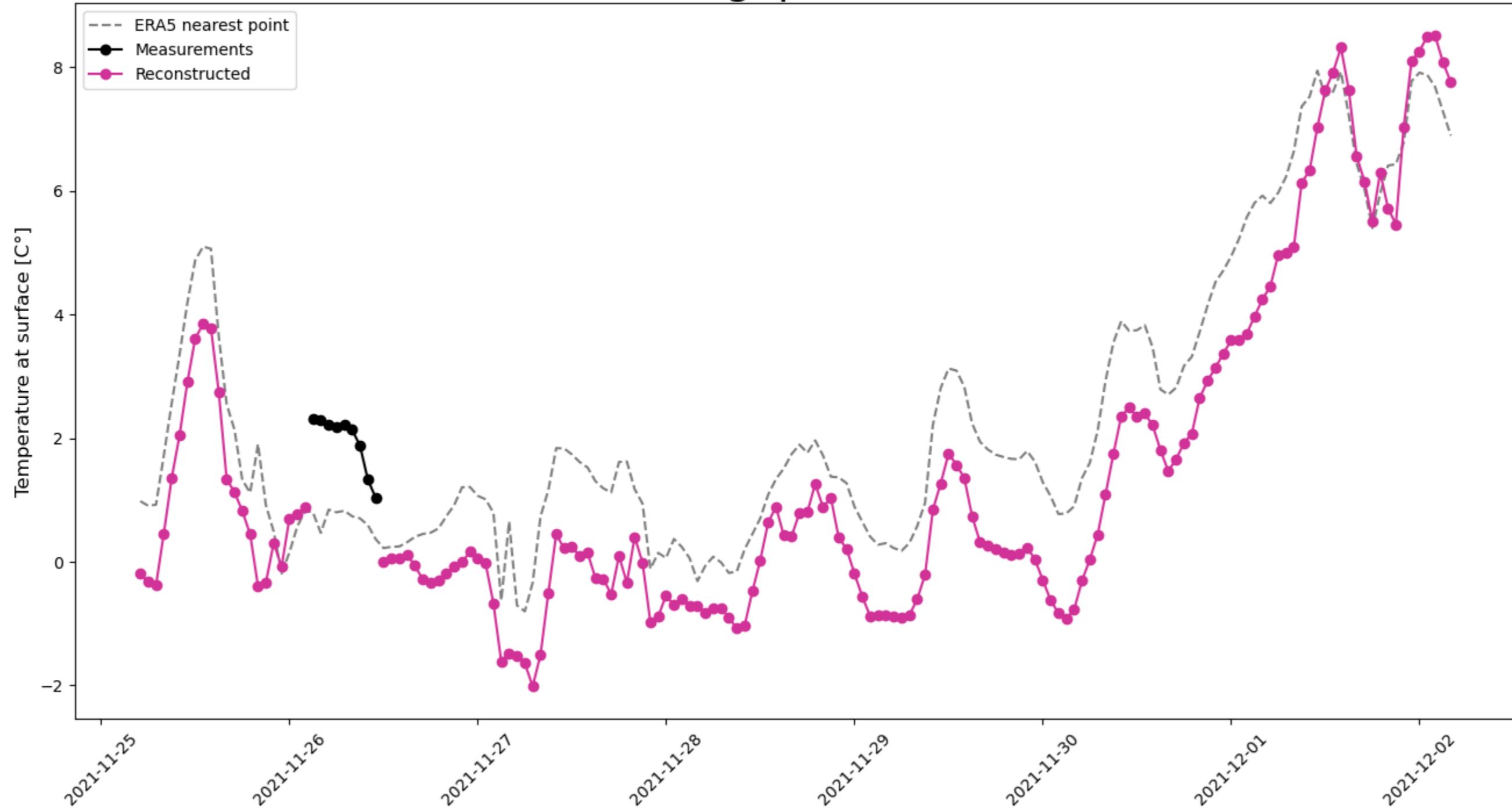
Vienna - reconstructed gaps - view of a random week



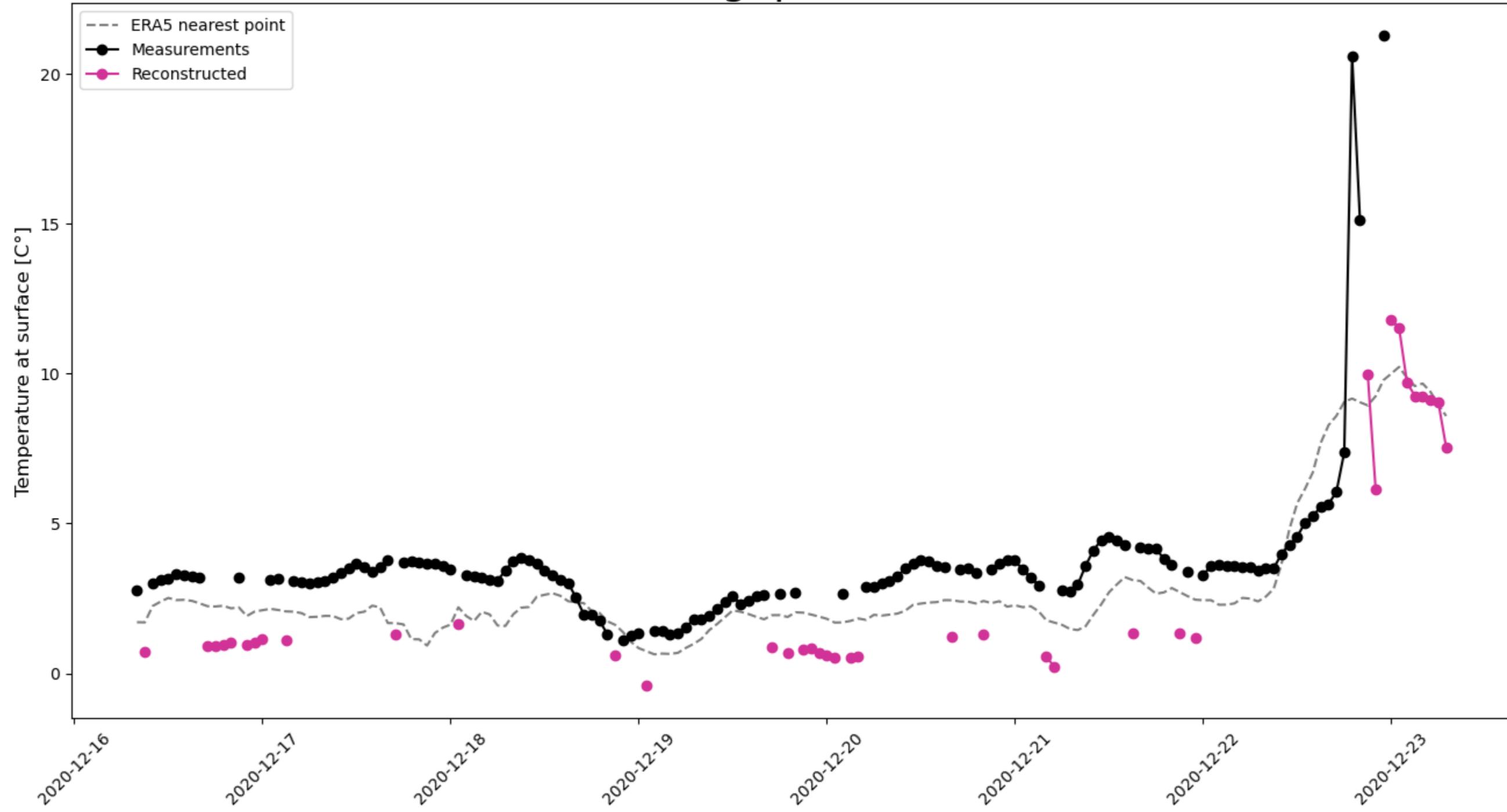
Vienna - reconstructed gaps - view of a random week



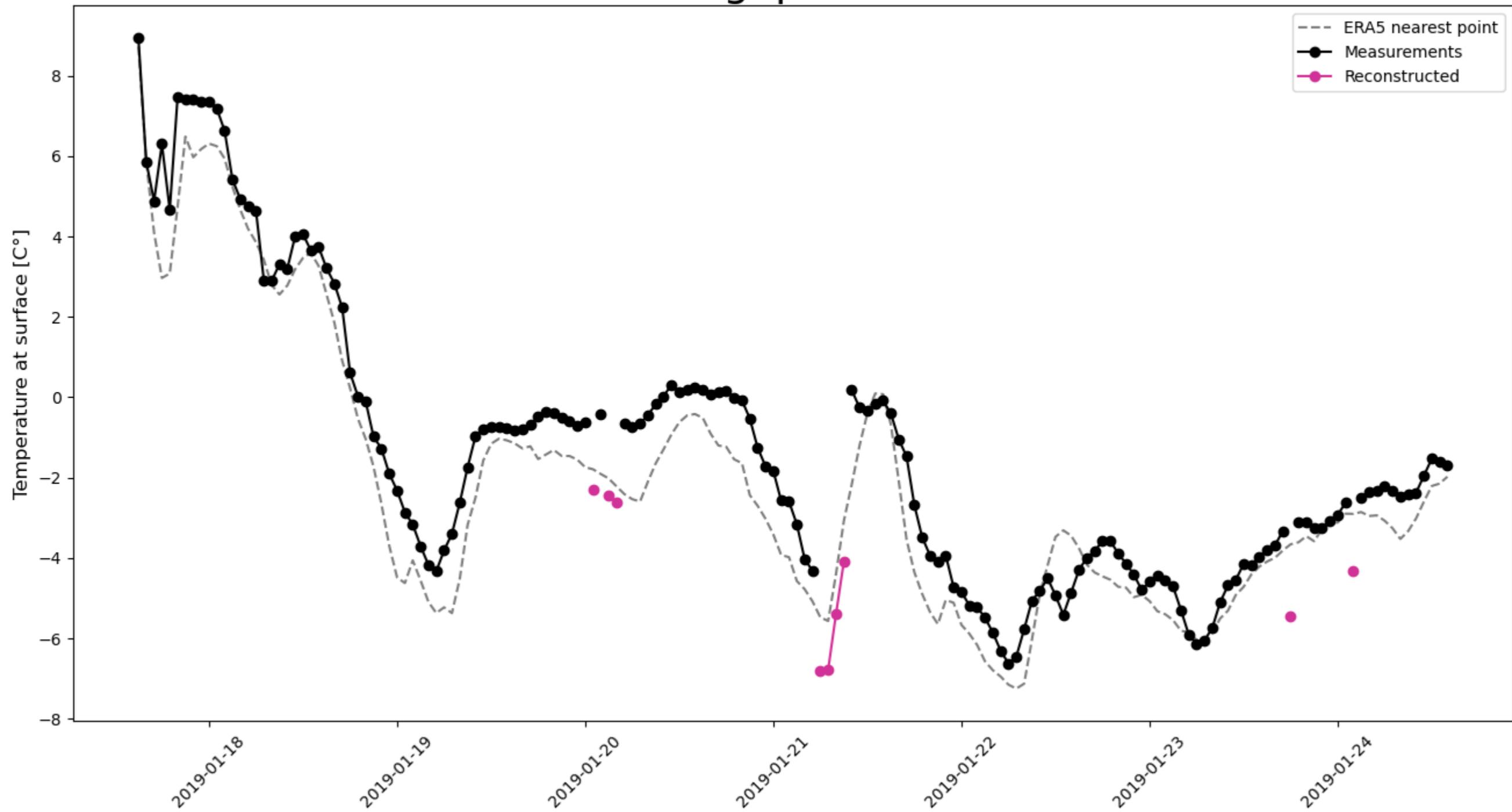
Vienna - reconstructed gaps - view of a random week



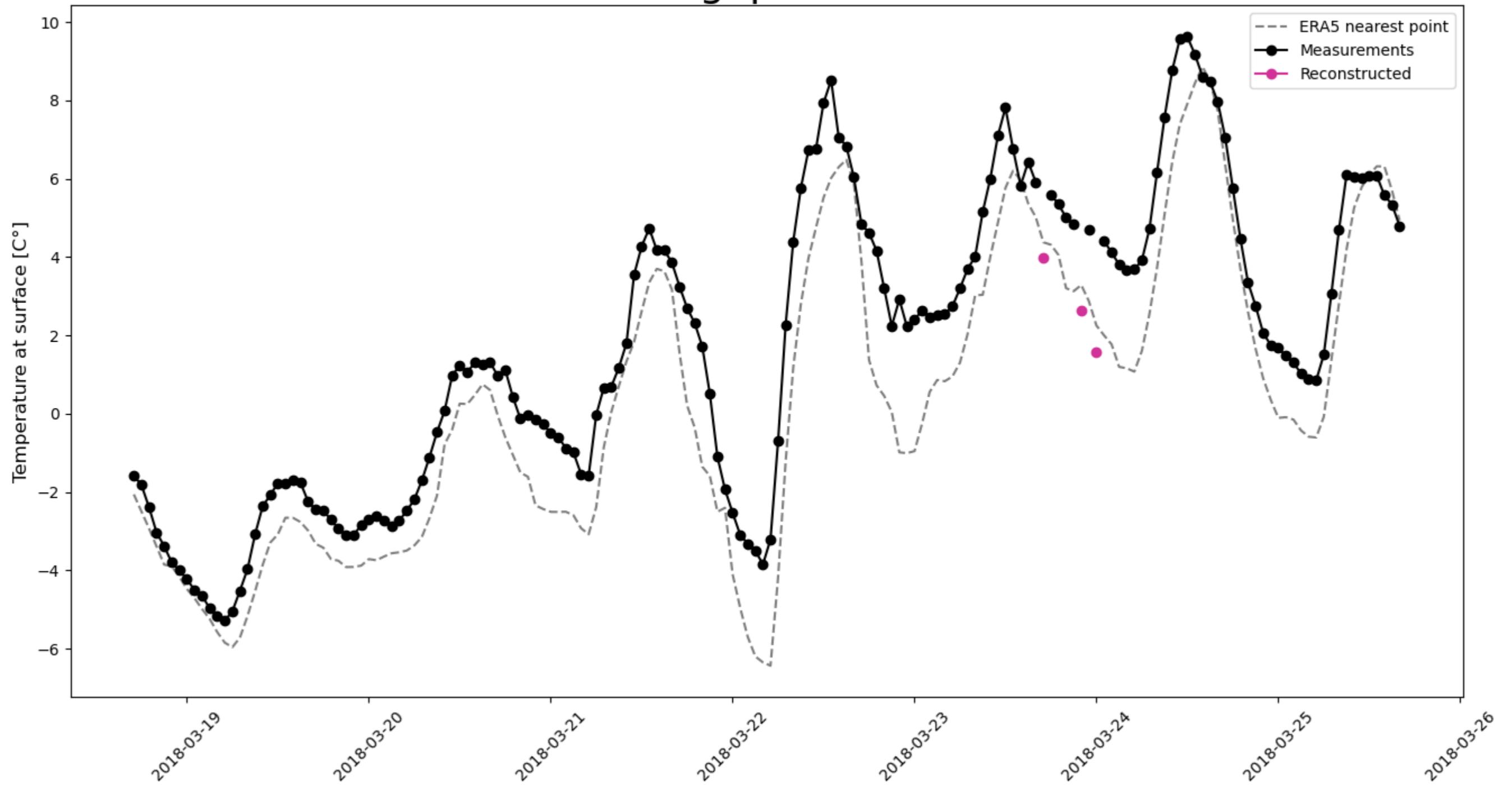
Vienna - reconstructed gaps - view of a random week



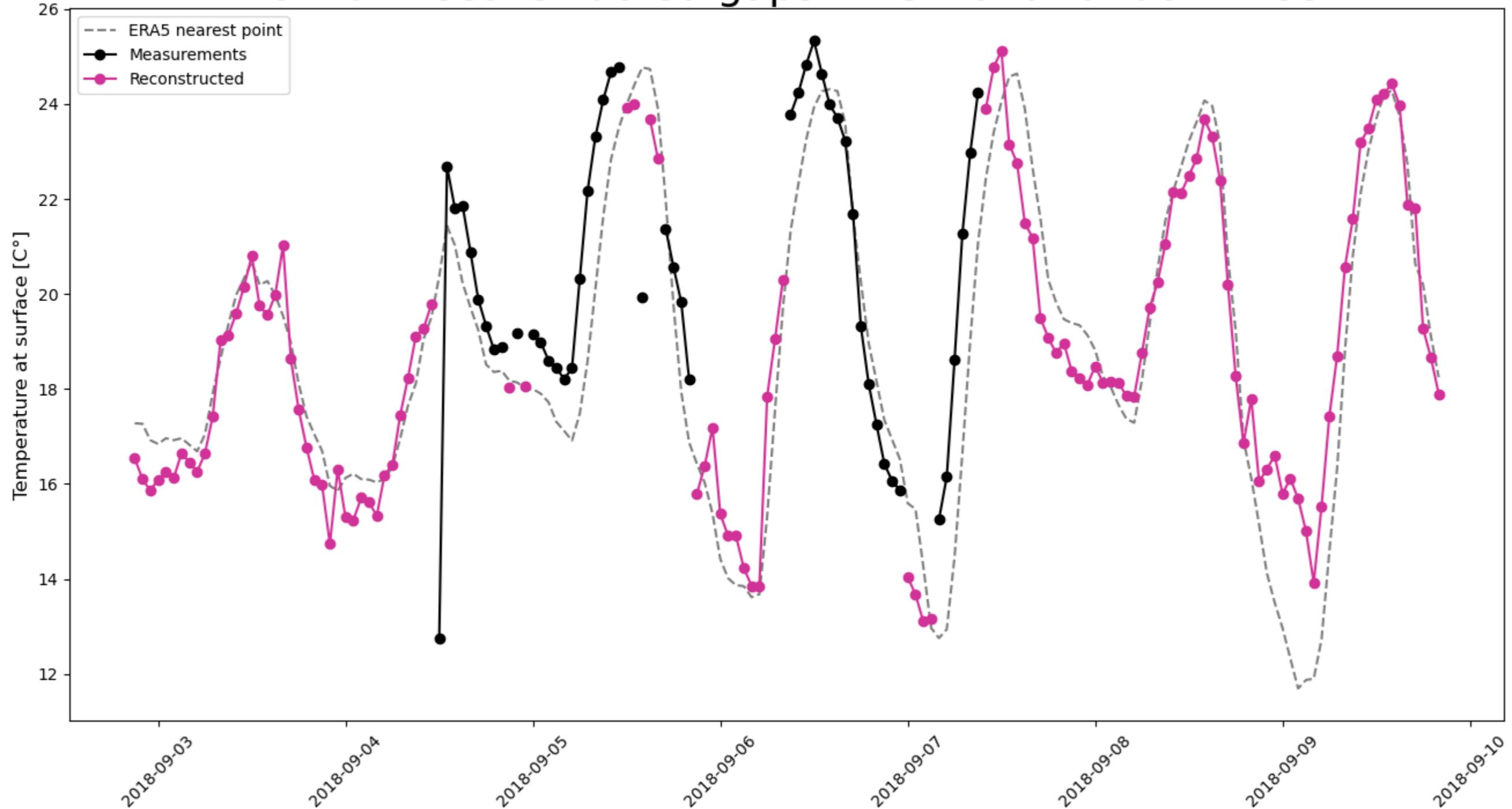
Vienna - reconstructed gaps - view of a random week



Vienna - reconstructed gaps - view of a random week



Vienna - reconstructed gaps - view of a random week



Vienna - reconstructed gaps - view of a random week

