

Imports

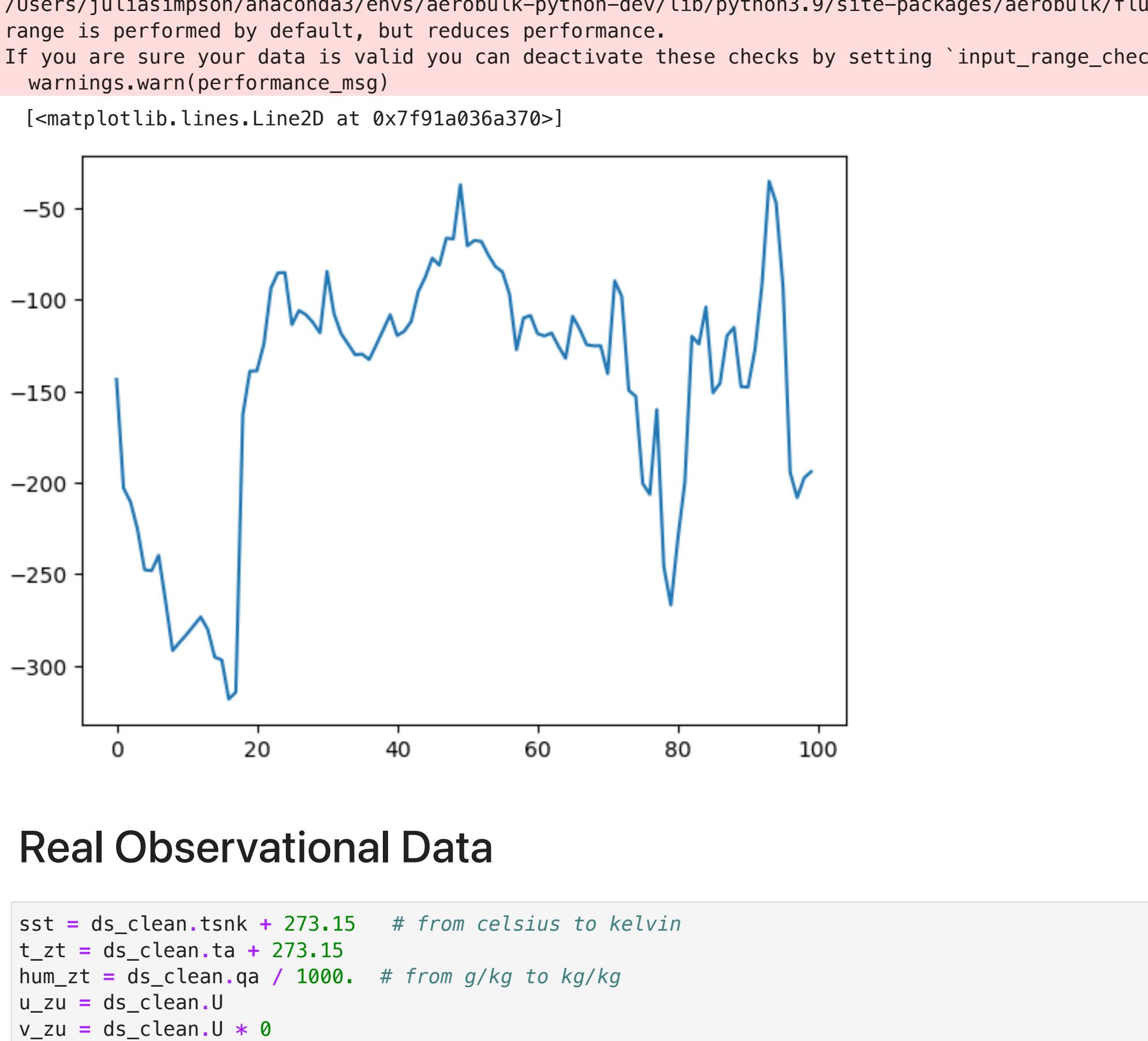
```
In [5]: import numpy as np  
import pandas as pd  
import datetime as dt  
from datetime import datetime  
import xarray as xr  
  
import matplotlib.pyplot as plt  
from matplotlib import cm  
import matplotlib.dates as mdates  
import cartopy.crs as ccrs  
  
import random  
import decimal  
import aerobulk  
from aerobulk.flux import noskin_np, skin_np, noskin, skin
```

Create Data

```
In [6]: ds = xr.load_dataset('~/ml-fluxes/data/fluxes_all_cruises_compilation.nc')  
ds_clean = ds.dropna(dim='time', how='any',  
subset=['taux','taucy','hsc','hlc','U','tsnk','ta','qa'])  
  
/Users/juliasimpson/anaconda3/envs/aerobulk-python-dev/lib/python3.9/site-packages/xarray/coding/times.py:254: RuntimeWarning: invalid value encountered in cast  
flat_num_dates_ns_int = (flat_num_dates * _NS_PER_TIME_DELTA[delta]).astype(  
/Users/juliasimpson/anaconda3/envs/aerobulk-python-dev/lib/python3.9/site-packages/xarray/coding/times.py:254: RuntimeWarning: invalid value encountered in cast  
flat_num_dates_ns_int = (flat_num_dates * _NS_PER_TIME_DELTA[delta]).astype(
```

```
In [7]: fig = plt.figure(figsize=(10, 5))  
ax = fig.add_subplot(1, 1, 1, projection=ccrs.Robinson())  
ax.set_global()  
ax.stock_img()  
ax.coastlines()  
colors = cm.tab10(np.linspace(0, 1, 10))  
  
ax.scatter(ds_clean.lon,ds_clean.lat,s=ds_clean.tsink,transform=ccrs.PlateCarree())
```

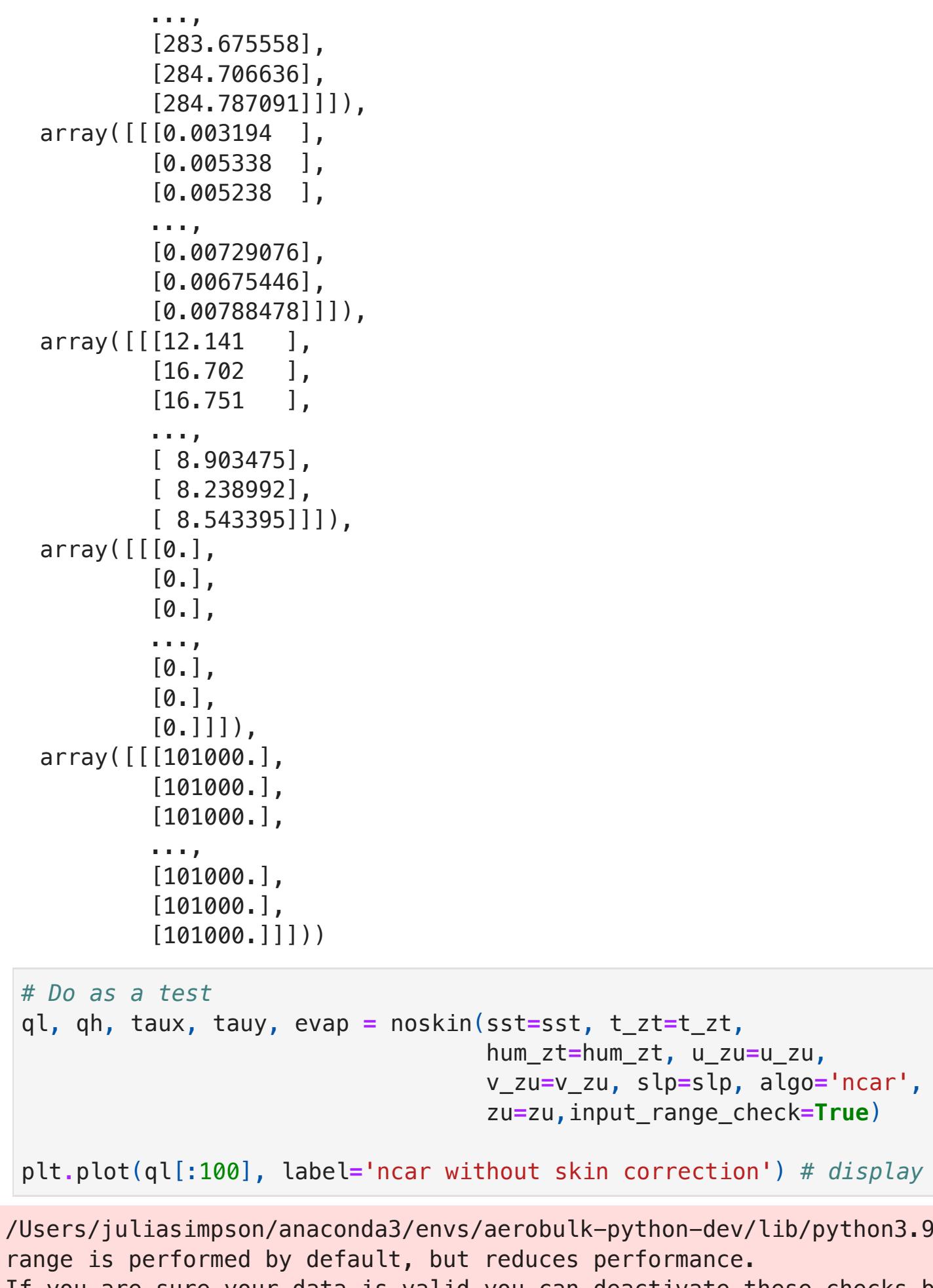
```
Out[7]: <matplotlib.collections.PathCollection at 0x7f919288f730>
```



```
In [8]: # Do as a test  
ql, qh, taux, taucy, evap = noskin(sst=ds_clean.tsink + 273.15, t_zt=ds_clean.ta + 273.15,  
hum_zt=ds_clean.qa / 1000., u_zu=ds_clean.U,  
v_zu=ds_clean.U@0, slp=ds_clean.U/ds_clean.U*101000.0, algo='ncar', zt=ds_clean.zu.to_numpy(),  
zu=ds_clean.zu.to_numpy())  
  
plt.plot(ql[:100], label='ncar without skin correction') # display first 100 points
```

```
/Users/juliasimpson/anaconda3/envs/aerobulk-python-dev/lib/python3.9/site-packages/aerobulk/flux.py:280: UserWarning: Checking for misaligned nans and values outside of the valid range is performed by default, but reduces performance.  
If you are sure your data is valid you can deactivate these checks by setting `input_range_check=False`  
warnings.warn(performance_msg)
```

```
Out[8]: <matplotlib.lines.Line2D at 0x7f91a036a370>
```



Real Observational Data

```
In [9]: sst = ds_clean.tsink + 273.15 # from celsius to kelvin  
t_zt = ds_clean.ta + 273.15  
hum_zt = ds_clean.qa / 1000. # from g/kg to kg/kg  
u_zu = ds_clean.U  
v_zu = ds_clean.U * 0  
zt = ds_clean.zt.to_numpy()  
zu = ds_clean.zu.to_numpy()  
slp=ds_clean.U/ds_clean.U*101000.0
```

```
In [10]: ocean_index = np.where(~np.isnan(sst))  
# Shrink the input data (i.e. remove all land points)  
args_shrunk = tuple(np.atleast_3d(ocean_index)) for a in (sst, t_zt, hum_zt, u_zu, v_zu, slp)
```

```
In [11]: ocean_index
```

```
Out[11]: (array([ 0, 1, 2, ..., 10076, 10077, 10078]),)
```

```
In [12]: args_shrunk
```

```
Out[12]: (array([[280.719 ],  
[284.971 ],  
[285.014 ],  
...,  
[284.051551],  
[284.525798],  
[285.807802]]),  
array([[278.091 ],  
[281.345 ],  
[281.742 ],  
...,  
[283.675558],  
[284.706636],  
[284.787091]]),  
array([[10.003194 ],  
[0.005338 ],  
[0.005238 ],  
...,  
[0.00729076],  
[0.00675446],  
[0.00884478]]),  
array([[12.141 ],  
[16.702 ],  
[16.751 ],  
...,  
[ 8.903475],  
[ 8.238992],  
[ 8.543395]])),  
array([[10.],  
[0.],  
[0.],  
...,  
[0.],  
[0.],  
[0.]]),  
array([[101000.],  
[101000.],  
[101000.],  
...,  
[101000.],  
[101000.],  
[101000.])))
```

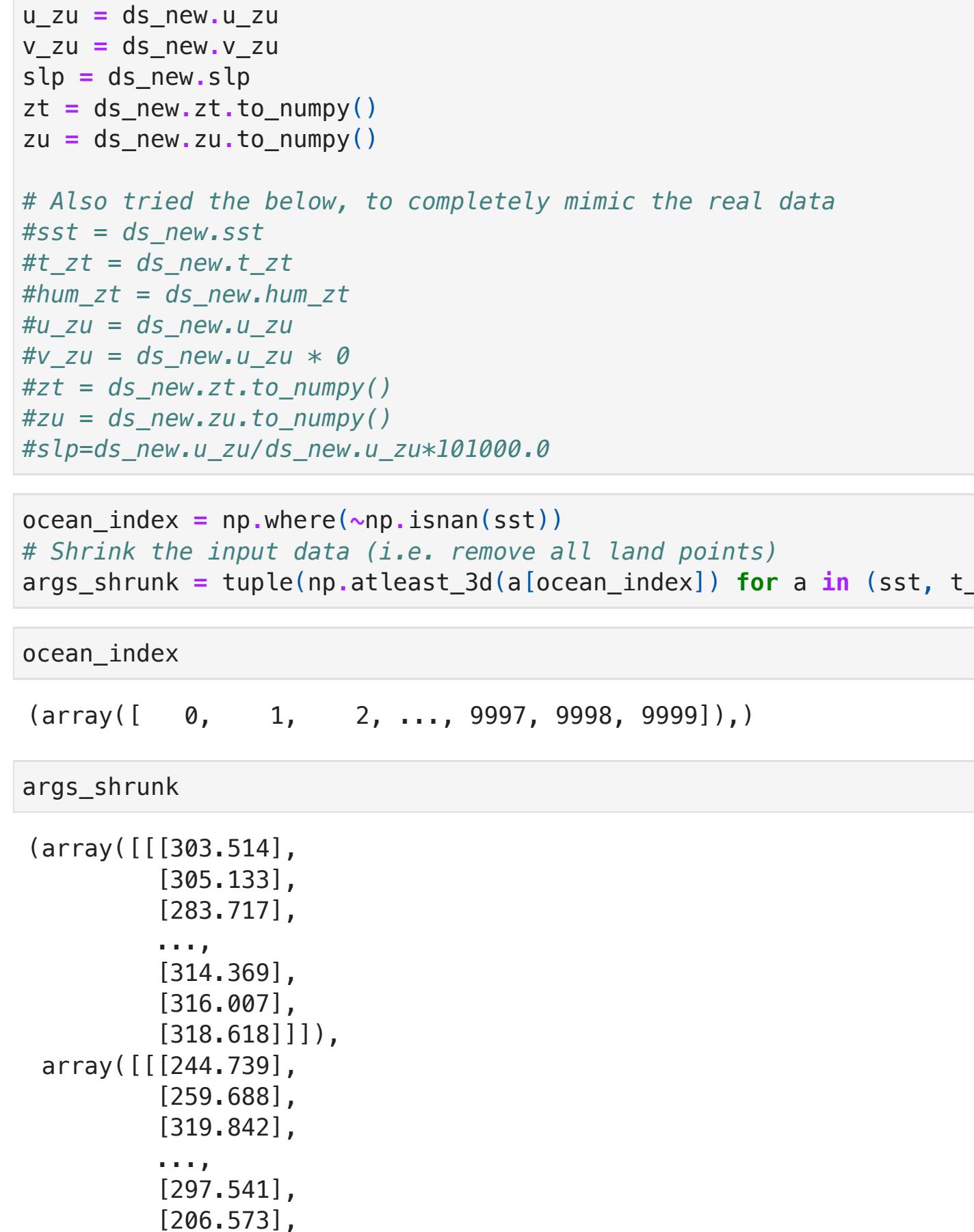
```
In [13]: # Do as a test
```

```
ql, qh, taux, taucy, evap = noskin(sst=sst, t_zt=t_zt,  
hum_zt=hum_zt, u_zu=u_zu,  
v_zu=v_zu, slp=slp, algo='ncar', zt=zt,  
zu=zu, input_range_check=True)
```

```
plt.plot(ql[:100], label='ncar without skin correction') # display first 100 points
```

```
/Users/juliasimpson/anaconda3/envs/aerobulk-python-dev/lib/python3.9/site-packages/aerobulk/flux.py:280: UserWarning: Checking for misaligned nans and values outside of the valid range is performed by default, but reduces performance.  
If you are sure your data is valid you can deactivate these checks by setting `input_range_check=False`  
warnings.warn(performance_msg)
```

```
Out[13]: <matplotlib.lines.Line2D at 0x7f9170393c40>
```



Synthetic Data

```
In [14]: # Valid data range according to documentation  
# https://github.com/gcm/aerobulk-python/blob/main/source/aerobulk/flux.py  
VALID_VALUE_RANGES = {'sst': [270, 320],  
't_zt': [180, 330],  
'hum_zt': [0, 0.08],  
'u_zu': [-50, 50],  
'v_zu': [-50, 50],  
'slp': [80000, 110000],  
'rad_sw': [0, 1500],  
'rad_lw': [0, 750]}
```

```
sst = []  
t_zt = []  
hum_zt = []  
u_zu = []  
v_zu = []  
slp = []  
zt = []  
zu = []
```

```
data_points = 10000
```

```
#INITIAL: don't specify any kind of distribution. Later, look at choices/ways in random sst, and adding for random temp  
for data_point in range(0, data_points):  
    sst.append(decimal.Decimal(random.randrange(VALID_VALUE_RANGES['sst'][0]*1000, VALID_VALUE_RANGES['sst'][1]*1000))/1000) #did *10 and then /10 to get 0.1 precision  
    t_zt.append(float(decimal.Decimal(random.randrange(VALID_VALUE_RANGES['t_zt'][0]*1000, VALID_VALUE_RANGES['t_zt'][1]*1000))/1000))  
    #t_zt.append(sst[data_point]*10) #instead of 10 do a random range of difference between 270-180 and 320-330  
    hum_zt.append(float(decimal.Decimal(random.randrange(VALID_VALUE_RANGES['hum_zt'][0]*1000000, VALID_VALUE_RANGES['hum_zt'][1]*1000000))/1000000))  
    u_zu.append(float(decimal.Decimal(random.randrange(VALID_VALUE_RANGES['u_zu'][0]*1000, VALID_VALUE_RANGES['u_zu'][1]*1000))/1000))  
    slp.append(101000.0)  
  
    zt.append(float(decimal.Decimal(random.randrange(12*10,19*10)/10))) #use 12 to 19, mirroring other dataset  
    zu.append(float(decimal.Decimal(random.randrange(15*10,22*10)/10))) #use 15 to 22, mirroring other dataset
```

```
In [15]: time = ds.time[:10000].data
```

```
ds_new = xr.Dataset()  
data_vars=dict()  
sst=[{"time": sst},  
t_zt=[{"time": t_zt},  
hum_zt=[{"time": hum_zt},  
u_zu=[{"time": u_zu},  
v_zu=[{"time": v_zu},  
slp=[{"time": slp},  
zt=[{"time": zt},  
zu=[{"time": zu},  
}],  
coords=dict(  
time=time  
),  
attrs=dict(description="Organizing synthetic data into a set.",),
```

```
In [16]: sst = ds_new.sst  
t_zt = ds_new.t_zt  
hum_zt = ds_new.hum_zt  
u_zu = ds_new.u_zu  
v_zu = ds_new.v_zu  
slp = ds_new.slp  
zt = ds_new.zt.to_numpy()  
zu = ds_new.zu.to_numpy()
```

```
# Also tried the below, to completely mimic the real data  
#sst = ds_new.sst  
#t_zt = ds_new.t_zt  
#hum_zt = ds_new.hum_zt  
#u_zu = ds_new.u_zu  
#v_zu = ds_new.v_zu * 0  
#zt = ds_new.zt.to_numpy()  
#zu = ds_new.zu.to_numpy()  
#slp=ds_new.u_zu/ds_new.u_zu*101000.0
```

```
In [17]: ocean_index = np.where(~np.isnan(sst))  
# Shrink the input data (i.e. remove all land points)  
args_shrunk = tuple(np.atleast_3d(ocean_index)) for a in (sst, t_zt, hum_zt, u_zu, v_zu, slp)
```

```
In [18]: ocean_index
```

```
Out[18]: (array([ 0, 1, 2, ..., 9997, 9998, 9999]),)
```

```
In [19]: args_shrunk
```

```
Out[19]: (array([[303.514],  
[305.133],  
[283.717],  
...,  
[314.369],  
[316.097],  
[318.618]]),  
array([[244.339],  
[259.688],  
[319.842],  
...,  
[297.541],  
[206.573],  
[280.632]]),  
array([[0.022024],  
[0.03621],  
[0.027963],  
...,  
[0.073098],  
[0.07194],  
[0.00215]]),  
array([[ 9.172],  
[ 26.483],  
[-26.573],  
...,  
[ 35.721],  
[-3.509],  
[-30.651]]),  
array([[0.1],  
[0.1],  
[0.1],  
...,  
[0.1],  
[0.1],  
[0.1]]),  
array([[101000.],  
[101000.],  
[101000.],  
...,  
[101000.],  
[101000.],  
[101000.])))
```

```
THE BELOW CELL CRASHES THE KERNEL
```

```
In [ ]: # Do as a test  
ql, qh, taux, taucy, evap = noskin(sst=sst, t_zt=t_zt,  
hum_zt=hum_zt, u_zu=u_zu,  
v_zu=v_zu, slp=slp, algo='ncar', zt=zt,  
zu=zu, input_range_check=True)
```

```
plt.plot(ql[:100], label='ncar without skin correction') # display first 100 points
```