

# Wide vs. Long Format Data

## About the data

In this notebook, we will be using daily temperature data from the [National Centers for Environmental Information \(NCEI\) API](#). We will use the Global Historical Climatology Network - Daily (GHCND) data set for the Boonton 1 station (GHCND:USC00280907); see the documentation [here](#).

*Note: The NCEI is part of the National Oceanic and Atmospheric Administration (NOAA) and, as you can see from the URL for the API, this resource was created when the NCEI was called the NCDC. Should the URL for this resource change in the future, you can search for the NCEI weather API to find the updated one.*

```
import matplotlib.pyplot as plt
import pandas as pd

wide_df = pd.read_csv('data/wide_data.csv', parse_dates=['date'])
long_df = pd.read_csv(
    'data/long_data.csv',
    usecols=['date', 'datatype', 'value'],
    parse_dates=['date']
)[['date', 'datatype', 'value']] # sort columns
```

## Wide format

Our variables each have their own column:

```
wide_df.head(6)
```

	date	TMAX	TMIN	TOBS
0	2018-10-01	21.1	8.9	13.9
1	2018-10-02	23.9	13.9	17.2
2	2018-10-03	25.0	15.6	16.1
3	2018-10-04	22.8	11.7	11.7
4	2018-10-05	23.3	11.7	18.9
5	2018-10-06	20.0	13.3	16.1

Describing all the columns is easy:

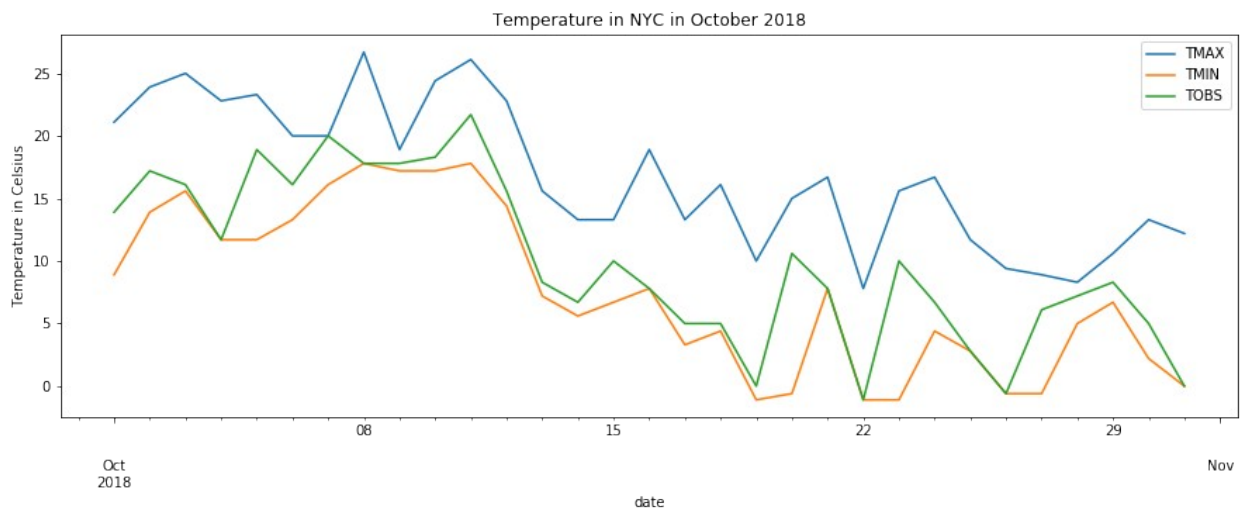
```
wide_df.describe(include='all')
```

	date	TMAX	TMIN	TOBS
count	31	31.000000	31.000000	31.000000
unique	31	NaN	NaN	NaN
top	2018-10-01 00:00:00	NaN	NaN	NaN

freq		1	NaN	NaN	NaN
first	2018-10-01 00:00:00		NaN	NaN	NaN
last	2018-10-31 00:00:00		NaN	NaN	NaN
mean		NaN	16.829032	7.561290	10.022581
std		NaN	5.714962	6.513252	6.596550
min		NaN	7.800000	-1.100000	-1.100000
25%		NaN	12.750000	2.500000	5.550000
50%		NaN	16.100000	6.700000	8.300000
75%		NaN	21.950000	13.600000	16.100000
max		NaN	26.700000	17.800000	21.700000

Easy to graph with `pandas` (covered in [chapter 5](#)):

```
wide_df.plot(
    kind='line', y=['TMAX', 'TMIN', 'TOBS'], x='date',
    title='Temperature in NYC in October 2018',
    figsize=(15, 5)
).set_ylabel('Temperature in Celsius')
plt.show()
```



## Long format

Our variable names are now in the `datatype` column and their values are in the `value` column. We now have 3 rows for each date, since we have 3 different `datatypes`:

```
long_df.head(6)
```

	date	datatype	value
0	2018-10-01	TMAX	21.1
1	2018-10-01	TMIN	8.9
2	2018-10-01	TOBS	13.9
3	2018-10-02	TMAX	23.9

```
4 2018-10-02      TMIN    13.9
5 2018-10-02      TOBS    17.2
```

Since we have many rows for the same date, using `describe()` is not trivial anymore:

```
long_df.describe(include='all')
```

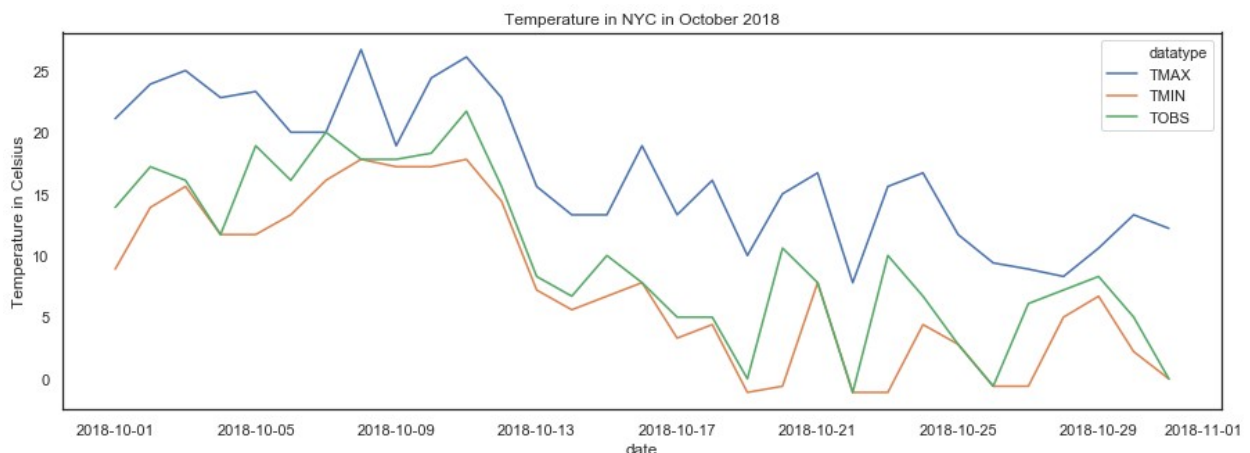
	date	datatype	value
count	93	93	93.000000
unique	31	3	NaN
top	2018-10-01 00:00:00	TOBS	NaN
freq	3	31	NaN
first	2018-10-01 00:00:00	NaN	NaN
last	2018-10-31 00:00:00	NaN	NaN
mean	NaN	NaN	11.470968
std	NaN	NaN	7.362354
min	NaN	NaN	-1.100000
25%	NaN	NaN	6.700000
50%	NaN	NaN	11.700000
75%	NaN	NaN	17.200000
max	NaN	NaN	26.700000

Plotting long format data in `pandas` can be rather tricky. Instead we use `seaborn` (covered in [chapter 6](#)):

```
import seaborn as sns

sns.set(rc={'figure.figsize':(15, 5)}, style='white')

ax = sns.lineplot(data=long_df, hue='datatype', y='value', x='date')
ax.set_ylabel('Temperature in Celsius')
ax.set_title('Temperature in NYC in October 2018')
plt.show()
```



With long data and `seaborn`, we can easily facet our plots:

```

sns.set(rc={'figure.figsize':(20, 10)}, style='white', font_scale=2)

g = sns.FacetGrid(long_df, col="datatype", height=10)
g = g.map(plt.plot, "date", "value")
g.set_titles(size=25)
g.set_xticklabels(rotation=45)
plt.show()

```

