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Python for Data Analytics

Pandas III

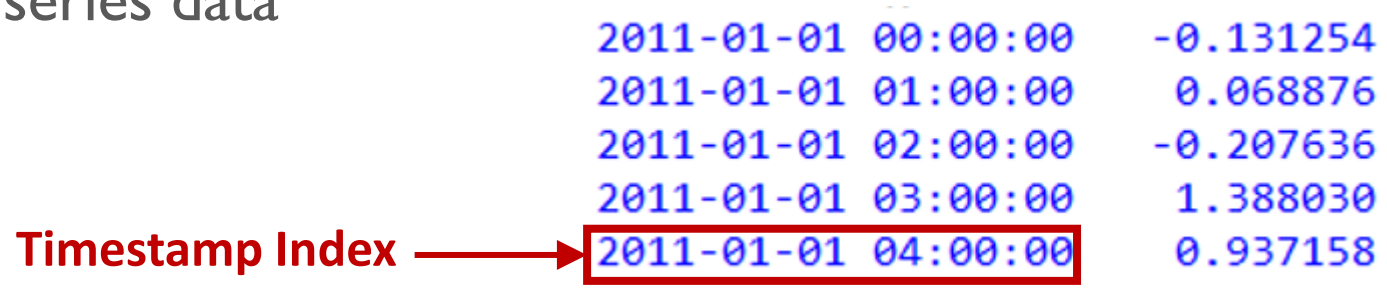
Outline

- Why Pandas?
- Pandas Series
- Pandas DataFrame
  - Creating DataFrame
  - Manipulating Columns
  - Manipulating Rows
  - Arithmetic operations
  - Group Aggregation
  - Hierarchical Indexing
  - Combining and Merging
  - Time Series Data

Time Series Data

Time Series Data

- How to analyze time series data?
  - Sample time series data



- Time series data is the data with the timestamp index
  - How to parse time series information from various sources and formats?
  - How to generate sequences of fixed-frequency dates and time spans
  - How to manipulate and convert date times with timezone information?
  - How to group data by time?
  - ...

Python datetime Module

- `datetime.datetime` class: a combination of date and time
  - year, month, day, hour, minute, second, microsecond, tzinfo
- `datetime.now()`: return the current local datetime

```
import numpy as np
import pandas as pd
```

```
import datetime as dt
now = dt.datetime.now()
print(now)
```

```
2021-01-09 22:56:48.365683
```

```
newyear = dt.datetime(2021, 1, 1)
print(newyear)
```

```
2021-01-01 00:00:00
```

```
print(now - newyear)
```

```
8 days, 22:56:48.365683
```

NumPy datetime64 Type

- NumPy supports datetime functionality with the data type called '`datetime64`'
  - No timezone support

```
import numpy as np
now = np.datetime64('now')
print(now)
```

```
2021-01-09T13:59:25
```

```
np.arange('2021-01', '2021-07', dtype='datetime64[M]')
```

```
array(['2021-01', '2021-02', '2021-03', '2021-04', '2021-05', '2021-06'],
      dtype='datetime64[M]')
```

```
newyear = np.datetime64('2021-1-1')
```

```
---------------------------------------------------------------------------
ValueError                                Traceback (most recent call last)
<ipython-input-11-bc5dab85414d> in <module>
----> 1 newyear = np.datetime64('2021-1-1')

ValueError: Error parsing datetime string "2021-1-1" at position 5
```

Converting to Datetime

- Pandas supports extensive capabilities and features for working with time series data based on NumPy `datetime64`.
- *pd*.`to_datetime(`*arg*, ...`)`
  - Convert argument to datetime.
  - Return type can be a DatetimeIndex, Series, or Timestamp
  - *arg*: integer, float, string, datetime, list, tuple, 1-D array, Series

```
t = np.array(['1/11/2021', '2021/1/12', '20210113', '2021-1-14', '2021 1 15',
              '2021, 1, 16', 'Jan. 17 2021', '18 Jan 2021'])
pd.to_datetime(t)
```

```
DatetimeIndex(['2021-01-11', '2021-01-12', '2021-01-13', '2021-01-14',
               '2021-01-15', '2021-01-16', '2021-01-17', '2021-01-18'],
              dtype='datetime64[ns]', freq=None)
```

Generating DatetimeIndex

- *pd.*`date_range(`*start=None, end=None, periods=None, freq=None, ...*`)`
  - Return a fixed frequency DatetimeIndex
  - *start*: left bound for generating dates
  - *end*: right bound for generating dates
  - *periods*: the number of datetime to generate
  - *freq*: the time interval between consecutive datetime values (default: 'D')

Freq string	Description	Freq string	Description
`'D'`	One absolute day	`'M'`	Calendar month end
`'H'`	One hour	`'MS'`	Calendar month begin
`'T' or 'min'`	One minute	`'BM'`	Business month end
`'S'`	One second	`'BMS'`	Business month begin
`'B'`	Business day (weekday)	`'WOM-2THU'`	Second Thursday of the month
`'W'`	One week	`'1h30min'`	One and half hour

date_range() Examples (1)

- Default: everyday

```
pd.date_range('2021-1-11', '2021-1-17')
```

```
DatetimeIndex(['2021-01-11', '2021-01-12', '2021-01-13', '2021-01-14',
               '2021-01-15', '2021-01-16', '2021-01-17'],
              dtype='datetime64[ns]', freq='D')
```

- 7 days since 2021-1-11

```
pd.date_range('2021 1 11', periods=7)
```

```
DatetimeIndex(['2021-01-11', '2021-01-12', '2021-01-13', '2021-01-14',
               '2021-01-15', '2021-01-16', '2021-01-17'],
              dtype='datetime64[ns]', freq='D')
```

date_range() Examples (2)

- Just weekdays

```
pd.date_range('2021 1 1', '2021/1/20', freq='B')
```

```
DatetimeIndex(['2021-01-01', '2021-01-04', '2021-01-05', '2021-01-06',
               '2021-01-07', '2021-01-08', '2021-01-11', '2021-01-12',
               '2021-01-13', '2021-01-14', '2021-01-15', '2021-01-18',
               '2021-01-19', '2021-01-20'],
              dtype='datetime64[ns]', freq='B')
```

- Every Sunday

```
pd.date_range('2021-1-4', '2021-2-26', freq='W-SUN')
```

```
DatetimeIndex(['2021-01-10', '2021-01-17', '2021-01-24', '2021-01-31',
               '2021-02-07', '2021-02-14', '2021-02-21'],
              dtype='datetime64[ns]', freq='W-SUN')
```

date_range() Examples (3)

- First business day every two months

```
pd.date_range('2021-1-1', '2021-12-31', freq='2BMS')
```

```
DatetimeIndex(['2021-01-01', '2021-03-01', '2021-05-03', '2021-07-01',
               '2021-09-01', '2021-11-01'],
              dtype='datetime64[ns]', freq='2BMS')
```

- Every one and half hour

```
pd.date_range('2021-1-11 8:30', periods=7, freq='1h30min')
```

```
DatetimeIndex(['2021-01-11 08:30:00', '2021-01-11 10:00:00',
               '2021-01-11 11:30:00', '2021-01-11 13:00:00',
               '2021-01-11 14:30:00', '2021-01-11 16:00:00',
               '2021-01-11 17:30:00'],
              dtype='datetime64[ns]', freq='90T')
```

Finding the Day of the Week

- *pd*.DatetimeIndex.day_name(**args*, ...)
  - Return the day names of the DatetimeIndex

```
idx = pd.date_range(start='2021-01-01', freq='D', periods=3)
idx
```

```
DatetimeIndex(['2021-01-01', '2021-01-02', '2021-01-03'], dtype='datetime64[n
s]', freq='D')
```

```
idx.day_name()
```

```
Index(['Friday', 'Saturday', 'Sunday'], dtype='object')
```

```
war = pd.date_range(start='1950-6-25', freq='D', periods=3)
war.day_name()
```

```
Index(['Sunday', 'Monday', 'Tuesday'], dtype='object')
```

Creating Time Series Data

- Create a range of `DatetimeIndex` object

```
ts = pd.date_range('1/1/2021', periods=4, freq='2H')
ts
```

```
DatetimeIndex(['2021-01-01 00:00:00', '2021-01-01 02:00:00',
               '2021-01-01 04:00:00', '2021-01-01 06:00:00'],
              dtype='datetime64[ns]', freq='2H')
```

- Use the `DatetimeIndex` object as Pandas Series or DataFrame index

```
df = pd.DataFrame({'A':np.random.randn(len(ts)),
                   'B':np.random.randn(len(ts))},
                  index=ts)
df
```

	A	B
2021-01-01 00:00:00	0.801227	1.536040
2021-01-01 02:00:00	0.004163	2.357660
2021-01-01 04:00:00	0.043119	-2.062572
2021-01-01 06:00:00	0.302858	0.446922

Example: Pandas Time Series Data (1)

- Create a dataframe: input dataset = <*timestamp, access count*>

```
import pandas as pd

data = {'date': ['2021-01-01 08:47:05.069722',
                 '2021-01-01 18:47:05.119994',
                 '2021-01-02 08:47:05.178768',
                 '2021-01-02 13:47:05.230071',
                 '2021-01-02 18:47:05.230071',
                 '2021-01-02 23:47:05.280592',
                 '2021-01-03 08:47:05.332662',
                 '2021-01-03 18:47:05.385109',
                 '2021-01-04 08:47:05.436523',
                 '2021-01-04 18:47:05.486877'],
 'counts': [34, 25, 26, 15, 15, 14, 26, 25, 62, 41]}
df = pd.DataFrame(data)
df
```

	date	counts
0	2021-01-01 08:47:05.069722	34
1	2021-01-01 18:47:05.119994	25
2	2021-01-02 08:47:05.178768	26
3	2021-01-02 13:47:05.230071	15
4	2021-01-02 18:47:05.230071	15
5	2021-01-02 23:47:05.280592	14
6	2021-01-03 08:47:05.332662	26
7	2021-01-03 18:47:05.385109	25
8	2021-01-04 08:47:05.436523	62
9	2021-01-04 18:47:05.486877	41

Example: Pandas Time Series Data (2)

- Convert `df['date']` from string to datetime

```
df.date = pd.to_datetime(df.date)
```

- Set `df['date']` as the index

```
df = df.set_index('date')
```

	counts
date	
2021-01-01 08:47:05.069722	34
2021-01-01 18:47:05.119994	25
2021-01-02 08:47:05.178768	26
2021-01-02 13:47:05.230071	15
2021-01-02 18:47:05.230071	15
2021-01-02 23:47:05.280592	14
2021-01-03 08:47:05.332662	26
2021-01-03 18:47:05.385109	25
2021-01-04 08:47:05.436523	62
2021-01-04 18:47:05.486877	41

Accessing Time Series Data (1)

- View data in 2021

```
df['2021']
```

date	counts
2021-01-01 08:47:05.069722	34
2021-01-01 18:47:05.119994	25
2021-01-02 08:47:05.178768	26
2021-01-02 13:47:05.230071	15
2021-01-02 18:47:05.230071	15
2021-01-02 23:47:05.280592	14
2021-01-03 08:47:05.332662	26
2021-01-03 18:47:05.385109	25
2021-01-04 08:47:05.436523	62
2021-01-04 18:47:05.486877	41

- View data in January 2021

```
df['2021-01']
```

date	counts
2021-01-01 08:47:05.069722	34
2021-01-01 18:47:05.119994	25
2021-01-02 08:47:05.178768	26
2021-01-02 13:47:05.230071	15
2021-01-02 18:47:05.230071	15
2021-01-02 23:47:05.280592	14
2021-01-03 08:47:05.332662	26
2021-01-03 18:47:05.385109	25
2021-01-04 08:47:05.436523	62
2021-01-04 18:47:05.486877	41

Accessing Time Series Data (2)

- Observations after 12:00, Jan. 3, 2021

```
df['2021/1/3 12:00':]
```

date	counts
2021-01-03 18:47:05.385109	25
2021-01-04 08:47:05.436523	62
2021-01-04 18:47:05.486877	41

These are the "views"!

- Observations between Jan. 1 - 2

```
df['1/1/2021':'1/2/2021']
```

date	counts
2021-01-01 08:47:05.069722	34
2021-01-01 18:47:05.119994	25
2021-01-02 08:47:05.178768	26
2021-01-02 13:47:05.230071	15
2021-01-02 18:47:05.230071	15
2021-01-02 23:47:05.280592	14

Accessing Time Series Data (3)

- Extracting years

```
df['Year'] = df.index.year
```

date	counts	Year
2021-01-01 08:47:05.069722	34	2021
2021-01-01 18:47:05.119994	25	2021
2021-01-02 08:47:05.178768	26	2021
2021-01-02 13:47:05.230071	15	2021
2021-01-02 18:47:05.230071	15	2021
2021-01-02 23:47:05.280592	14	2021
2021-01-03 08:47:05.332662	26	2021
2021-01-03 18:47:05.385109	25	2021
2021-01-04 08:47:05.436523	62	2021
2021-01-04 18:47:05.486877	41	2021

```
.year
.month
.day
.hour
.minute
.second
.microsecond
…
```

- Extracting months

```
df['Month'] = df.index.month
```

date	counts	Month
2021-01-01 08:47:05.069722	34	1
2021-01-01 18:47:05.119994	25	1
2021-01-02 08:47:05.178768	26	1
2021-01-02 13:47:05.230071	15	1
2021-01-02 18:47:05.230071	15	1
2021-01-02 23:47:05.280592	14	1
2021-01-03 08:47:05.332662	26	1
2021-01-03 18:47:05.385109	25	1
2021-01-04 08:47:05.436523	62	1
2021-01-04 18:47:05.486877	41	1

Accessing Time Series Data (4)

- Truncate observations after Jan. 3, 2021

```
df.truncate(after='1/3/2021')
```

date	counts
2021-01-01 08:47:05.069722	34
2021-01-01 18:47:05.119994	25
2021-01-02 08:47:05.178768	26
2021-01-02 13:47:05.230071	15
2021-01-02 18:47:05.230071	15
2021-01-02 23:47:05.280592	14

- Total counts per day

```
df.resample('D').sum()
```

date	counts
2021-01-01	59
2021-01-02	70
2021-01-03	51
2021-01-04	103

Resampling

- The process of converting a time series from one frequency to another
  - Downsampling (similar to groupby operation): aggregating higher frequency data to lower frequency
  - Upsampling: converting lower frequency to higher frequency

- *df*.`resample(`*rule, axis=0, closed=None, label=None, ...*`)`
  - *rule*: the offset string or object representing target conversion
  - *axis*: axis to use for up- or down-sampling
  - *closed*: which side of bin interval is closed (default: 'right' for 'M', 'A', 'Q', 'BM', 'BA', 'BQ', and 'W', and 'left' for others)
  - *label*: which bin edge label to label bucket with (same default value as *closed*)

Downsampling

closed='left'

	counts
2021-01-01 00:00:00	0
2021-01-01 00:01:00	1
2021-01-01 00:02:00	2
2021-01-01 00:03:00	3
2021-01-01 00:04:00	4
2021-01-01 00:05:00	5
2021-01-01 00:06:00	6
2021-01-01 00:07:00	7
2021-01-01 00:08:00	8
2021-01-01 00:09:00	9
2021-01-01 00:10:00	10
2021-01-01 00:11:00	11

```
df.resample('5min').sum()
```

	counts
2021-01-01 00:00:00	10
2021-01-01 00:05:00	35
2021-01-01 00:10:00	21

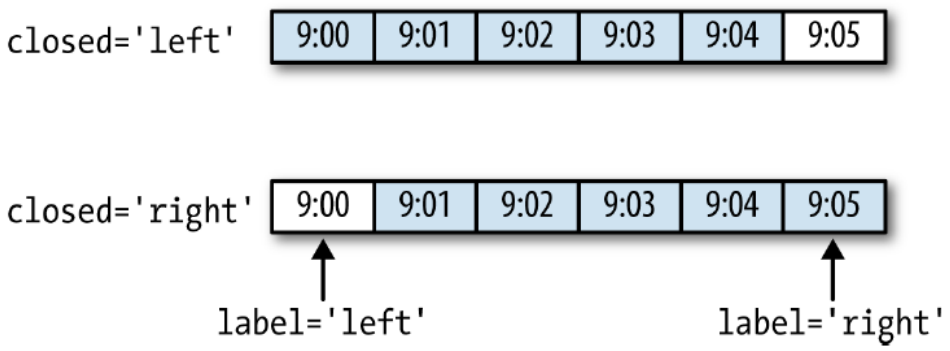
```
df.resample('5min', closed='right').sum()
```

	counts
2020-12-31 23:55:00	0
2021-01-01 00:00:00	15
2021-01-01 00:05:00	40
2021-01-01 00:10:00	11

closed='right'

```
df.resample('5min', closed='right', label='right').sum()
```

	counts
2021-01-01 00:00:00	0
2021-01-01 00:05:00	15
2021-01-01 00:10:00	40
2021-01-01 00:15:00	11



OHLC Resampling

- Open-High-Low-Close resampling
  - Used in finance

```
df.resample('5min').ohlc()
```

	counts			
	open	high	low	close
2021-01-01 00:00:00	0	4	0	4
2021-01-01 00:05:00	5	9	5	9
2021-01-01 00:10:00	10	11	10	11

	counts
2021-01-01 00:00:00	0
2021-01-01 00:01:00	1
2021-01-01 00:02:00	2
2021-01-01 00:03:00	3
2021-01-01 00:04:00	4
2021-01-01 00:05:00	5
2021-01-01 00:06:00	6
2021-01-01 00:07:00	7
2021-01-01 00:08:00	8
2021-01-01 00:09:00	9
2021-01-01 00:10:00	10
2021-01-01 00:11:00	11

Upsampling

```
df.resample('20s').ffill()
```

	counts
2021-01-01 00:00:00	0
2021-01-01 00:00:20	0
2021-01-01 00:00:40	0
2021-01-01 00:01:00	1
2021-01-01 00:01:20	1
2021-01-01 00:01:40	1
2021-01-01 00:02:00	2
2021-01-01 00:02:20	2
2021-01-01 00:02:40	2
2021-01-01 00:03:00	3

```
df.resample('20s').asfreq()
```

	counts
2021-01-01 00:00:00	0.0
2021-01-01 00:00:20	NaN
2021-01-01 00:00:40	NaN
2021-01-01 00:01:00	1.0
2021-01-01 00:01:20	NaN
2021-01-01 00:01:40	NaN
2021-01-01 00:02:00	2.0
2021-01-01 00:02:20	NaN
2021-01-01 00:02:40	NaN
2021-01-01 00:03:00	3.0

```
df.resample('20s').bfill(limit=1)
```

	counts
2021-01-01 00:00:00	0.0
2021-01-01 00:00:20	NaN
2021-01-01 00:00:40	1.0
2021-01-01 00:01:00	1.0
2021-01-01 00:01:20	NaN
2021-01-01 00:01:40	2.0
2021-01-01 00:02:00	2.0
2021-01-01 00:02:20	NaN
2021-01-01 00:02:40	3.0
2021-01-01 00:03:00	3.0

	counts
2021-01-01 00:00:00	0
2021-01-01 00:01:00	1
2021-01-01 00:02:00	2
2021-01-01 00:03:00	3

Downsampling with Periods

```
ts = pd.date_range('2021-01-01', periods=365, freq='D')
df = pd.DataFrame({'Date':np.random.randn(365)}, index=ts)
df
```

	Date
2021-01-01	0.242348
2021-01-02	-2.324153
2021-01-03	-0.829714
2021-01-04	1.253311
2021-01-05	-0.663266
...	...
2021-12-27	0.611949
2021-12-28	1.002227
2021-12-29	1.509228
2021-12-30	1.071371
2021-12-31	-1.015844

365 rows × 1 columns

```
df.resample('M').mean()
```

	Date
2021-01-31	-0.171708
2021-02-28	-0.171608
2021-03-31	-0.077201
2021-04-30	0.072524
2021-05-31	0.342635
2021-06-30	0.158424
2021-07-31	0.013654
2021-08-31	0.037618
2021-09-30	-0.080903
2021-10-31	-0.009613
2021-11-30	-0.020476
2021-12-31	0.100228

```
df.resample('M', kind='period').mean()
```

	Date
2021-01	-0.171708
2021-02	-0.171608
2021-03	-0.077201
2021-04	0.072524
2021-05	0.342635
2021-06	0.158424
2021-07	0.013654
2021-08	0.037618
2021-09	-0.080903
2021-10	-0.009613
2021-11	-0.020476
2021-12	0.100228

```
df.resample('Q').mean()
```

	Date
2021-03-31	-0.139125
2021-06-30	0.192859
2021-09-30	-0.009105
2021-12-31	0.023856

```
df.resample('Q', kind='period').mean()
```

	Date
2021Q1	-0.139125
2021Q2	0.192859
2021Q3	-0.009105
2021Q4	0.023856

Aggregation Functions

function	Description
`.asfreq()`	Return the values at the new freq, essentially a reindex
`.fillna()`	Fill missing values introduced by upsampling ('ffill', 'bfill', 'nearest')
`.ffill() / .bfill()`	Forward (or backward) fill the values
`.first() / .last()`	Compute first (or last) of group values
`.min() / .max()`	Compute min (or max) of group values
`.mean() / .median()`	Compute mean (or median) of groups, excluding missing values
`.sum()`	Compute sum of group values
`.std() / .var()`	Compute standard deviation (or variance) of groups, excluding missing values
`.ohlc()`	Compute open, high, low and close values of a group, excluding missing values
`.count()`	Compute count of group, excluding missing values
`.nunique()`	Return number of unique elements in the group
`.quantile()`	Return value at the given quantile
`.interpolate()`	Interpolate values according to different methods

Moving Window Functions: rolling()

- *df*.`rolling(`*window, min_periods=None, ...*`)`
  - Provide rolling window calculations
  - *window*: size of the moving window
  - *min_periods*: minimum number of observations in window required to have a value

```
ts = pd.date_range('1/1/2022', periods=8, freq='D')
df = pd.DataFrame({'Counts': np.arange(len(ts))}, index=ts)
df
```

	Counts
2022-01-01	0
2022-01-02	1
2022-01-03	2
2022-01-04	3
2022-01-05	4
2022-01-06	5
2022-01-07	6
2022-01-08	7

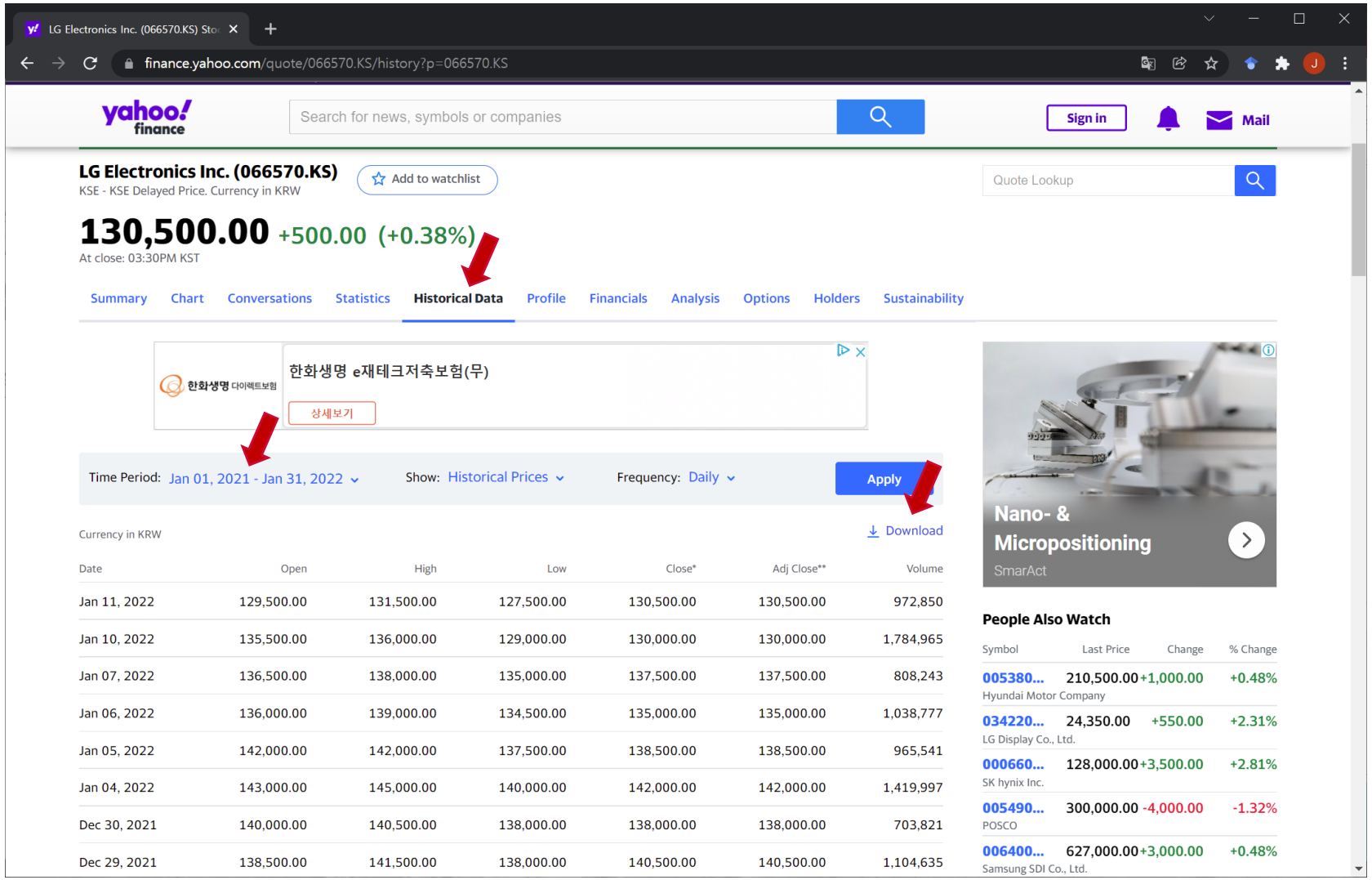
```
df.rolling(3).mean()
```

	Counts
2022-01-01	NaN
2022-01-02	NaN
2022-01-03	1.0
2022-01-04	2.0
2022-01-05	3.0
2022-01-06	4.0
2022-01-07	5.0
2022-01-08	6.0

```
df.rolling(3, min_periods=2).sum()
```

	Counts
2022-01-01	NaN
2022-01-02	1.0
2022-01-03	3.0
2022-01-04	6.0
2022-01-05	9.0
2022-01-06	12.0
2022-01-07	15.0
2022-01-08	18.0

Example: Stock Data (1)



Example: Stock Data (2)

```
lge = pd.read_csv('066570.KS.csv')
lge.head()
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	2021-01-04	136500	144000	136500	142000	142000	1868234
1	2021-01-05	139000	140000	137000	140000	140000	1602818
2	2021-01-06	148500	150500	137000	137500	137500	5283488
3	2021-01-07	139500	155000	137000	150000	150000	9782722
4	2021-01-08	145000	151000	140500	147500	147500	9472733

Example: Stock Data (3)

```
lge.tail()
```

	Date	Open	High	Low	Close	Adj Close	Volume
249	2022-01-05	142000	142000	137500	138500	138500	965541
250	2022-01-06	136000	139000	134500	135000	135000	1038777
251	2022-01-07	136500	138000	135000	137500	137500	808243
252	2022-01-10	135500	136000	129000	130000	130000	1784965
253	2022-01-11	129500	131500	127500	130500	130500	972850

Example: Stock Data (4)

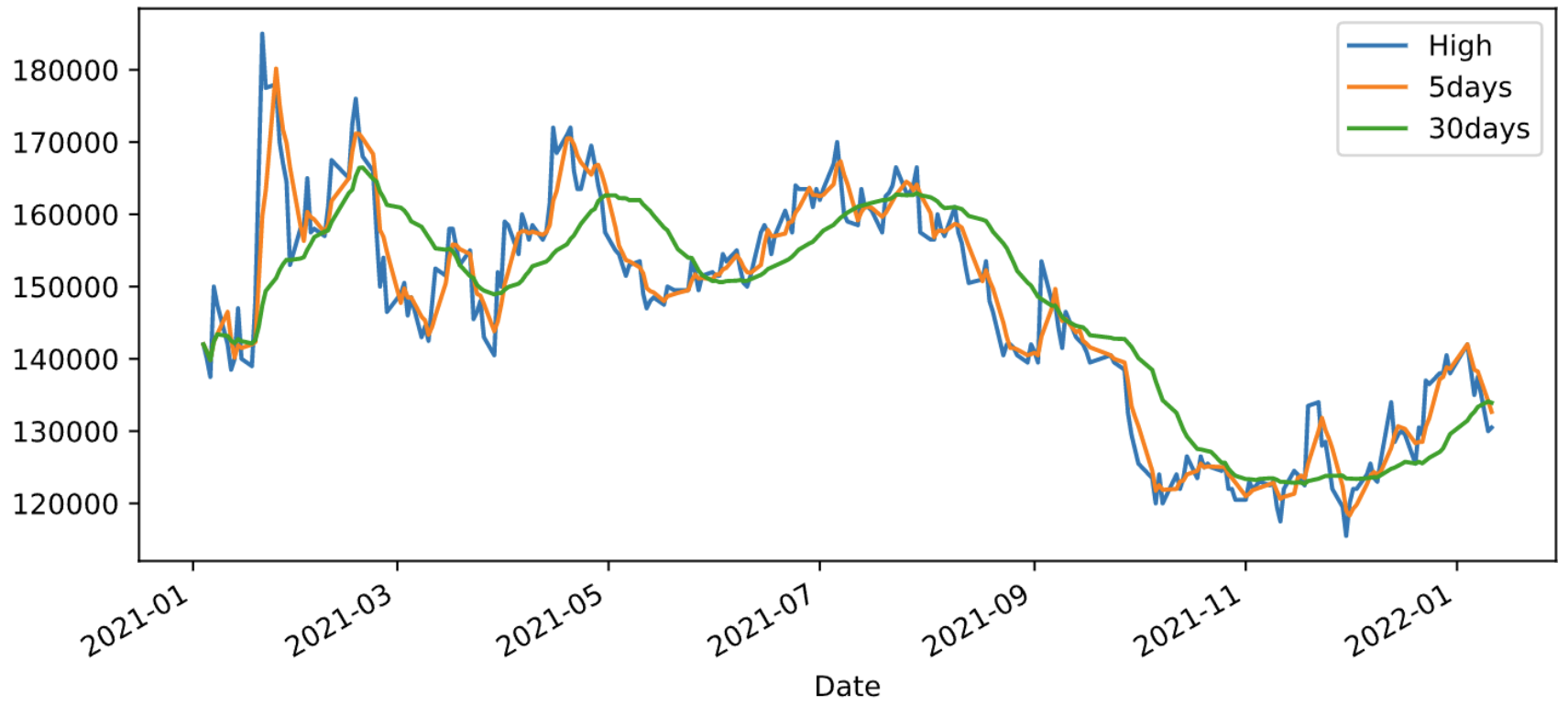
```
lge.Date = pd.to_datetime(lge.Date)
lge = lge.set_index('Date')
lge.describe()
```

	Open	High	Low	Close	Adj Close	Volume
count	254.000000	254.000000	254.000000	254.000000	254.000000	2.540000e+02
mean	147175.196850	149761.811024	144442.913386	146814.960630	146814.960630	1.620855e+06
std	15444.463011	15863.404024	14941.773980	15345.082403	15345.082403	1.693077e+06
min	116500.000000	119500.000000	115000.000000	115500.000000	115500.000000	3.600960e+05
25%	136500.000000	138625.000000	134125.000000	136625.000000	136625.000000	7.648182e+05
50%	149750.000000	152250.000000	147250.000000	149500.000000	149500.000000	1.108556e+06
75%	159000.000000	162000.000000	156500.000000	158000.000000	158000.000000	1.821710e+06
max	183000.000000	193000.000000	177500.000000	185000.000000	185000.000000	1.630495e+07

Example: Stock Data (5)

```
lge.Close.plot(figsize=(9,4))
lge.Close.rolling('5d').mean().plot()
lge.Close.rolling('30d').mean().plot()
plt.legend(['High', '5days', '30days'])
```

```
<matplotlib.legend.Legend at 0x21f3927f3a0>
```



Moving Window Functions: ewm()

- *df*.ewm(*alpha=None, min_periods=0, adjust=True, ...*)
  - Provide exponential weighted (EW) functions
  - *alpha*: specify smoothing factor α directly, $0 < \alpha \leq 1$
  - When *adjust*=True (default), the weighted moving average is calculated with $w_i = (1-\alpha)^i$:

$$y_t = \frac{\sum_{i=0}^{t} w_i x_{t-1}}{\sum_{i=0}^{t} w_i} = \frac{x_t + (1-\alpha)x_{t-1} + (1-\alpha)^2 x_{t-2} + \cdots + (1-\alpha)^t x_0}{1 + (1-\alpha) + (1-\alpha)^2 + \cdots + (1-\alpha)^t}$$

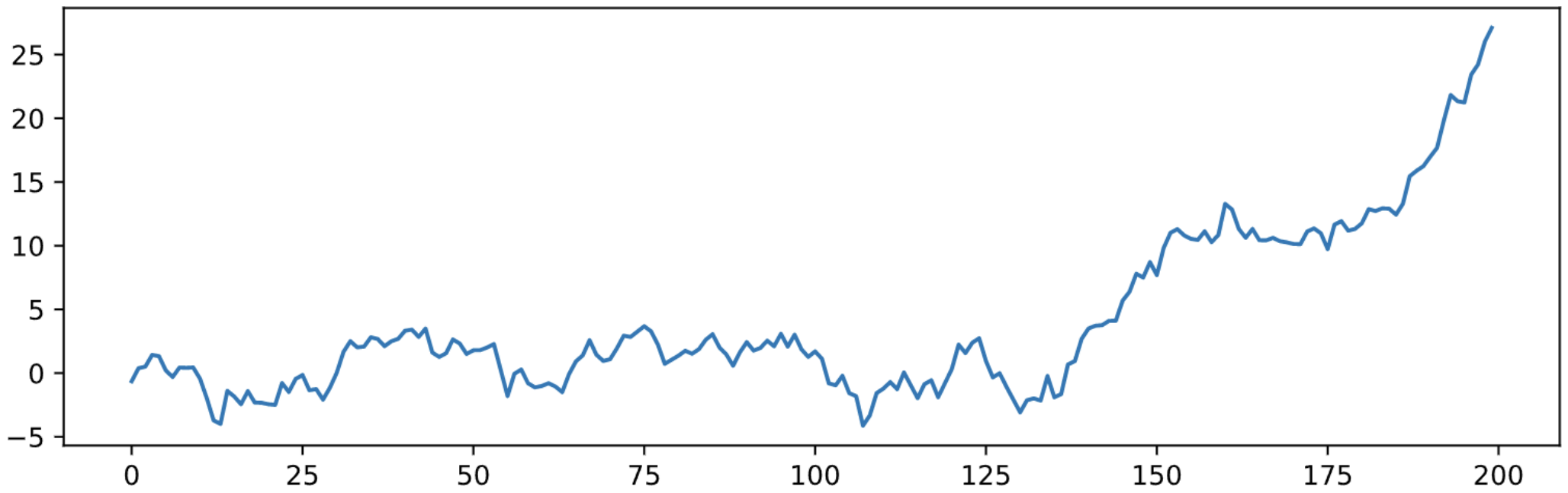
  - When *adjust*=False, the weighted moving average is calculated recursively:

$$y_0 = x_0$$
$$y_t = (1-\alpha)y_{t-1} + \alpha x_t$$

emw(): Example (1)

```
df = pd.DataFrame({'step': np.random.randn(200)})
df['dist'] = df['step'].cumsum()
df['dist'].plot(figsize=(10,3))
```

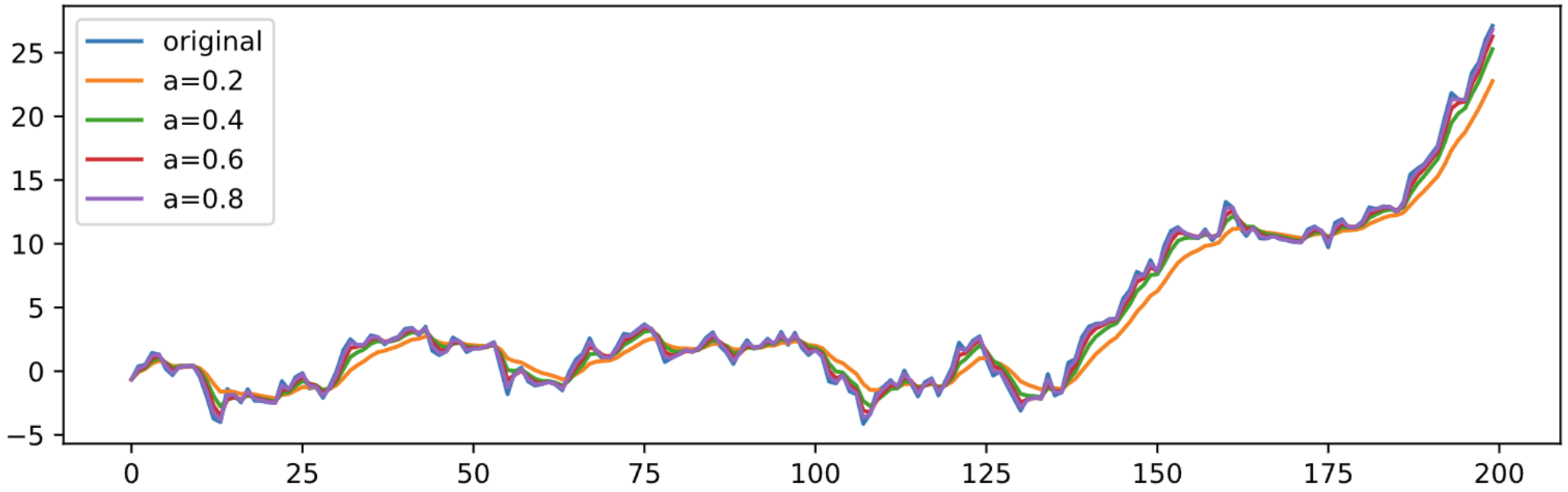
```
<AxesSubplot:>
```



emw(): Example (2)

```
df['dist'].plot(figsize=(10,3))
df['dist'].ewm(alpha=0.2).mean().plot()
df['dist'].ewm(alpha=0.4).mean().plot()
df['dist'].ewm(alpha=0.6).mean().plot()
df['dist'].ewm(alpha=0.8).mean().plot()
plt.legend(['original', 'a=0.2', 'a=0.4', 'a=0.6', 'a=0.8'])
```

```
<matplotlib.legend.Legend at 0x21f391691c0>
```



emw(): Example (3)

```
df['dist'][:10].plot(figsize=(10,3))
df['dist'].ewm(alpha=0.2).mean()[:10].plot()
df['dist'].ewm(alpha=0.4).mean()[:10].plot()
df['dist'].ewm(alpha=0.6).mean()[:10].plot()
df['dist'].ewm(alpha=0.8).mean()[:10].plot()
plt.legend(['original', 'a=0.2', 'a=0.4', 'a=0.6', 'a=0.8'])
```

```
<matplotlib.legend.Legend at 0x21f39220100>
```