

Get the average weekly market value of Bitcoin Cash.

```
data.loc[data.symbol == 'BCH', 'market'].resample('W').mean().head()
'Output':
date
2017-07-23    0.000000e+00
2017-07-30    0.000000e+00
2017-08-06    3.852961e+09
2017-08-13    4.982661e+09
2017-08-20    7.355117e+09
Freq: W-SUN, Name: market, dtype: float64
```

Convert to Datetime Datatype Using ‘to_datetime’

Pandas uses the **to_datetime** method to convert strings to Pandas datetime datatype.

The **to_datetime** method is smart enough to infer a **datetime** representation from a string of dates passed with different formats. The default output format of **to_datetime** is in the following order: **year, month, day, minute, second, millisecond, microsecond, nanosecond**.

The input to **to_datetime** is recognized as **month, day, year**. Although, it can easily be modified by setting the attributes **dayfirst** or **yearfirst** to **True**.

For example, if **dayfirst** is set to **True**, the input is recognized as **day, month, year**.

Let’s see an example of this.

```
# create list of dates
my_dates = ['Friday, May 11, 2018', '11/5/2018', '11-5-2018', '5/11/2018',
            '2018.5.11']
pd.to_datetime(my_dates)
'Output':
DatetimeIndex(['2018-05-11', '2018-11-05', '2018-11-05', '2018-05-11',
               '2018-05-11'],
              dtype='datetime64[ns]', freq=None)
```

Let's set `dayfirst` to `True`. Observe that the first input in the string is treated as a day in the output.

```
# set dayfirst to True
pd.to_datetime('5-11-2018', dayfirst = True)
'Output':
Timestamp('2018-11-05 00:00:00')
```

The `shift()` Method

A typical step in a timeseries use case is to convert the timeseries dataset into a supervised learning framework for predicting the outcome for a given time instant. The **`shift()`** method is used to adjust a Pandas DataFrame column by shifting the observations forward or backward. If the observations are pulled backward (or lagged), **NaNs** are attached at the tail of the column. But if the values are pushed forward, the head of the column will contain **NaNs**. This step is important for adjusting the **target** variable of a dataset to predict outcomes *n*-days or steps or instances into the future. Let's see some examples.

Subset columns for the observations related to Bitcoin Cash.

```
# subset a few columns
data_subset_BCH = data.loc[data.symbol == 'BCH',
['open', 'high', 'low', 'close']]
data_subset_BCH.head()
'Output':
```

| | open | high | low | close |
|------------|--------|--------|--------|--------|
| date | | | | |
| 2017-07-23 | 555.89 | 578.97 | 411.78 | 413.06 |
| 2017-07-24 | 412.58 | 578.89 | 409.21 | 440.70 |
| 2017-07-25 | 441.35 | 541.66 | 338.09 | 406.90 |
| 2017-07-26 | 407.08 | 486.16 | 321.79 | 365.82 |
| 2017-07-27 | 417.10 | 460.97 | 367.78 | 385.48 |

Now let's create a target variable that contains the closing rates 3 days into the future.

```
data_subset_BCH['close_4_ahead'] = data_subset_BCH['close'].shift(-4)
data_subset_BCH.head()
```