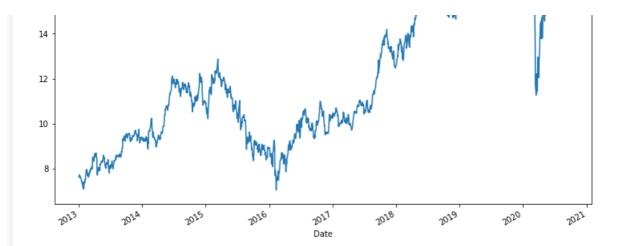
# Using Monte Carlo Simulation to predict the future stock of

```
Fortum company in Finland
In [1]:
import numpy as np
import pandas as pd
import pandas_datareader.data as web
import datetime
from scipy.stats import norm
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
C:\Users\chumj\Anaconda3\Ben\lib\site-packages\pandas_datareader\compat\__init_
FutureWarning: pandas.util.testing is deprecated. Use the functions in the public API at
pandas.testing instead.
 from pandas.util.testing import assert_frame_equal
In [2]:
start=datetime.datetime (2013,1,1)
end=datetime.datetime (2020,9,9)
We will work from the framework of Drift and Volatility.
Drift, is the best approximation of future return of a stock.
Drift=U-0.5.var or Drift=U-0.5σ^2
Expected daily returns of stock.
U=mean,var=variance.
Volatity=Random variable
Random variable=\sigma xZ(Rand(0;1))
Browian motion,
r=Drift + std*e^r
Price Today=Price Yesterday [ xe] ^{(U-0.5\sigma^{(2)})+\sigma z[Rand(0;1)]})
In [3]:
Fortum=web.DataReader('FORTUM.HE','yahoo',start,end)['Adj Close']
In [4]:
Fortum.plot(figsize=(12,8))
Out[4]:
<matplotlib.axes._subplots.AxesSubplot at 0x1c874350388>
```

20 18 16



## In [5]:

```
log_ret=np.log(Fortum/Fortum.shift(1))
```

## In [6]:

```
log_ret.head(4)
```

#### Out[6]:

```
Date
2013-01-02 NaN
2013-01-03 0.006248
2013-01-04 0.008956
2013-01-07 -0.008264
Name: Adj Close, dtype: float64
```

## In [7]:

```
U=log_ret.mean()
U
```

#### Out[7]:

0.00042350135239911456

#### In [8]:

```
var=log_ret.var()
var
```

#### Out[8]:

0.00023686545496048953

## In [9]:

```
std=log_ret.std()
std
```

## Out[9]:

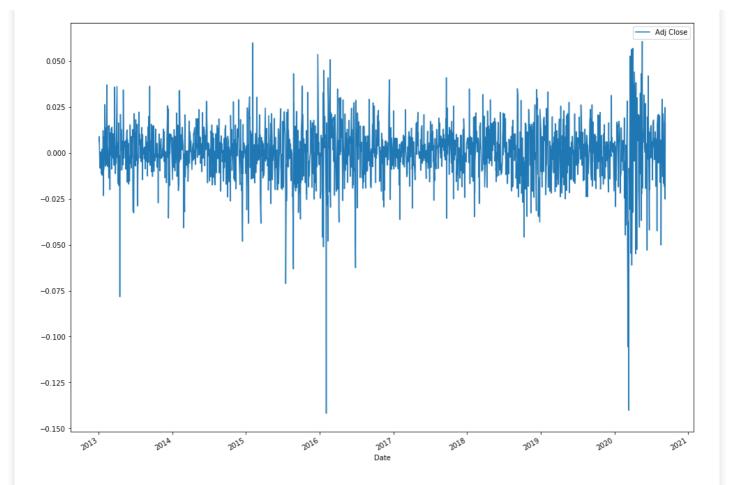
0.01539043387824039

## In [10]:

```
log_ret.plot(figsize=(16,12),legend=True)
```

## Out[10]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1c87462e408>

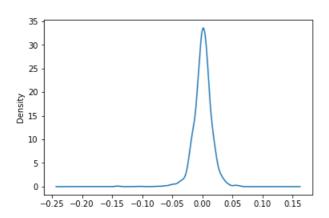


## In [11]:

```
log_ret.plot(kind='kde')
```

## Out[11]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1c87483b488>



### In [12]:

```
drift=U-(0.5*var)
drift
```

## Out[12]:

0.0003050686249188698

#### In [13]:

```
drif=np.array(drift)
drif
```

```
Out[13]:
array(0.00030507)
In [ ]:
In [14]:
sd=np.array(std)
sd
Out[14]:
array(0.01539043)
In [15]:
norm.ppf(0.95)
Out[15]:
1.6448536269514722
In [16]:
x=np.random.rand(10,2)
Out[16]:
array([[0.44018166, 0.8172463],
        [0.0983318 , 0.35456285],
        [0.64422325, 0.72474606],
        [0.10640619, 0.02144597],
[0.69115401, 0.58837673],
[0.83468723, 0.15790488],
        [0.24782871, 0.71538995],
        [0.42095991, 0.91793699],
        [0.62129818, 0.04552043],
        [0.65544681, 0.68556365]])
In [17]:
norm.ppf(x)
Out[17]:
array([[-0.15050867, 0.9049207],
        [-1.29111561, -0.37303057],
        [ 0.36977049, 0.59699926],
[-1.24586929, -2.02476074],
        [ 0.49912407, 0.22337129], [ 0.97285467, -1.00310594], [-0.68133836, 0.56920047],
        [-0.1994384 , 1.39132788],
        [ 0.30889205, -1.68993255],
        [ 0.40006807, 0.48331414]])
In [18]:
z=norm.ppf(np.random.rand(10,2))
Z
Out[18]:
array([[-0.80131775, -0.76574802],
        [-0.44878055, -2.32206744],
        [ 2.11363954, 0.90281707],
```

```
[ 1.33998112, 0.49801245],
       [-0.24215659, -1.70697299],
       [-1.57413964, 1.40093314],
[-0.49101149, 1.44116878],
       [ 2.1289281 , 0.66584854],
       [ 0.5418879 , 0.53453952], [ 1.38588859, -1.33839192]])
In [19]:
t intervals=350
iterations=10
In [20]:
\verb|daily_returns=np.exp(drif+sd*norm.ppf(np.random.rand(t_intervals,iterations))||
daily_returns
Out[20]:
array([[0.97927315, 0.9853733 , 1.01275675, ..., 0.99333584, 0.99578463,
        0.99786791],
       [1.0042481 , 1.01263921, 1.00920426, ..., 0.98258402, 1.01758068,
        0.99853448],
       [0.98129996, 1.00176591, 1.01276718, ..., 0.99267426, 0.9937563,
        0.98282945],
       [1.01374625, 1.00665872, 0.97330718, ..., 0.98760873, 0.96878194,
        0.98438723],
       [1.01309399, 1.0369216 , 0.9733088 , ..., 0.98951514, 1.01340701,
        1.02443418],
       [0.98014154, 0.97437733, 0.99106171, ..., 1.02068582, 0.99987536,
        1.03569693]])
In [21]:
so=Fortum.iloc[-1]
so
Out[21]:
17.21500015258789
In [22]:
price list=np.zeros like(daily returns)
price_list
Out[22]:
array([[0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]])
In [23]:
price list[0]=so
price_list
Out[23]:
array([[17.21500015, 17.21500015, 17.21500015, ..., 17.21500015,
        17.21500015, 17.21500015],
                 , 0.
       [ 0.
                                  , 0.
                                               , ..., 0.
         0.
                       0.
                                  ],
                    , 0.
                                      0.
                                                , ..., 0.
        [ 0.
                    , 0.
                                  ],
```

```
, 0.
                        0.
[ 0.
 0.
            0.
                      1,
[ 0.
            0.
                         0.
          , 0.
 0.
                      ],
          , 0.
[ 0.
                      , 0.
                                   , ..., 0.
          , 0.
 0.
                      ]])
```

#### In [24]:

```
for t in range(1,t_intervals):
    price_list[t]=price_list[t-1]*daily_returns[t]
```

#### In [25]:

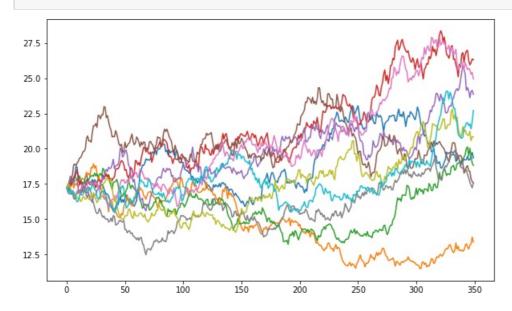
```
price_list
```

#### Out[25]:

```
array([[17.21500015, 17.21500015, 17.21500015, ..., 17.21500015, 17.21500015, 17.21500015], [17.28813127, 17.43258424, 17.37345149, ..., 16.9151841, 17.51765148, 17.1897713], [16.96484252, 17.46336863, 17.59526153, ..., 16.79126782, 17.40827652, 16.89461341], ..., [19.47535882, 13.2362082, 19.49643457, ..., 17.40520395, 20.56946346, 21.4040791], [19.730369, 13.7249102, 18.97605129, ..., 17.22271286, 20.8452385, 21.92707022], [19.33855425, 13.37324134, 18.8064379, ..., 17.57897873, 20.84264042, 22.70979938]])
```

#### In [26]:

```
plt.figure(figsize=(10,6))
plt.plot(price_list);
```



#### In [ ]: