```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from functools import partial
from scipy.stats import norm
from scipy.optimize import brentq, fsolve, minimize_scalar, curve_fit
```

## ▼ Bachelier (1990), Black (1976), QNM (2023) classes

Bachelier (1990)

$$\begin{split} dF_t &= \sigma_a dW_t \\ C_{BC}(F,K,\sigma_a,r,\tau) &= e^{-r\tau} \left[ (F-K)N(m_a) + \sigma_a \sqrt{\tau} n(m_a) \right] \\ P_{BC}(F,K,\sigma_a,r,\tau) &= e^{-r\tau} \left[ (K-F)(1-N(m_a)) + \sigma_a \sqrt{\tau} n(m_a) \right] \\ \text{where} \\ m_a &= \frac{F-K}{\sigma_a \sqrt{\tau}} \end{split}$$

Black (1976)

$$\begin{split} C_{BL}\left(F,K,\sigma_{G},r,\tau\right) &= e^{-r\tau}\left[FN(m_{G}+\frac{\sigma_{G}\sqrt{\tau}}{2})-KN(m_{G}-\frac{\sigma_{G}\sqrt{\tau}}{2})\right] \\ P_{BL}\left(F,K,\sigma_{G},r,\tau\right) &= e^{-r\tau}\left[K(1-N(m_{G}-\frac{\sigma_{G}\sqrt{\tau}}{2}))-F(1-N(m_{G}+\frac{\sigma_{G}\sqrt{\tau}}{2}))\right] \end{split}$$

where 
$$m_G = \frac{ln(F - K)}{\sigma_G \sqrt{\tau}}$$

 $dF_t = \sigma_G F_t dW_t$ 

Quadratic Normal Model (Bouchouev, 2023)

$$\begin{split} dF_t &= \sigma(F_t) dW_t = (\sigma_{ATM} + a + bF_t + cF_t^2) dW_t \\ C(F, K, a, b, c, r, \tau) &= C_{BC}(F, K, \sigma_a = \sigma_{ATM}, r, \tau) + e^{-r\tau} U \\ P(F, K, a, b, c, r, \tau) &= P_{BC}(F, K, \sigma_a = \sigma_{ATM}, r, \tau) + e^{-r\tau} U \end{split}$$

where

$$U = \sqrt{\tau} n \left( \frac{F - K}{\sigma_{ATM} \sqrt{\tau}} \right) \left[ a + \frac{b}{2} (F + K) + \frac{c}{3} (F^2 + FK + K^2 + \frac{\sigma_{ATM}^2 \tau}{2}) \right]$$

```
# Bachelier (1900)
class Bachelier:
    def __init__(self, F, vol, r, tau):
      self.F = F
      self.vol = vol
      self.r = r
      self.tau = tau
    def option_pricer(self, K, vol = None, option_type = 'call'):
      Bachelier formula
      return call/put option price
      # default parameter (to compute implied vol)
      if vol == None:
          vol = self.vol
      m = (self.F - K) / (vol * self.tau**0.5)
      if option_type == 'call':
          \texttt{return np.exp(-self.r * self.tau) * ((self.F - K) * norm.cdf(m) + vol * self.tau**0.5 * norm.pdf(m))} \\
          \texttt{return np.exp(-self.r * self.tau) * ((K - self.F) * (1 - norm.cdf(m)) + vol * self.tau**0.5 * norm.pdf(m))}
# Black (1976)
class Black:
    def __init__(self, F, vol, r, tau):
      self.F = F
      self.vol = vol
      self.r = r
      self.tau = tau
```

```
def option_pricer(self, K, vol = None, option_type = 'call'):
      Black formula
      return call/put option price
       # default parameter (to compute implied vol)
      if vol == None:
        vol = self.vol
      m = np.log(self.F / K) / (vol * self.tau**0.5)
if option_type == 'call':
          return np.exp(-self.r * self.tau) * ( self.F * norm.cdf(m + 0.5*vol*self.tau**0.5) -
                                                            K * norm.cdf(m - 0.5*vol*self.tau**0.5))
      elif option_type == 'put':
           return np.exp(-self.r * self.tau) * ( K * (1 - norm.cdf(m - 0.5*vol*self.tau**0.5)) -
                                                            self.F * (1 - norm.cdf(m + 0.5*vol*self.tau**0.5)))
# Quadratic Normal Model (2023)
class ONM:
    def __init__(self, F, sig_atm, a, b, c, r, tau):
      self.F = F
      self.sig_atm = sig_atm
      self.a = a
      self.b = b
      self.c = c
      self.r = r
      self.tau = tau
    def option_pricer(self, K, option_type = 'call'):
      The method of linearization
      return call/put option price
      m = (self.F - K)/(self.sig_atm * self.tau**0.5)
       \texttt{C\_BC} = \texttt{np.exp}(-\texttt{self.r} * \texttt{self.tau}) * ((\texttt{self.F} - \texttt{K}) * \texttt{norm.cdf}(\texttt{m}) + \texttt{self.sig\_atm} * \texttt{self.tau} * \texttt{0.5} * \texttt{norm.pdf}(\texttt{m})) 
      P\_BC = np.exp(-self.r * self.tau) * ((K - self.F)*(1-norm.cdf(m)) + self.sig_atm * self.tau**0.5 * norm.pdf(m))
      U = self.tau**0.5 * norm.pdf(m) * (self.a + self.b*(self.F + K)/2 +
                              self.c*(self.F**2 + self.F*K + K**2 + 0.5*self.sig_atm**2*self.tau)/3)
      if option type == 'call':
          return C_BC + U*np.exp(-self.r * self.tau)
      elif option_type == 'put':
          return P_BC + U*np.exp(-self.r * self.tau)
```

### → IBV and INV function

```
# IBV and INV
def implied_volatility(option_price, F, K, r, tau, option_type = 'call', model = 'black', method='brent', disp=True):
       Return Implied volatility
       model: black (default), bachelier
       methods: brent (default), fsolve, minimization
    # model
    if model == 'bachelier':
       bachelier_ = Bachelier(F, 30, r, tau)
       obj_fun = lambda vol : option_price - bachelier_.option_pricer(K = K, vol = vol, option_type = option_type)
    else: # model == 'black'
       black_ = Black(F, 0.1, r, tau)
       obj fun = lambda vol : option price - black .option pricer(K = K, vol = vol, option type = option type)
    # numerical method
   if method == 'minimization':
       obj square = lambda vol : obj fun(vol)**2
        res = minimize_scalar( obj_square, bounds=(1e-15, 8), method='bounded')
        if res.success == True:
           return res.x
    elif method == 'fsolve':
       X0 = [0.1, 0.5, 1, 3] # set of initial guess points
        for x0 in X0:
           x, _, solved, _ = fsolve(obj_fun, x0, full_output=True, xtol=1e-8)
           if solved == 1:
               return x[0]
    else:
        x, r = brentq( obj_fun, a = 1e-15, b = 500, full_output = True)
        if r.converged == True:
   # display strikes with failed convergence
   if disp == True:
   print(method, K)
```

#### Parameters and market data

```
July 07 2023
```

Options on WTI Sep '23 (CLU23)

The option data is downloaded from Bartchart.com

```
# Market parameters
futures_price = 73.77
risk_free rate = 0.05235
time\_to\_maturity = 40/365
# Mount Google Drive
from google.colab import drive
drive.mount('/content/gdrive')
# Read option data (csv file) from Google Drive
filename = '/content/gdrive/MyDrive/job preparation/data/' +\
                                  'clu23-options-07-07-2023.csv'
df_raw_data = pd.read_csv(filename, header=0, index_col = 5)
df_raw_data = df_raw_data.iloc[:-1]
\# Keep OTM options with sufficiently high trading volume (>100)
list_data = []
for index_ in df_raw_data.index:
    if index_ < futures_price:</pre>
        if df_raw_data.loc[index_, 'Volume.1'] > 100:
            list_data.append((index_, df_raw_data.loc[index_, 'Last.1'], 'put'))
       if df_raw_data.loc[index_, 'Volume'] > 100:
            list_data.append((index_, df_raw_data.loc[index_, 'Last'], 'call'))
arr_data = np.array(list_data)
 df_{data} = pd.DataFrame( \{ 'price': arr_data.T[1].astype(float), 'option type': arr_data.T[2] \}, index = arr_data.T[0].astype(float) \} 
df_data.drop(index = 50.0, inplace=True)
print(df_data)
    Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive", force_remount=True).
          price option type
    54.5
           0.10
                        put
    55.0
                         put
    57.0
           0.15
    60.0
           0.26
                         put
    61.0
           0.31
                         put
    62.0
           0.38
                         put
    62.5
           0.42
                        put
                         put
     63.5
           0.51
                         put
    64.0
           0.57
                         put
    65.0
           0.69
                         put
    66.5
           0.91
                        put
     67.0
           1.00
                         put
     67.5
           1.10
                         put
    68.0
           1.21
                         put
    68.5
           1.33
                         put
          1.75
    70.0
                         put
     72.5
           2.64
                        put
     73.0
           2.86
                        put
     74.0
           3.10
                        call
    75.0
           2.61
                        call
    76.5
           2.00
                       call
     77.0
                        call
           1.82
     77.5
                       call
           1.65
     78.0
                        call
    78.5
           1.36
                        call
    79.0
           1.23
                       call
     79.5
           1.11
                        call
```

```
# Compute INV for the ATM option
atm_strike = 75
atm_option_price = df_data[df_data.index == atm_strike].iloc[0,0]
inv_atm = implied_volatility(atm_option_price, futures_price, atm_strike, risk_free_rate, time_to_maturity, model = 'bachelier')
print(f'INV for {atm_strike}-strike option is {inv_atm}')
```

INV for 75-strike option is 24.249067350166275

call call call

call

call

call

call

1.00

0.81

0.66

0.53

0.43

0.35

81.0

82.0

83.0

84.0

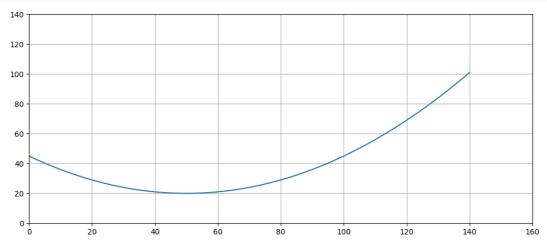
85.0

```
# calibrate parameters a, b, c on historical realized volatility
sigma_qnm = lambda f : 0.01*(f-50)**2 + 20
arr_f = np.arange(0,141,2)

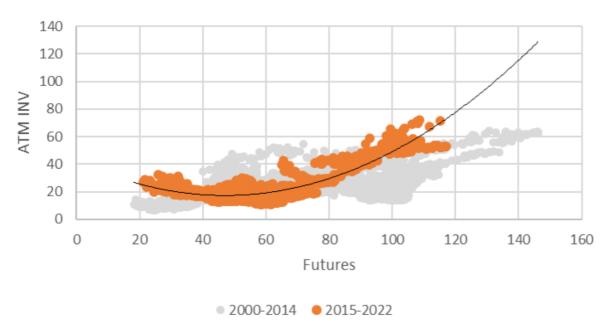
plt.figure(figsize=(12,5))
plt.plot(arr_f, sigma_qnm(arr_f))

plt.xlim(0,160)
plt.ylim(0,140)
plt.grid()

plt.show()
```



# ATM INV vs Futures



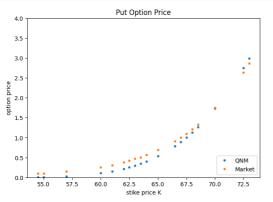
```
# Quadratic Normal Model parameters
c = 0.01
b = -50 * 2 * c
a = 20 + b**2/4/c - inv_atm
print(f'(a, b, c) = ({a}, {b}, {c})')

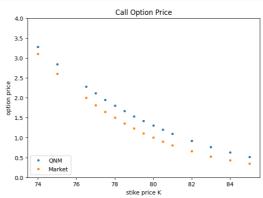
(a, b, c) = (20.750932649833725, -1.0, 0.01)
```

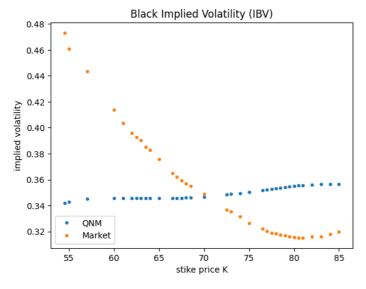
```
qnm = QNM(futures_price, inv_atm, a, b, c, risk_free_rate, time_to_maturity)
arr_qnm_call = qnm.option_pricer(K = df_data[df_data['option type']=='call'].index, option_type = 'call')
arr_mkt_call = df_data[df_data['option type']=='call'].values[:,0]
arr_qnm_put = qnm.option_pricer(K = df_data[df_data['option type']=='put'].index, option_type = 'put')
arr_mkt_put = df_data[df_data['option type']=='put'].values[:,0]

# create subplots
fig = plt.figure(figsize=(16,5))
```

```
ax1 = fig.add_subplot(121)
ax1.plot(df_data[df_data['option type']=='put'].index, arr_qnm_put, '.', linewidth = 0.5, label = 'QNM')
ax1.plot(df_data[df_data['option type']=='put'].index, arr_mkt_put, '.', linewidth = 0.5, label = 'Market')
ax1.set_ylim([0, 4])
ax1.set_title('Put Option Price')
ax1.set_xlabel('stike price K')
ax1.set_ylabel('option price')
ax1.legend(loc = 'lower right')
ax2 = fig.add subplot(122)
ax2.plot(df_data[df_data['option type']=='call'].index, arr_qnm_call, '.', linewidth = 0.5, label = 'QNM')
ax2.plot(df_data[df_data['option type']=='call'].index, arr_mkt_call, '.', linewidth = 0.5, label = 'Market')
ax2.set_ylim([0, 4])
ax2.set_title('Call Option Price')
ax2.set_xlabel('stike price K')
ax2.set_ylabel('option price')
ax2.legend(loc = 'lower left')
plt.show()
```







Calibrate QNM on market by minimizing the mean squared error

$$\theta^* = argmin_{\theta} \frac{1}{N} \sum_{i=1}^{N} \left( P_i(K_i) - f(K_i|\Theta) \right)^2$$

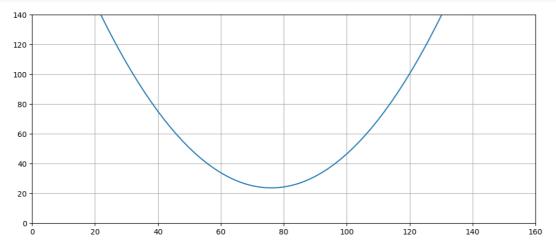
(a, b, c) = (228.66240029218525, -6.02888532189243, 0.039649083686508296)

```
# plot sig(f)
sigma_qnm = lambda f : sig_atm + params[0] + params[1]*f + params[2]*f**2
arr_f = np.arange(0,141,2)

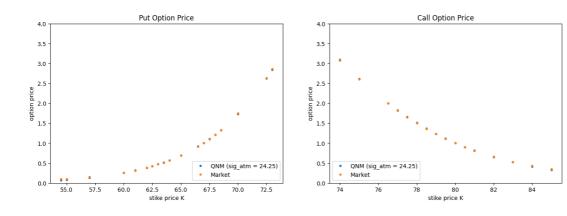
plt.figure(figsize=(12,5))
plt.plot(arr_f, sigma_qnm(arr_f), label = f'{sig_atm}')

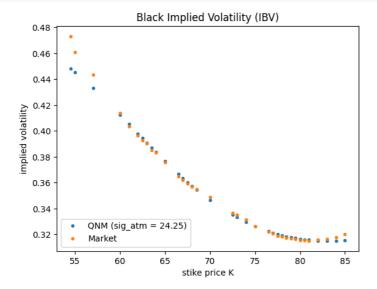
plt.xlim(0,160)
plt.ylim(0,140)
plt.grid()

plt.show()
```



```
qnm1 = QNM(futures_price, sig_atm, params[0], params[1], params[2], risk_free_rate, time_to_maturity)
arr_qnm_call = qnml.option_pricer(K = df_data[df_data['option type']=='call'].index, option_type = 'call')
arr_qnm_put = qnm1.option_pricer(K = df_data[df_data['option type']=='put'].index, option_type = 'put')
# create subplots
fig = plt.figure(figsize=(16,5))
ax1 = fig.add subplot(121)
 \texttt{ax1.plot(df\_data[df\_data['option type'] =='put'].index, arr\_qnm\_put, '.', linewidth = 0.5, label = f'QNM (sig\_atm = \{sig\_atm:.2f\})') } 
ax1.plot(df_data[df_data['option type'] == 'put'].index, arr_mkt_put, '.', linewidth = 0.5, label = 'Market')
ax1.set ylim([0, 4])
ax1.set title('Put Option Price')
ax1.set_xlabel('stike price K')
ax1.set_ylabel('option price')
ax1.legend(loc = 'lower right')
ax2 = fig.add_subplot(122)
ax2.plot(df_data[df_data['option type']=='call'].index, arr_qnm_call, '.', linewidth = 0.5, label = f'QNM (sig_atm = {sig_atm:.2f})')
ax2.plot(df_data[df_data['option type'] == 'call'].index, arr_mkt_call, '.', linewidth = 0.5, label = 'Market')
ax2.set_ylim([0, 4])
ax2.set_title('Call Option Price')
ax2.set_xlabel('stike price K')
ax2.set_ylabel('option price')
ax2.legend(loc = 'lower left')
plt.show()
```

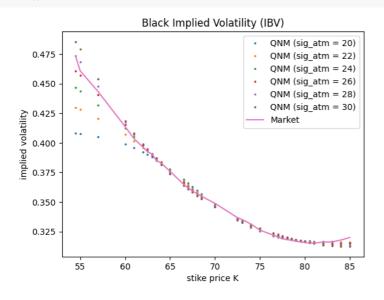




## ullet Sensitivity analysis on $\sigma_{ATM}$

```
(\text{sig\_atm, a, b, c}) = (18, 899.2461672456526, -23.960464670178027, 0.1605258599061128)
     (\text{sig atm, a, b, c}) = (20, 510.55183529656784, -13.509390005276353, 0.08997802463620516)
     (sig_atm, a, b, c) = (22, 319.97150687378837, -8.421342738904618, 0.055696649077468505)
     (\operatorname{sig\_atm},\ a,\ b,\ c)\ =\ (24,\ 234.62030273037092,\ -6.181420127339503,\ 0.04066672988463384)
     (sig_atm, a, b, c) = (26, 206.22543485841908, -5.482761595510559, 0.0360484646673681)
(sig_atm, a, b, c) = (28, 208.9825483661317, -5.626902017796557, 0.03711854668626201)
     (sig_atm, a, b, c) = (30, 228.71667552708115, -6.230245269285999, 0.041283686074145766)
      140
                                                                                                                        16
                                                                                                                        18
      120
                                                                                                                        20
                                                                                                                       22
                                                                                                                        24
      100
                                                                                                                        26
                                                                                                                        28
       80
                                                                                                                       30
                                                                                                                     -- INV ATM
       60
       40
       20
                        20
                                       40
                                                      60
                                                                     80
                                                                                   100
                                                                                                  120
                                                                                                                 140
                                                                                                                               160
arr_sig_atm = np.arange(20,32,2) # [20, 22, 24, 26, 28, 30]
for sig_atm in arr_sig_atm:
    params, _ = curve_fit(partial(qnm_option_pricer, sig_atm), df_data.index, df_data.loc[:,'price']
                         , p0=[a, b, c], bounds=([-np.inf, -np.inf, -np.inf], [np.inf, np.inf, np.inf]))
    qnm2 = QNM(futures_price, sig_atm, params[0], params[1], params[2], risk_free_rate, time_to_maturity)
    arr_qnm_call = qnm2.option_pricer(K = df_data[df_data['option type']=='call'].index, option_type = 'call')
    arr_qnm_put = qnm2.option_pricer(K = df_data[df_data['option type']=='put'].index, option_type = 'put')
    list_ibv_qnm = []
    for stirke_, option_type_, qnm_ in zip(df_data.index, df_data['option type'].values, np.concatenate((arr_qnm_put, arr_qnm_call))):
        list_ibv_qnm.append(implied_volatility(qnm_, futures_price, stirke_, risk_free_rate, time_to_maturity,
                                                              option_type = option_type_, model = 'black', method='fsolve', disp=True))
    plt.plot(df_data.index, list_ibv_qnm, '.', label = f'QNM (sig_atm = {sig_atm})', markersize = 3)
list ibv mkt = []
for stirke_, option_type_, mkt_ in zip(df_data.index, df_data['option type'].values, np.concatenate((arr_mkt_put, arr_mkt_call))):
    list_ibv_mkt.append(implied_volatility(mkt_, futures_price, stirke_, risk_free_rate, time_to_maturity,
                                                             option_type = option_type_, model = 'black', method='fsolve', disp=True))
plt.plot(df_data.index, list_ibv_mkt, '-', label = 'Market')
plt.title('Black Implied Volatility (IBV)')
plt.xlabel('stike price K')
```

 $(\operatorname{sig\_atm}, \ \mathsf{a, b, c}) = (16, \ 1667.44235861412, \ -44.685742316515785, \ 0.3005727349924079)$ 



plt.ylabel('implied volatility')
plt.legend(loc = 'upper right')

plt.show()