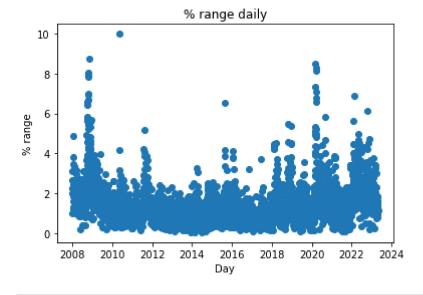
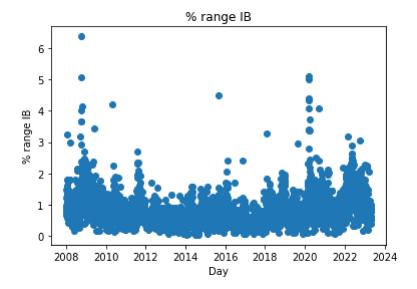
```
dat1 =pd.read_csv('NQ_1min_continuous_adjusted.txt',names = ["times","open","high","low
In [2]:
         # Define a function to extract the date and time from a datetime object
         def extract date time(datetime str):
             datetime_obj = pd.to_datetime(datetime_str)
             return pd.Series([datetime_obj.date(), datetime_obj.time()])
         # Apply the function to the 'times' column and create new columns for 'date' and 'time'
         dat1[['date', 'time']] = dat1['times'].apply(extract_date_time)
In [3]:
         #Filter to times you want for opening range (IB) and just the day range
         opening_range = dat1[(dat1.time >= datetime.time(9, 30)) & (dat1.time <= datetime.time(
         open_to_close = dat1[(dat1.time >= datetime.time(9, 30)) & (dat1.time <= datetime.time(</pre>
         # Group by 'date' and calculate the max and min for 'high' and 'low' for both opening_r
In [4]:
         opening_range_stats = opening_range.groupby('date').agg(high_ib=('high', 'max'), low_ib
         open_to_close_stats = open_to_close.groupby('date').agg(high_daily=('high', 'max'), low
         # Calculate the ranges
         opening_range_stats['opening_range_ib'] = 100 * (opening_range_stats['high_ib'] - openi
         open_to_close_stats['range_daily'] = 100 * (open_to_close_stats['high_daily'] - open_to_
         # Merge the two sets of statistics into a single DataFrame
         data_save_optimized = opening_range_stats.merge(open_to_close_stats, on='date')
         #Plot Daily range
In [5]:
         plt.plot(data_save_optimized.range_daily,'o')
         plt.xlabel('Day')
         plt.ylabel('% range')
         plt.title('% range daily')
         plt.show()
```



```
In [6]: #Plot Opening range
   plt.plot(data_save_optimized.opening_range_ib,'o')
   plt.xlabel('Day')
   plt.ylabel('% range IB')
   plt.title('% range IB')
   plt.show()
```



```
In [7]: # Calculate correlation
    x = data_save_optimized.range_daily
    y = data_save_optimized.opening_range_ib
    correlation_matrix = np.corrcoef(x, y)
    correlation = correlation_matrix[0, 1]
    print(correlation)
# 1 means they are 100% correlated so having a .78 shows they are very closely related
```

0.7853384551491553

This analysis aims to demonstrate that the Initial Balance (IB) range is a critical indicator of the daily market range, suggesting that the volatility within the first hour of trading can be predictive of the day's overall volatility. This insight is particularly useful for traders, as it assists in identifying optimal times for initiating or holding positions. By integrating this concept with the Average Daily Range (ADR), specifically a 5-day mean, traders can make more informed decisions about when to commence trading activities. Furthermore, incorporating this approach into a Brownian motion simulation enhances its realism and efficacy in mirroring actual market movements, thereby improving predictive capabilities for real-world market dynamics.