

# Seasonality in Cryptocurrency Volatility

May 20, 2021

## 1 Dataset: 1 minute OHLCV

	ETHUSDC	ETHUSDT	ETHUSD	BTCUSDT	BTCUSD	Sample Period
Kraken			O		O	Jan. 2019~now
Coinbase		O	O	O	O	Jan. 2019~now
Binance	O	O		O		Jan. 2019~now
Uniswap_v2	O					Aug, 2020~now
Sushiswap	O					Feb, 2021~now

For comparability, we choose the sample for 2019~now for centralized exchanges. We do not consider Sushiswap because the sample is too small. Pairs marked with the red circle in the above table will be dealt in this report. Volatility and volume patterns of pairs marked with the black circle show the same pattern to those with the red circle.

## 2 Measure of Volatility

We assume that daily return  $r_t$  follows a conditional normal distribution with conditional variance  $\sigma_t$  given the information set  $\mathcal{F}_{t-1}$ .

$$r_t | \mathcal{F}_{t-1} \sim N(0, \sigma_t^2)$$

We can estimate conditional variance by the simple quadratic variance. For example, daily conditional variance  $\sigma_t^2 \approx RV_t^{day} = \sum_{h=1}^{1440} \left( \frac{p_{t,h} - p_{t,h-1}}{p_{t,h-1}} \right)^2 = \sum_{h=1}^{1440} (r_{t,h})^2$  where  $p_{t,h}$  is close price and  $r_{t,h}$  is return for minute  $h$  of day  $t$ , respectively. Note that  $p_{t,h-1440k} = p_{t-k,h}$ .

## 3 Seasonality

### 3.1 Weekly Pattern:

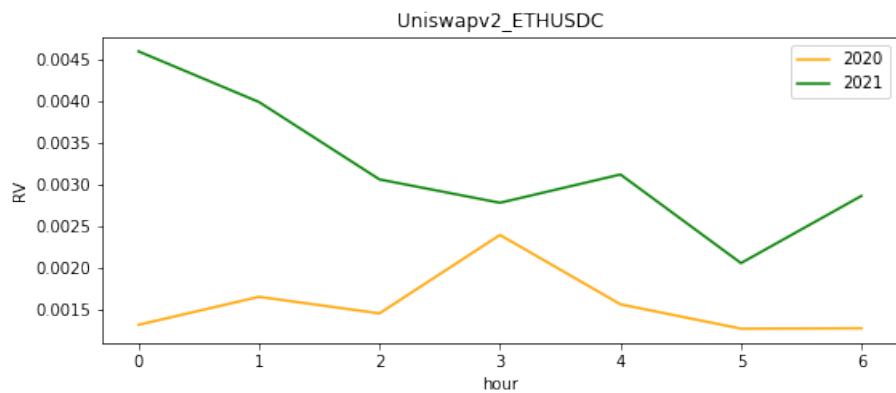
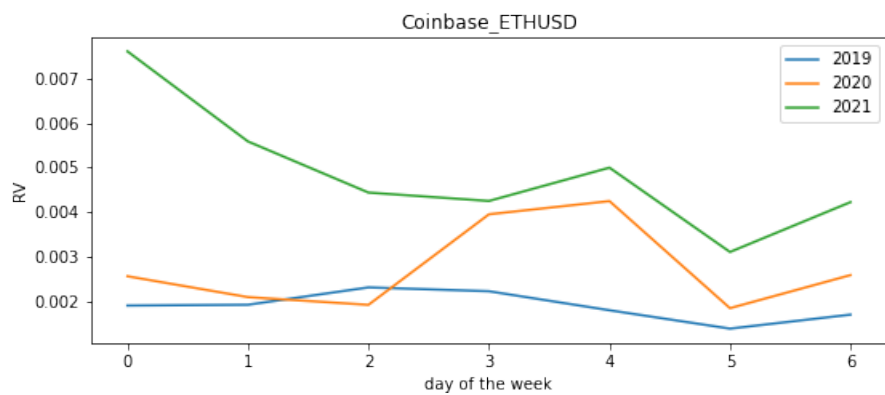
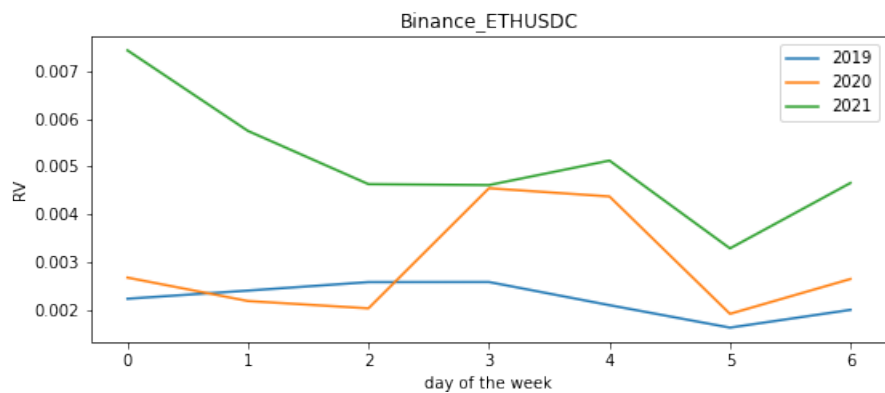
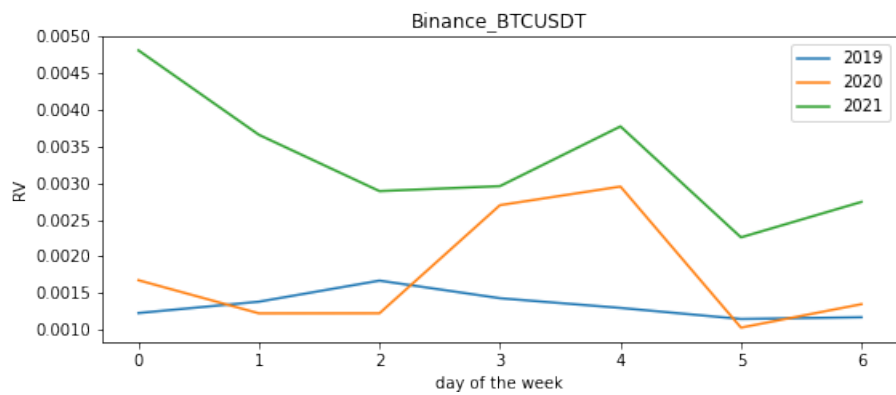
#### 3.1.1 Volatility: 1minute RV

We define the mean of the realized variance for a day of the week  $i$  given a year  $y$  as

$$\overline{RV}_{y,i}^{dow} = \frac{1}{W} \sum_{w=0}^{W-1} RV_{y,7w+i}^{day}$$

where  $y = 2019, 2020, 2021$  and  $i = 0, \dots, 6$ , 0 for Monday and 6 for Sunday.  $W$  is the number of weeks for  $y$ .  $RV_{y,7w+i}^{day}$  is a realized variance of the  $i$ -th day of the  $w$ -th week for a year  $y$ .

$\overline{RV}_{y,i}^{dow}$  is a simple average of realized variance for the  $i$ -th day of the week. The following plots show  $\overline{RV}_{y,i}^{dow}$  for each year.  $i$  is on x-axis and each line represents  $\overline{RV}_{y,i}^{dow}$  for the same  $y$ .



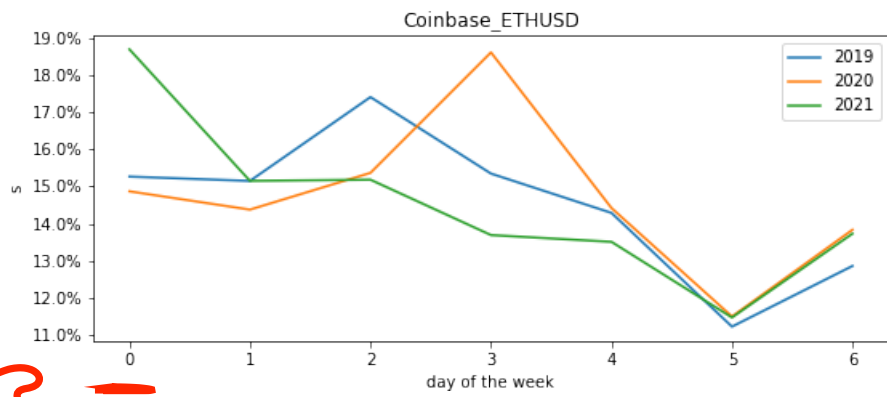
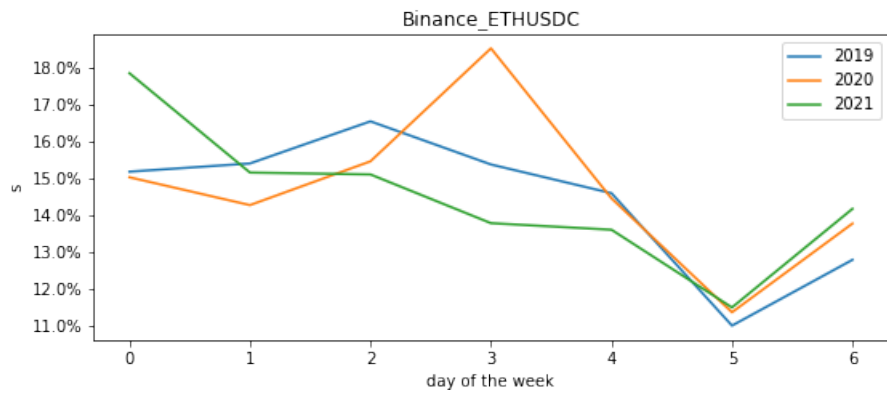
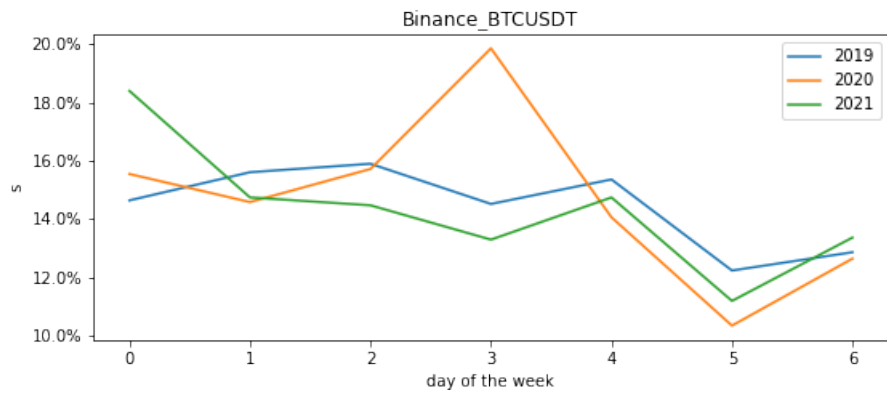
Prices in centralized exchanges show similar volatility patterns. The RV of Uniswap\_v2 shows similar patterns but the different level. That is because market microstructure noise such as bid-asks bounce or tick size impact 1-minute RV of centralized market, which does not happen in the AMM based decentralized exchanges.

The level of volatility is high in 2021 and low in the other years. This makes comparison hard. We can adjust this level by the following statistics that represents seasonality.

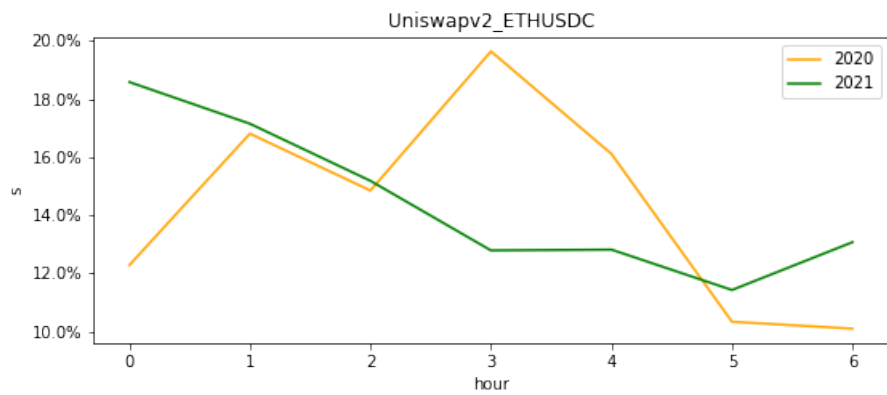
$$s_{y,i}^{dow} = \frac{100}{W} \sum_{w=0}^{W-1} \frac{RV_{y,7w+i}^{day}}{\sum_{j=0}^6 RV_{y,7w+i-j}^{day}}$$

$\sum_{j=0}^6 RV_{y,7w+i-j}^{day}$  is a realized variance for the past week, so  $s_{y,i}^{dow}$  is an average percentage of weekly realized variance that daily realized variance for the  $i$ -th day of the week accounts for. Note that  $\sum_{i=0}^6 s_{y,i}^{day} \approx 100$  since weekly realized variance is the sum of the daily realized variances. Since  $s_{y,i}^{dow}$  is bounded above by 1, it is more robust to outliers than  $\overline{RV}_{y,i}^{dow}$ .

The following plots show  $s_{y,i}^{dow}$ .  $i$  is on the x-axis.



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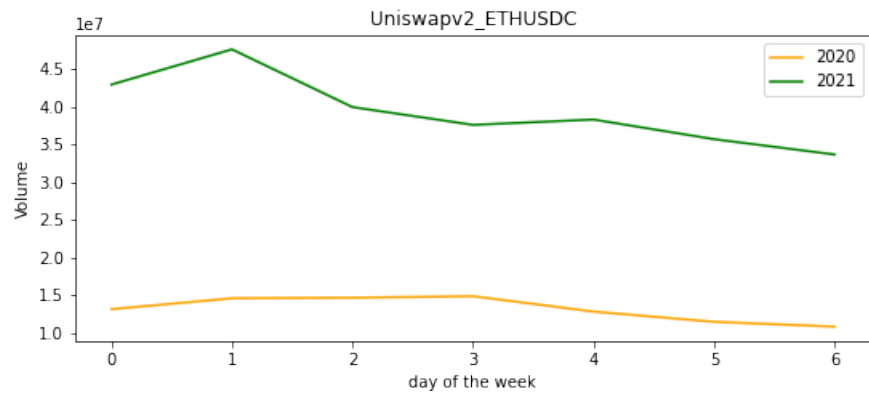
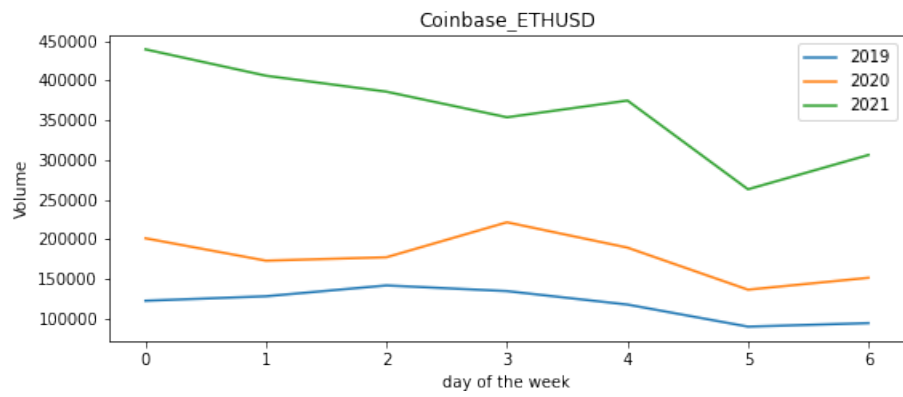
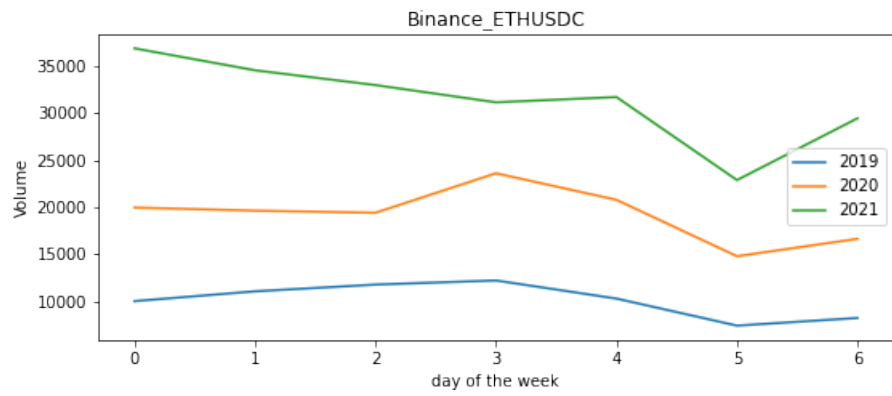
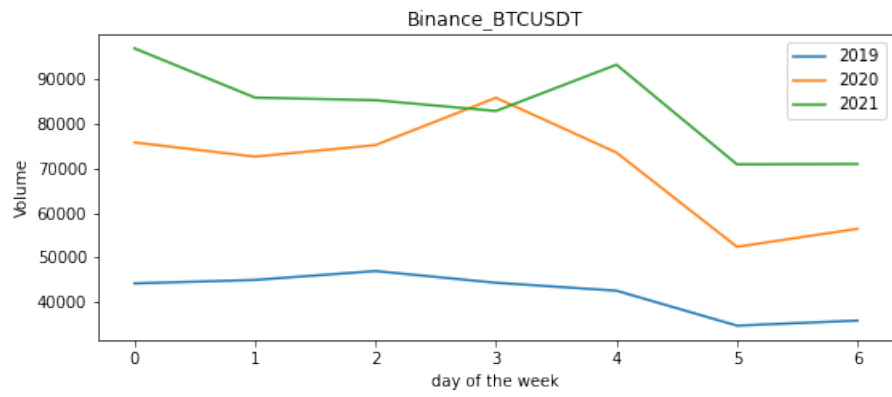
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### 3.1.2 Volume

Similar to the realized volatility, we define the mean of the volume for a day of the week  $i$  given a year  $y$  as

$$\bar{V}_{y,i}^{dow} = \frac{1}{W} \sum_{w=0}^{W-1} V_{y,7w+i}^{day}$$

where  $y = 2019, 2020, 2021$  and  $i = 0, \dots, 6$ .  $W$  is the number of weeks for  $y$ .  $V_{y,7w+i}^{day}$  is a daily volume of the  $i$ -th day of the  $w$ -th week for a year  $y$ .  $\bar{V}_{y,i}^{dow}$  is a simple average of volume for the  $i$ -th day of the week.

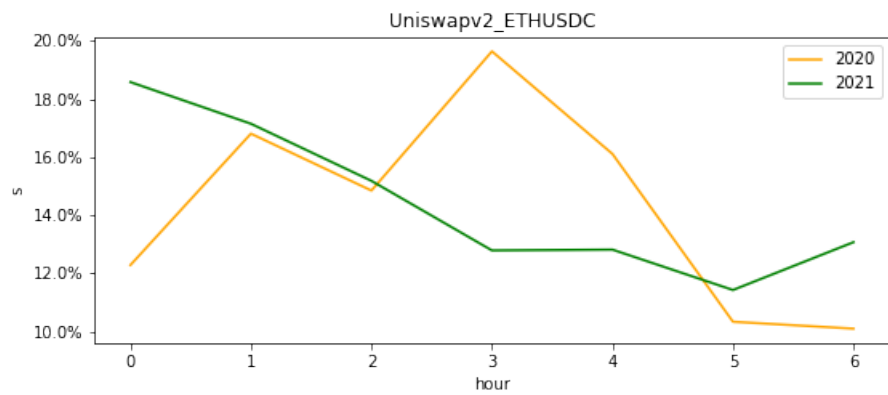
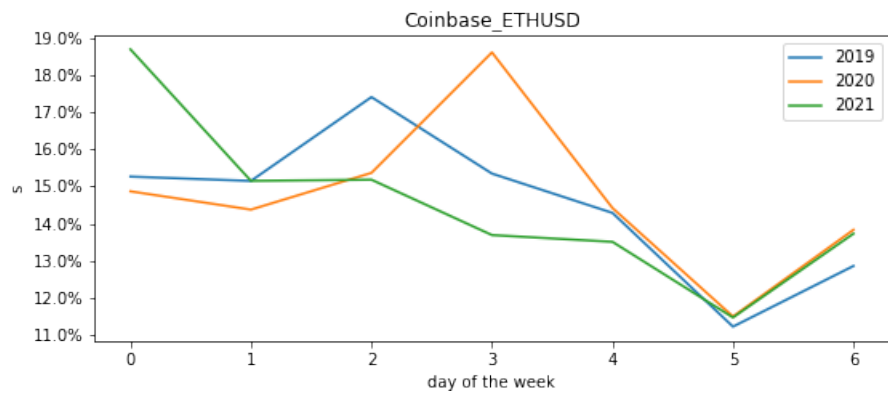
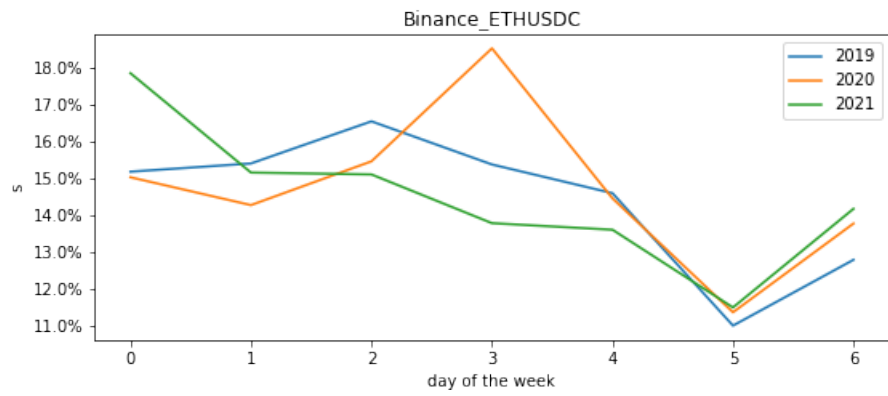
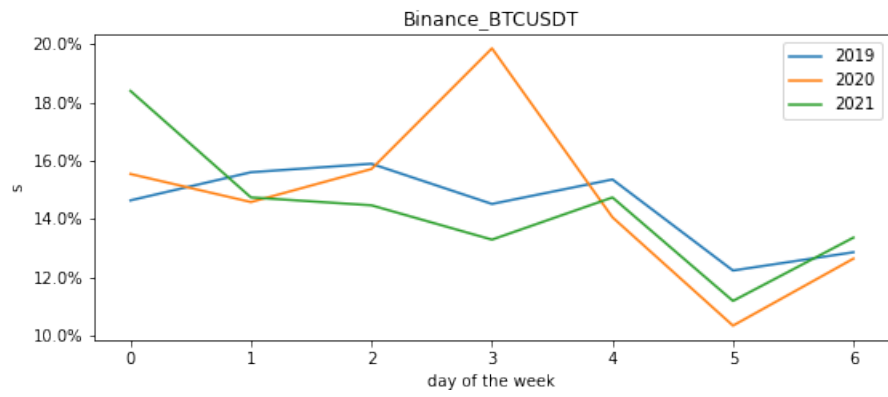


Volumes in centralized exchanges show similar patterns regardless of currency pairs. Uniswap\_v2 has been rapidly growing over the past years, so its volume patterns are affected by the growth effect as well. Similar to the volatility analysis, we introduce seasonality statistics which is more robust to outliers.

$$S_{y,i}^{dow} = \frac{100}{W} \sum_{w=0}^{W-1} \frac{V_{y,7w+i}^{day}}{\sum_{j=0}^6 V_{y,7w+i-j}^{day}}$$

$\sum_{j=0}^6 V_{y,7w+i-j}^{day}$  is the volume for the past week, so  $S_{y,i}^{dow}$  is an average percentage of weekly volume that the  $i$ -th day of the week accounts for. Note that  $\sum_{i=0}^6 S_{y,i}^{dow} \approx 100$  since weekly realized variance is the sum of the daily realized variances.

The following plots show  $S_{y,i}^{dow}$ .  $i$  is on the x-axis.





## 3.2 Intraday Pattern

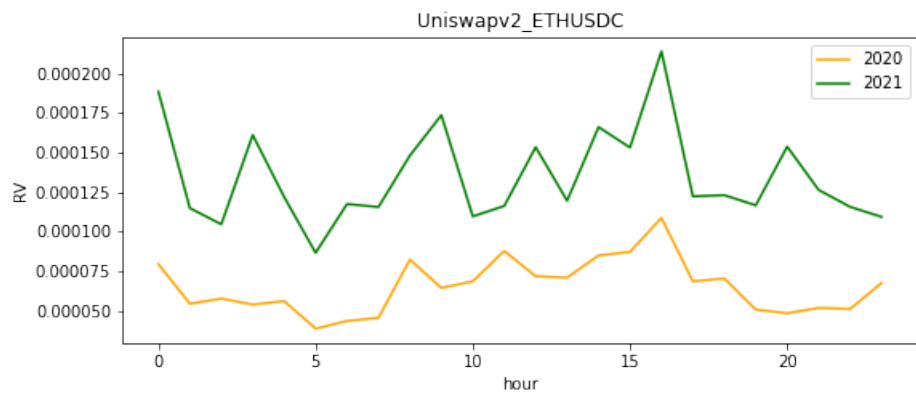
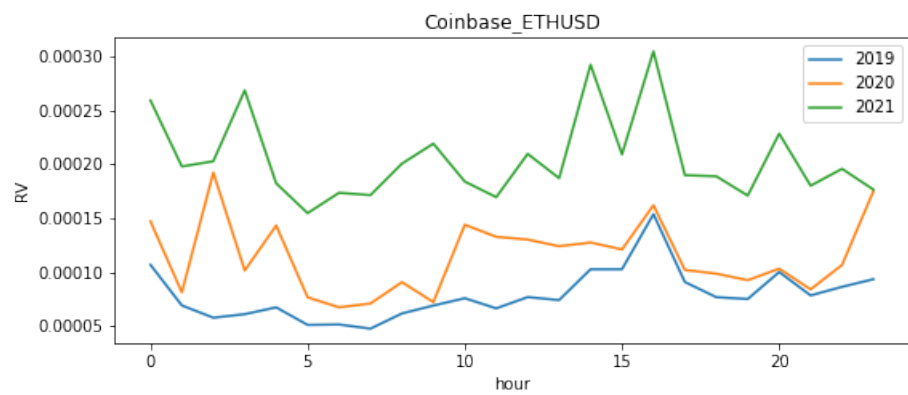
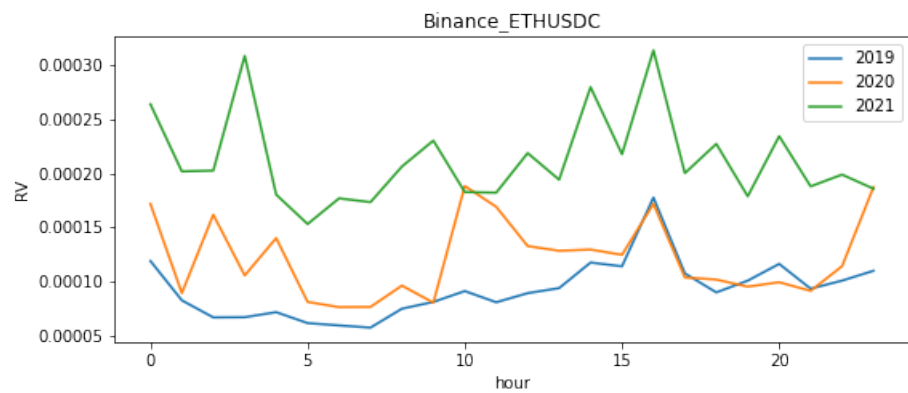
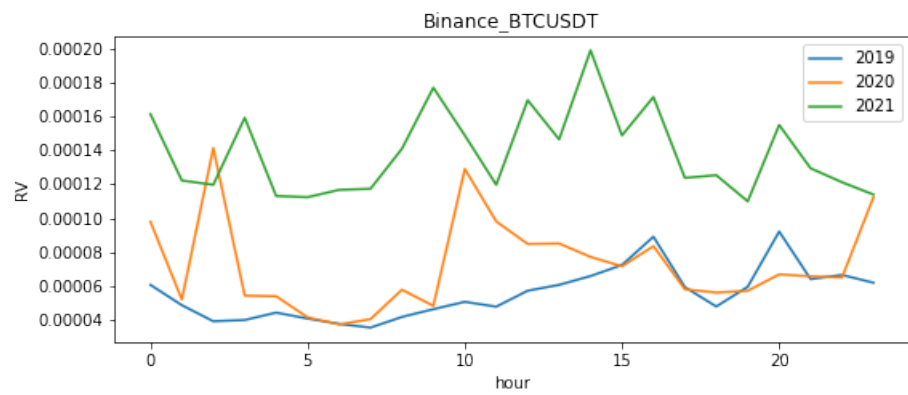
### 3.2.1 Volatility: 1minute RV

We define mean of the realized variance for a day of the week  $i$  given a year  $y$  as

$$\overline{RV}_{y,i}^{hour} = \frac{1}{D} \sum_{d=0}^{D-1} RV_{y,24d+i}^{hour}$$

where  $y = 2019, \dots, 2021$  and  $i = 0, \dots, 23$  refers the hour of the day.  $D$  is the number of hours for  $y$ .  $RV_{y,24d+i}^{hour}$  is a realized variance of the  $i$ -th hour of the  $d$ -th day for a year  $y$ .

$\overline{RV}_{y,i}^{hour}$  is a simple average of realized variance for the  $i$ -th hour of the day. The following plots show  $\overline{RV}_{y,i}^{hour}$  for each year.

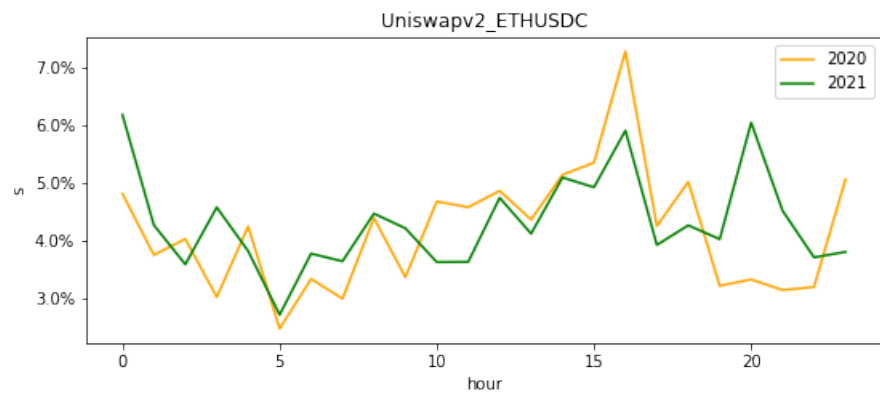
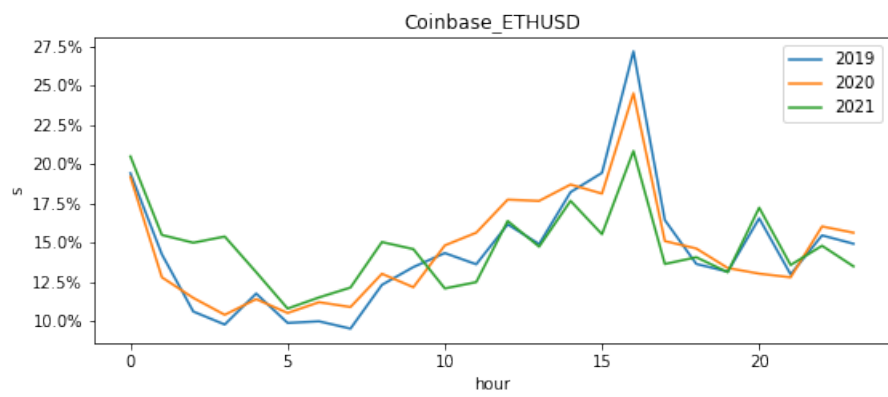
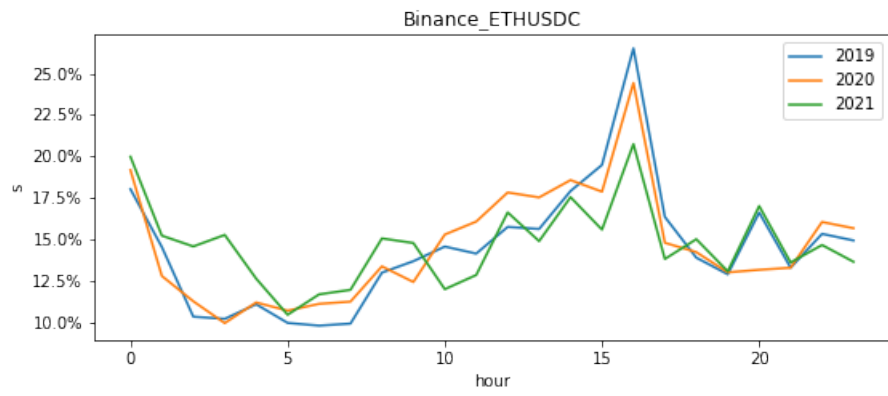
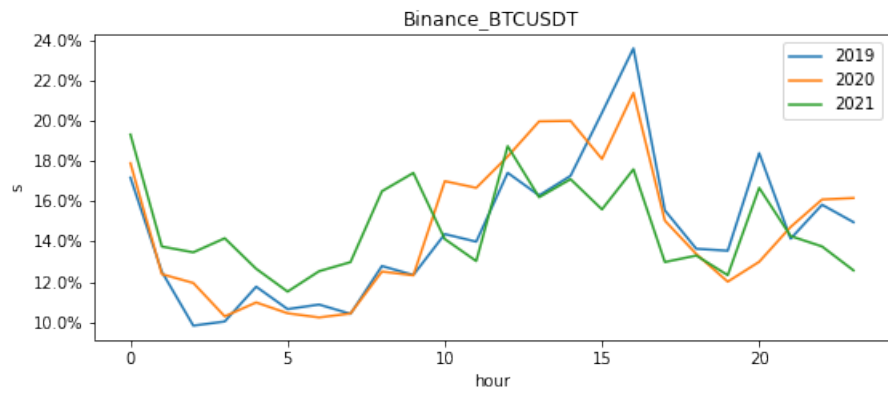


Similar to the weekly case, Uniswap and the centralized exchanges exhibit similar patterns. Seasonlity statistics for hourly volatility is

$$s_{y,i}^{hour} = \frac{100}{365} \sum_{d=0}^{365} \frac{RV_{y,24d+i}^{hour}}{\sum_{j=0}^{23} RV_{y,24d+i-j}^{hour}}$$

$\sum_{j=0}^{23} RV_{y,24d+i-j}^{hour}$  is a realized variance for the past day, so  $s_{y,i}^{hour}$  is an average percentage of daily realized variance that the  $i$ -th hour of the day accounts for. Note that  $\sum_{i=0}^{23} s_{y,i}^{id} \approx 100$  since daily realized variance is the sum of the hourly realized variances.

The following plots show  $s_{y,i}^{hour}$ .  $i$  is on the x-axis.



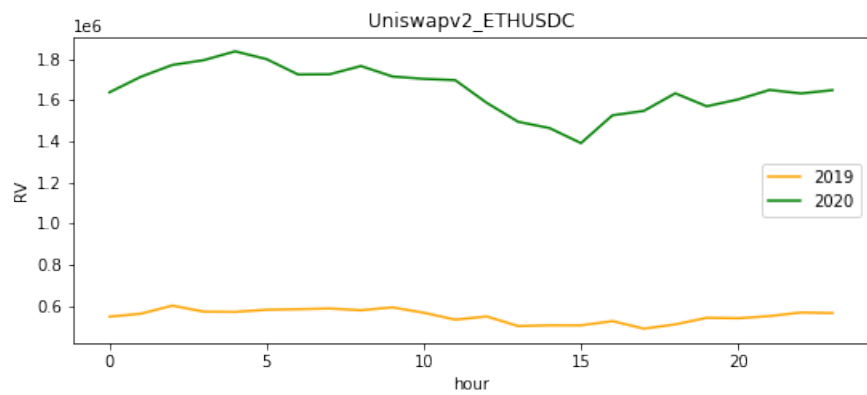
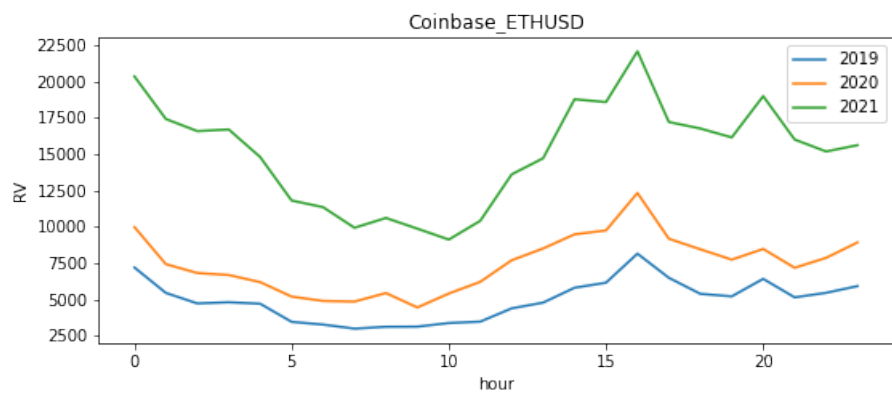
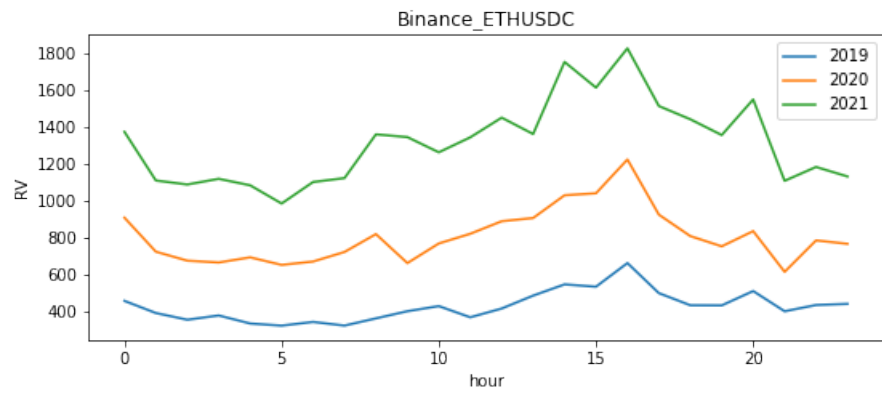
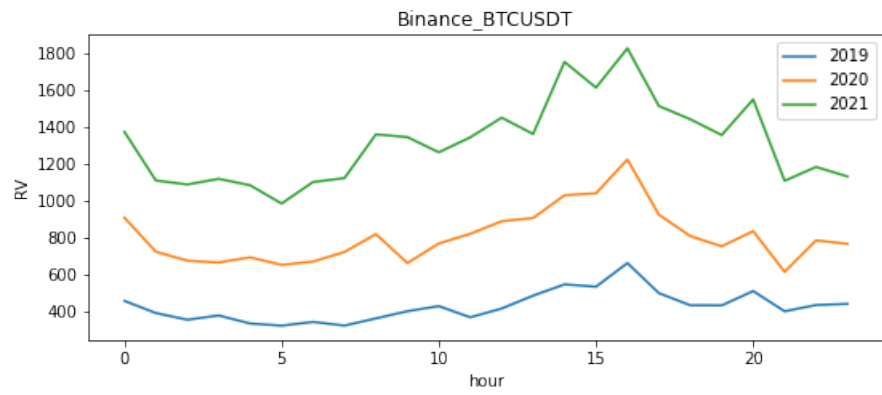
### 3.2.2 Volume

We define the mean of the volume for an hour of the day given a year  $y$  as

$$\bar{V}_{y,i}^{hour} = \frac{1}{D} \sum_{d=0}^{D-1} V_{y,24d+i}^{hour}$$

where  $y = 2019, \dots, 2021$  and  $i = 0, \dots, 23$  refers the hour of the day.  $D$  is the number of hours for  $y$ .  $V_{y,24d+i}^{hour}$  is a volume of the  $i$ -th hour of the  $d$ -th day for a year  $y$ .

$\bar{V}_{y,i}^{hour}$  is a simple average of hourly volume for the  $i$ -th hour of the day. The following plots show  $\bar{V}_{y,i}^{hour}$  for each year.  $i$  is on x-axis and each line represents  $\bar{V}_{y,i}^{hour}$  for the same  $y$ .

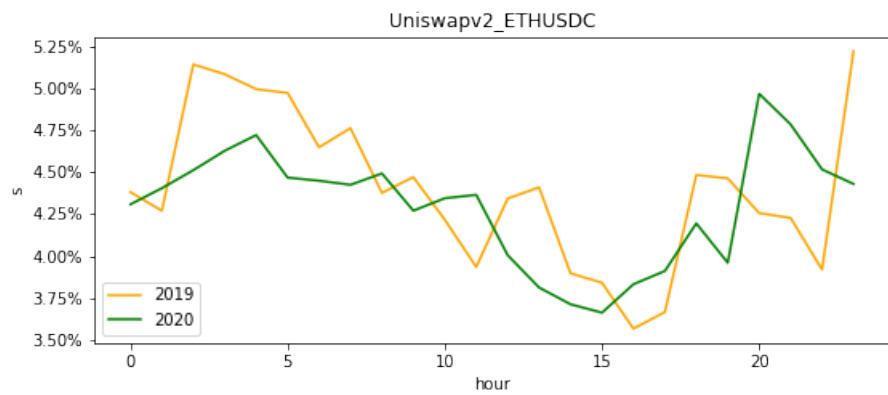
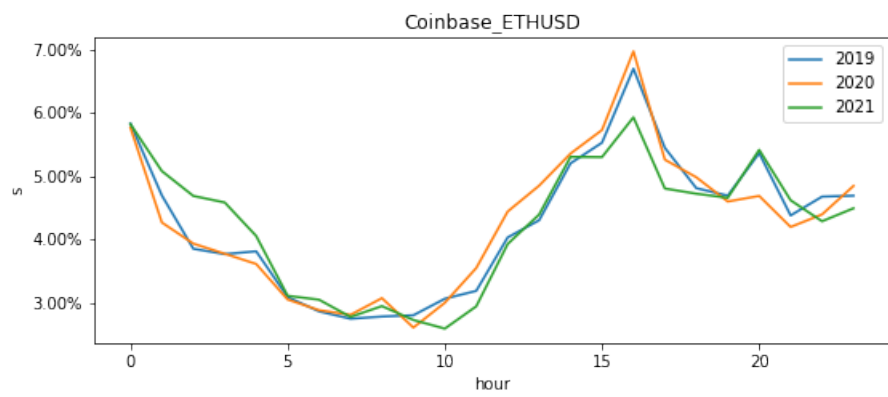
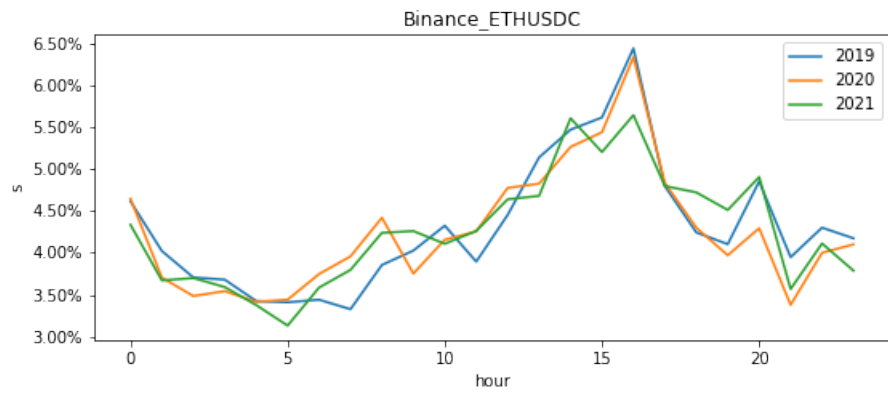
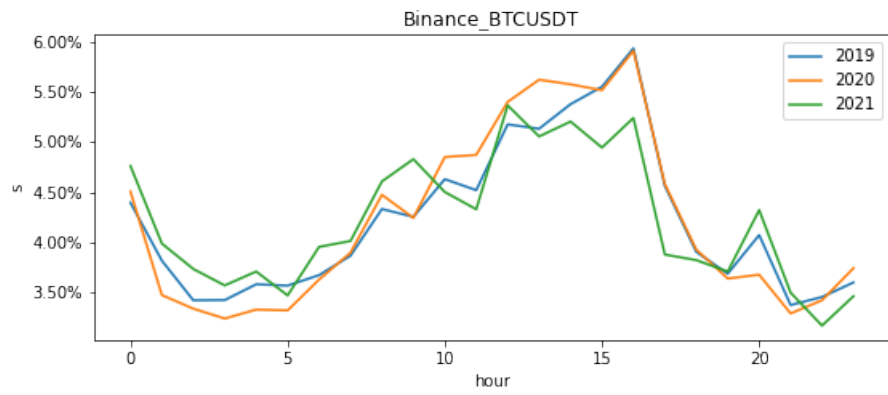


Volumes in centralized exchanges show similar patterns regardless of currency pairs. Since Uniswap\_v2 has grown exponentially over the past years, it shows very different volatility patterns compared the others and it makes comparison hard. We can reduce this effect this by the following statistics.

$$S_{y,i}^{hour} = \frac{100}{D} \sum_{d=0}^{D-1} \frac{V_{y,24d+i}^{hour}}{\sum_{j=0}^{23} V_{y,24d+i-j}^{hour}}$$

$\sum_{j=0}^{23} V_{y,24d+i-j}^{hour}$  is a realized variance for the past day, so  $S_{y,i}^{hour}$  is an average percentage of daily volume that hourly volume of the  $i$ -th hour of the day accounts for. Note that  $\sum_{i=0}^{23} S_{y,i}^{hour} \approx 100$

The following plots show  $S_{y,i}^{hour}$ .  $i$  is on the x-axis.





The volume in Uniswap moves in the opposite direction to that of the centralized exchanges. It hits local maximum around 5 and decreases over time until 15. It is still open question what causes the difference. It may be caused by geographical component. For example, Chinese traders have limited access to the main centralized exchanges, so they may choose to use Uniswap instead.

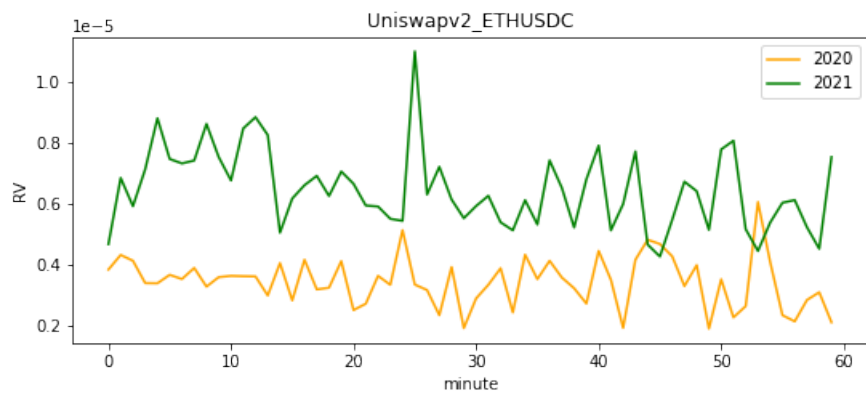
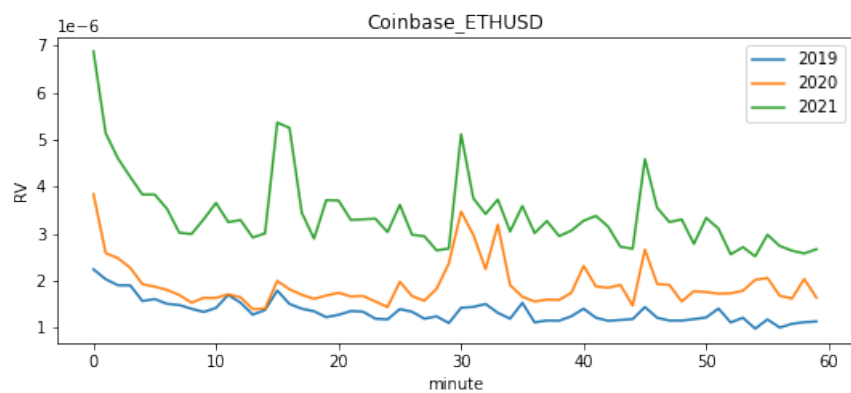
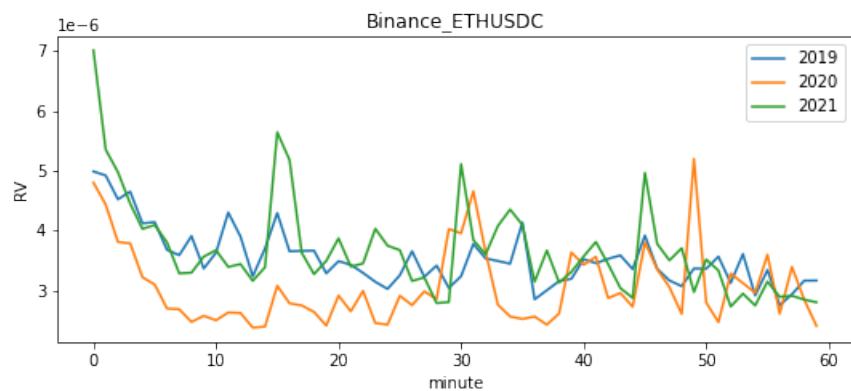
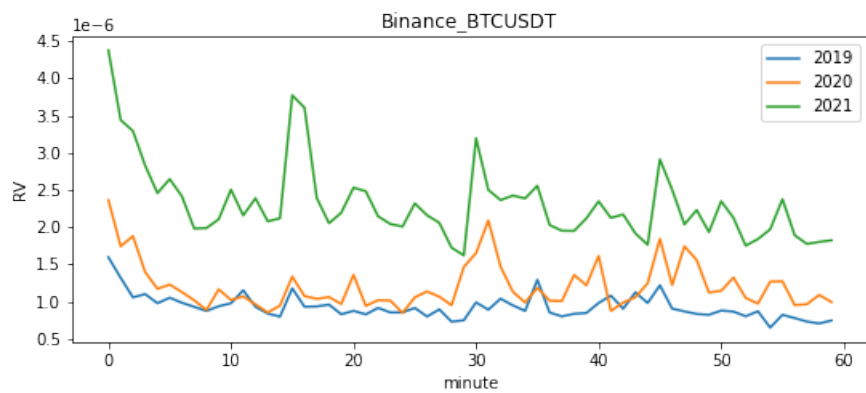
### 3.3 Intra-hour Pattern

#### 3.3.1 Volatility

Similarly, we define

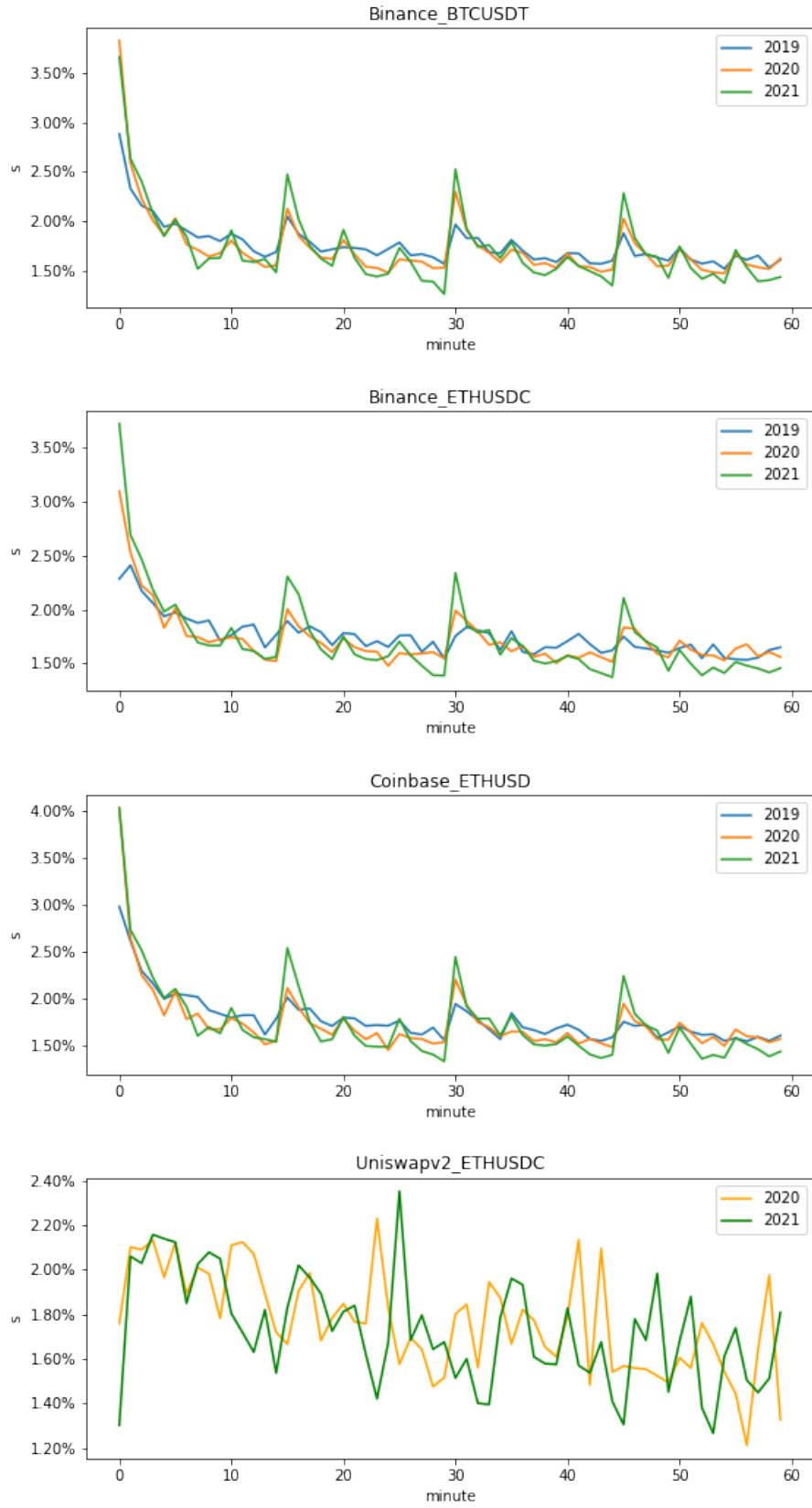
$$\overline{RV}_{y,i}^{min} = \frac{1}{H} \sum_{h=0}^{H-1} RV_{y,60h+i}^{min}$$

for  $i = 0, \dots, 59$  and  $H$  is the number of hours for year  $y$ , so  $H = 24 \times 365$ .



## Seasonlity statistics

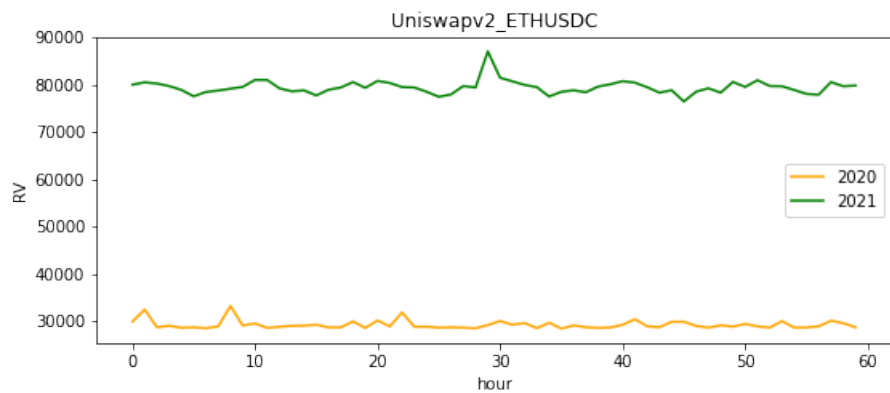
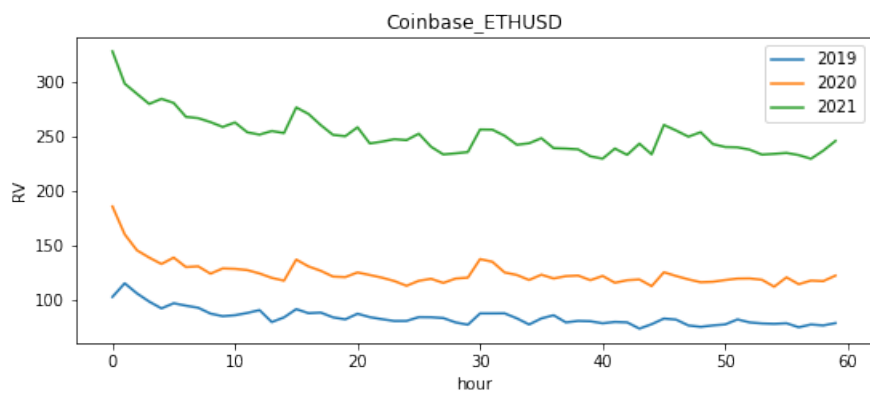
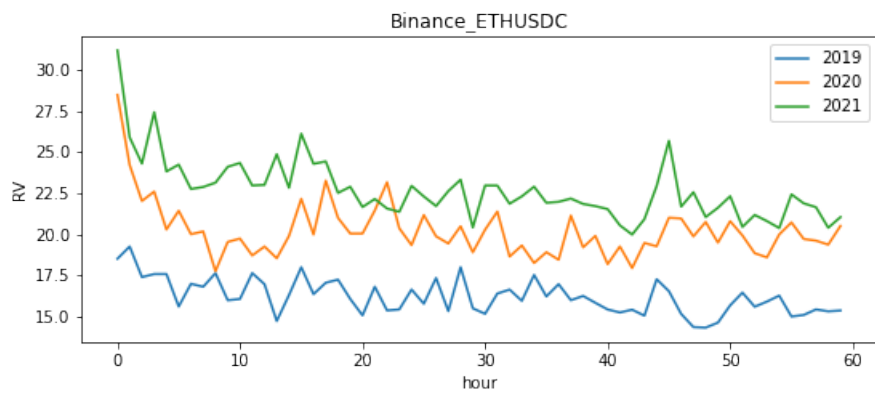
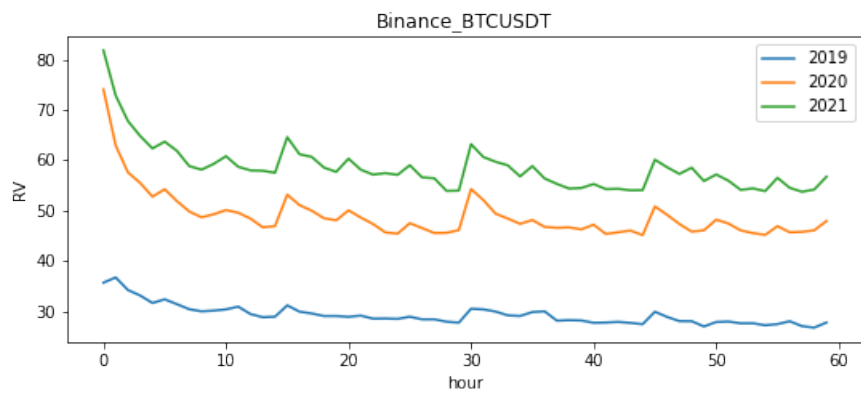
$$s_{y,i}^{min} = \frac{100}{H} \sum_{h=0}^H \frac{RV_{y,60h+i}^{min}}{\sum_{j=0}^{59} RV_{y,60h+i-j}^{min}}$$



In the centralized exchanges, the volatility sharply increases at 0,15,30,45 minutes and this pattern has been intensified over the periods. However, Uniswap does not show this pattern.

### 3.3.2 Volume

$$\bar{V}_{y,i}^{min} = \frac{1}{H} \sum_{h=0}^{H-1} V_{y,60h+i}^{min}$$



$$S_{y,i}^{min} = \frac{100}{H} \sum_{h=0}^{H-1} \frac{V_{y,60h+i}^{min}}{\sum_{j=0}^{59} V_{y,60h+i-j}^{min}}$$

