

Exchange Traded Funds

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```
library(QRM)
library(qrmtools)
library(readr)
library(tidyverse)
library(zoo)
library(xts)
library(quantmod)
library(ggplot2)
library(magrittr)
library(broom)
library(lubridate)
library(PerformanceAnalytics)
```

Reading the data

- In this investigation, the following are the 4 ETFs - VOO, QQQ, XBI, VYM are selected from NYSE which are listed in US and analyzed. They are considered to be among the best ETFs for many years. This investigation will compare and benchmark their performance with the S&P500 to compare how they fare in terms of risk, returns, standard deviation, correlations and dividend paid. The data used is generated from Yahoo Finance and is from 2016-01-04 to 2020-10-21.
- ^GSPC: S&P500
- VOO: Vanguard 500 Index Fund ETF invests in stocks in the S&P 500 Index, representing 500 of the largest U.S. companies. VOO offers high potential for investment growth, their share value rises and falls more sharply than that of funds holding bonds and is more appropriate for long-term goals where money growth is essential.
- QQQ: Tracks the Nasdaq 100 Index and focuses on large international and U.S. companies in the technology, healthcare, industrial, consumer discretionary, and telecommunications sectors. Where the S&P 500 tracks large-cap stocks across both major US stock exchanges, QQQ is limited to just the NASDAQ, so you can expect investment to be more heavily influenced by big news in the technology sector more than other industries.
- XBI: SPDR S&P Biotech ETF. XBI tracks an equal-weighted index of US biotechnology stocks. XBI is one of a handful of biotech ETFs available, offering exposure to a corner of the market that can perform well during periods of consolidation and is capable of big jumps in the event of major drug approvals.
- VYM: Vanguard High Dividend Yield ETF seeks to track the performance of the FTSE® High Dividend Yield Index, which measures the investment return of common stocks of companies characterized by high dividend yields. VYM provides a convenient way to track the performance of stocks that are forecasted to have above-average dividend yields and follows a passively managed, full-replication approach.

```
## 'getSymbols' currently uses auto.assign=TRUE by default, but will
## use auto.assign=FALSE in 0.5-0. You will still be able to use
## 'loadSymbols' to automatically load data. getOption("getSymbols.env")
## and getOption("getSymbols.auto.assign") will still be checked for
## alternate defaults.
##
## This message is shown once per session and may be disabled by setting
## options("getSymbols.warning4.0"=FALSE). See ?getSymbols for details.
```

```
## [1] "QQQ" "XBI" "VYM" "VOO" "^GSPC"
```

Calculating returns

- The returns are calculated on a daily basis from 2016-01-24 to 2020-10-21 and an average is generated for each month. In total there are 57 monthly periods.

```
etf1 <- apply.monthly(etf, mean)
etf1_return_discrete = Return.calculate(etf1, method = c("discrete"))
head(etf1)
```

##	GSPC.Adjusted	VOO.Adjusted	QQQ.Adjusted	XBI.Adjusted	VYM.Adjusted
## 2016-01-29	1918.598	159.9719	99.61326	56.45628	54.57969
## 2016-02-29	1904.418	159.1022	96.72721	48.16275	55.47697
## 2016-03-31	2021.954	169.3284	102.63474	50.25948	58.82765
## 2016-04-29	2075.535	174.0643	105.36269	55.46805	60.18922
## 2016-05-31	2065.550	173.5177	102.86296	52.86124	60.22427
## 2016-06-30	2083.891	175.4221	104.08364	54.83117	61.32005

```
str(etf1)
```

```
## An 'xts' object on 2016-01-29/2020-10-21 containing:
##   Data: num [1:58, 1:5] 1919 1904 2022 2076 2066 ...
##   - attr(*, "dimnames")=List of 2
##   ..$ : NULL
##   ..$ : chr [1:5] "GSPC.Adjusted" "VOO.Adjusted" "QQQ.Adjusted" "XBI.Adjusted" ...
##   Indexed by objects of class: [POSIXct,POSIXt] TZ: UTC
##   xts Attributes:
##   List of 2
##   $ src      : chr "yahoo"
##   $ updated: POSIXct[1:1], format: "2020-10-31 00:22:22"
```

```
str(etf1_return_discrete)
```

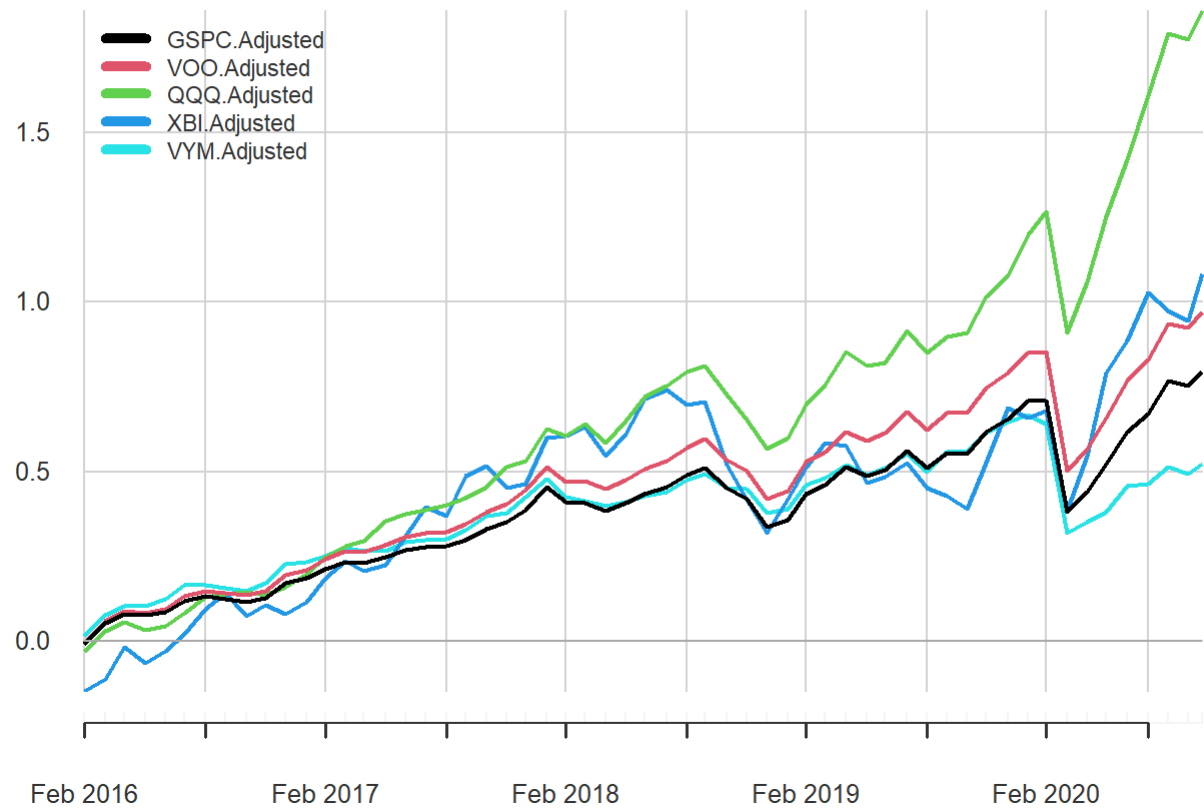
```
## An 'xts' object on 2016-01-29/2020-10-21 containing:
##   Data: num [1:58, 1:5] NA -0.00739 0.06172 0.0265 -0.00481 ...
##   - attr(*, "dimnames")=List of 2
##     ..$ : NULL
##     ..$ : chr [1:5] "GSPC.Adjusted" "VOO.Adjusted" "QQQ.Adjusted" "XBI.Adjusted" ...
##   Indexed by objects of class: [POSIXct,POSIXt] TZ: UTC
##   xts Attributes:
## List of 4
##  $ src           : chr "yahoo"
##  $ updated        : POSIXct[1:1], format: "2020-10-31 00:22:22"
##  $ ret_type       : chr "discrete"
##  $ coredata_content: chr "discreteReturn"
```

Performance charts

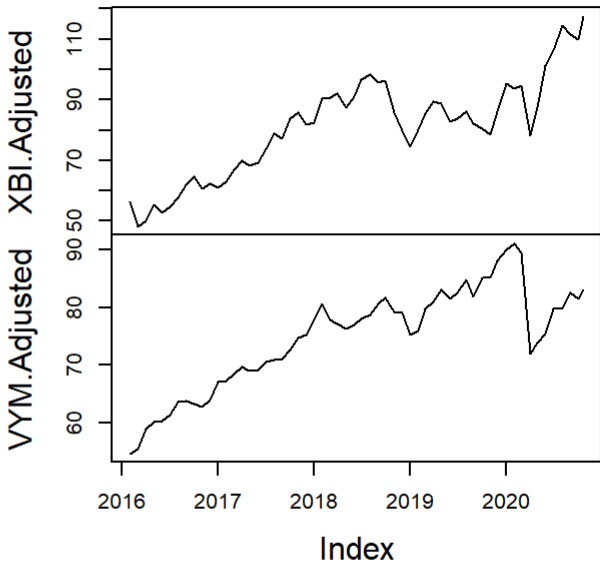
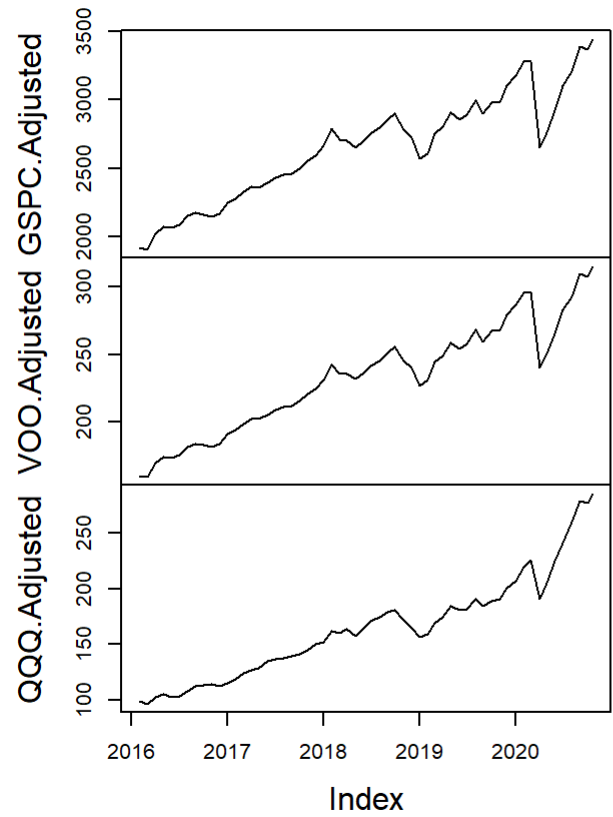
- The cumulative returns represents the total change in investment price over the 57 month periods. QQQ appears to be ahead in terms of cumulative return while VOO is second with VYM and XBI coming close behind.
- In terms of

ETF Absolute Performance

2016-02-29 / 2020-10-21



etf1

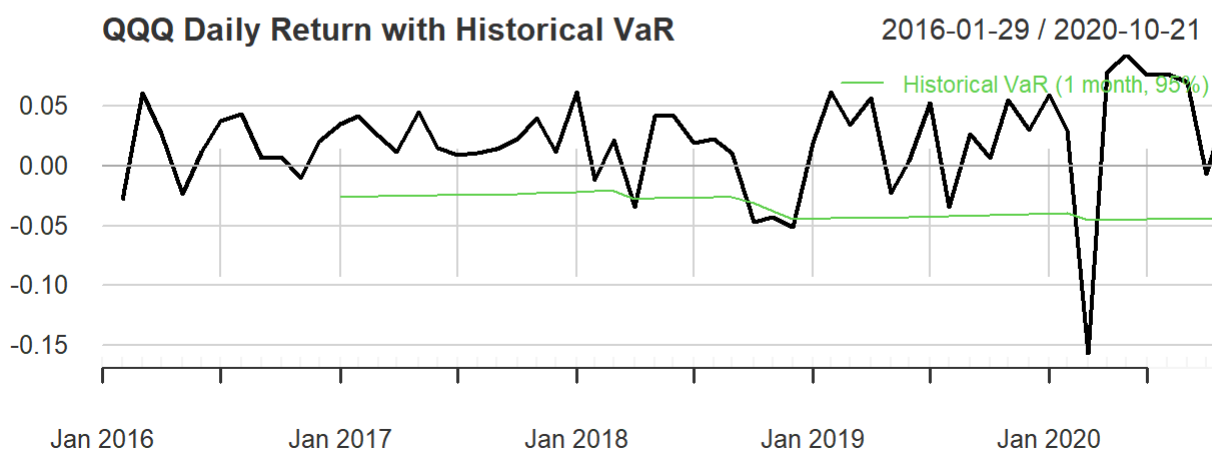
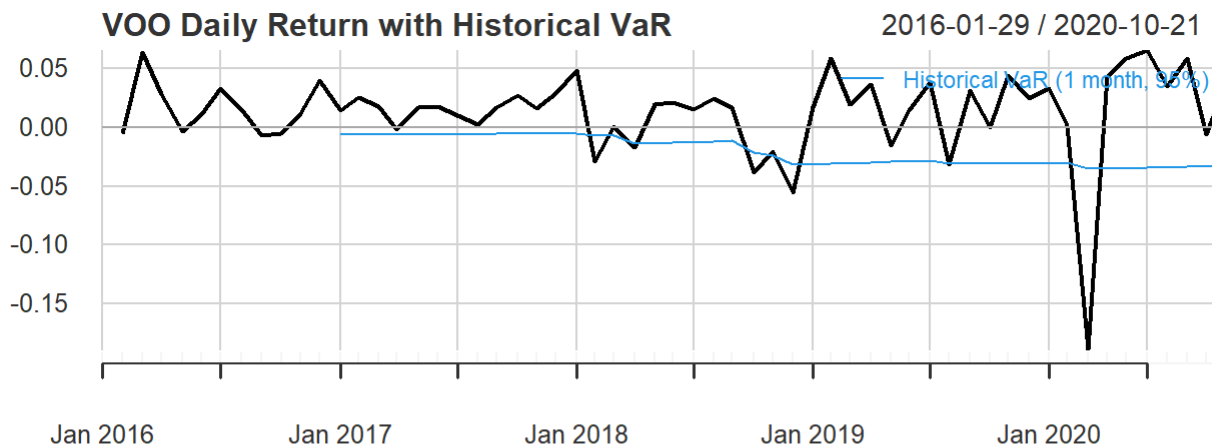


Value at Risk

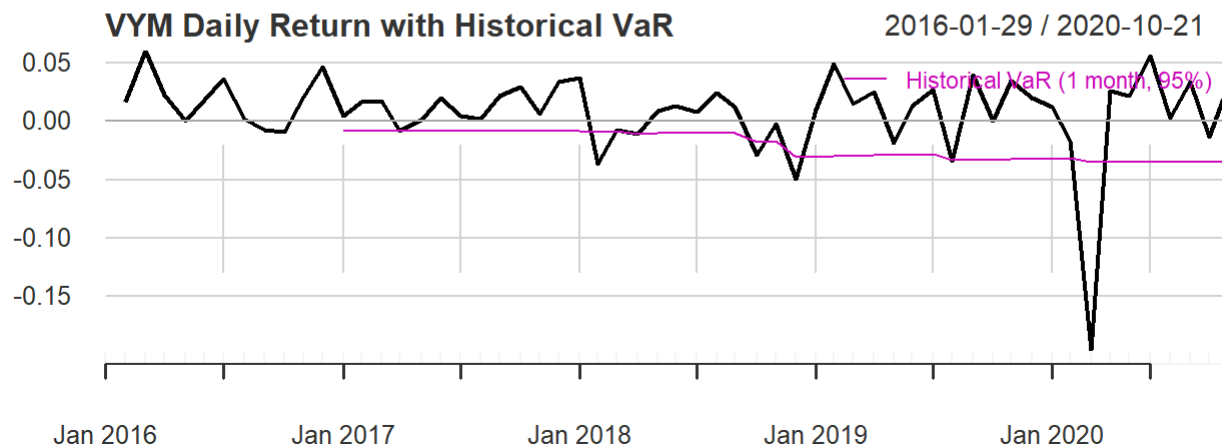
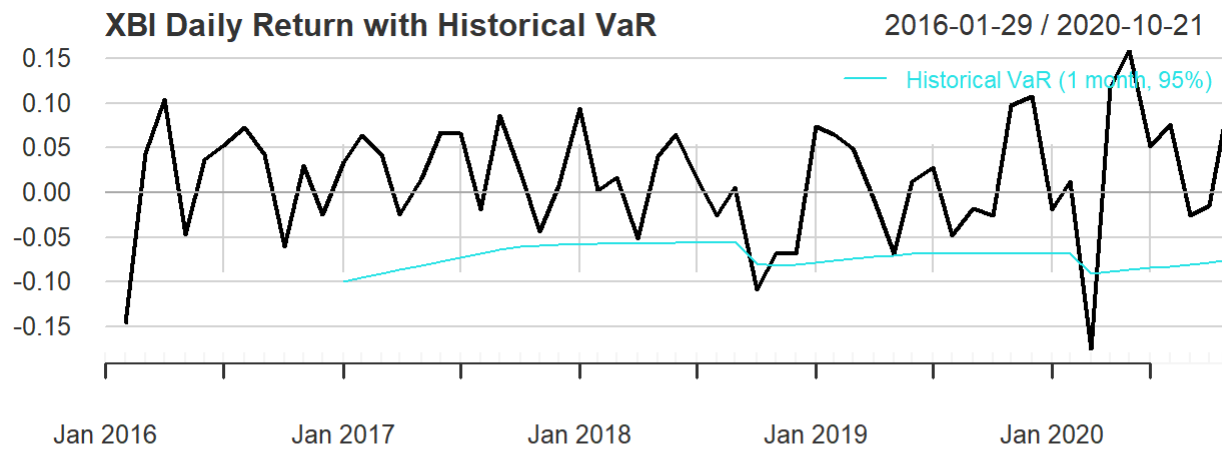
- Historical Value at Risk (VaR) reorganizes historical returns ranking them from worst to best and assuming history will repeat itself. These plots shows the daily returns overlaid with a rolling measure of tail risk referred as a modified VAR.
- VOO Value at Risk is at -1.76%, this means that in a day there is a 5% probability that VOO return could lose more than 1.76% of its stock price.
- QQQ is at -2.14%, XBI at -3.25% and VYM at -1.66%. Therefore, XBI has the highest historical VaR.

```
par(mfrow=c(2,1))
```

```
chart.BarVaR(etf1_return_discrete[,2], methods="HistoricalVaR",lty=1, lwd=2, main="VOO Daily Ret
urn with Historical VaR", legend.loc="topright", colorset=4)
chart.BarVaR(etf1_return_discrete[,3], gap=12, methods="HistoricalVaR",lty=1, lwd=2, main='QQQ D
aily Return with Historical VaR', legend.loc="topright", colorset=3)
```



```
chart.BarVaR(etf1_return_discrete[,4], gap=12, methods="HistoricalVaR",lty=1, lwd=2, main='XBI D
aily Return with Historical VaR', legend.loc="topright", colorset=5)
chart.BarVaR(etf1_return_discrete[,5], gap=12, methods="HistoricalVaR",lty=1, lwd=2, main='VYM D
aily Return with Historical VaR', legend.loc="topright", colorset=6)
```

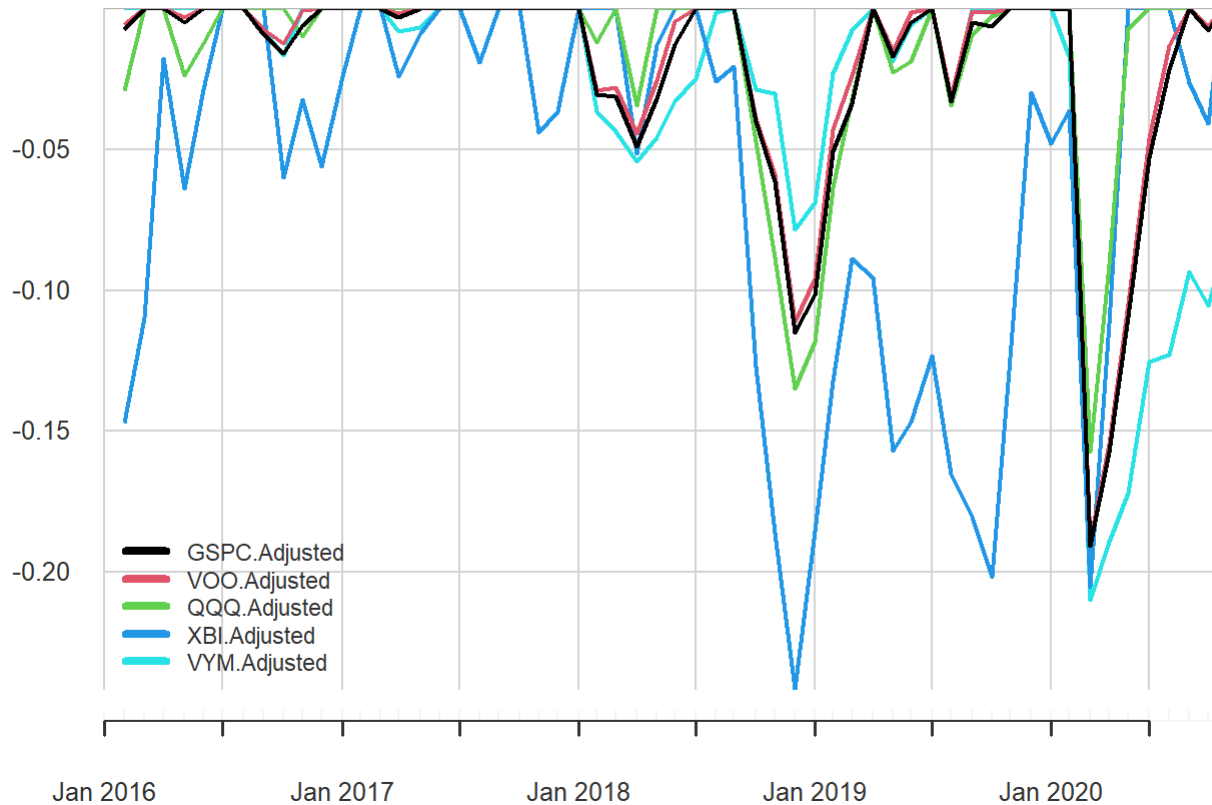


- The drawdown refers to how much the ETF has dropped down from the peak before it recovers back to the peak. XBI appears to have the greatest negative drawdown which means the highest downside volatility.

```
chart.Drawdown(etf1_return_discrete[,c(1:5)],main='Drawdown Performance',
               legend.loc="bottomleft", lwd=4)
```

Drawdown Performance

2016-01-29 / 2020-10-21



- Based on the table, QQQ has the highest average discrete monthly return 1.94%, VOO at 1.27%, XBI at 1.50%, VYM at 0.81%.
- In addition, standard deviation and variance are also useful for comparing the volatility of the returns of each ETF.
- Furthermore skewness of VOO, QQQ and VYM are less than -1 and are considered their returns are highly skewed.
- In terms of kurtosis which represents the heaviness of the tails of stock price, VOO and VYM has the heaviest extreme values. That means the investor will experience occasional extreme returns either positive or negative.

```
table.Stats(etf1_return_discrete[,c(1:5)])
```

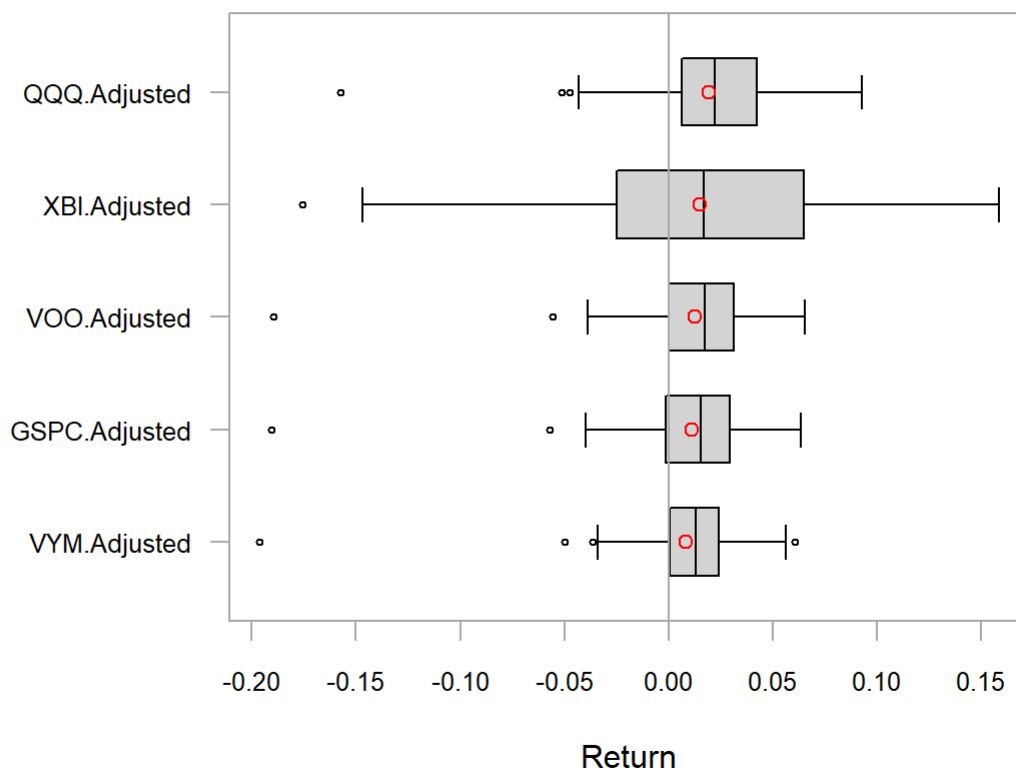
```
##          GSPC.Adjusted V00.Adjusted QQQ.Adjusted XBI.Adjusted
## Observations      57.0000      57.0000      57.0000      57.0000
## NAs                1.0000      1.0000      1.0000      1.0000
## Minimum           -0.1907     -0.1894     -0.1573     -0.1756
## Quartile 1        -0.0015     -0.0001      0.0065     -0.0247
## Median            0.0153      0.0172      0.0223      0.0169
## Arithmetic Mean    0.0110      0.0127      0.0194      0.0150
## Geometric Mean     0.0103      0.0120      0.0186      0.0130
## Quartile 3         0.0292      0.0312      0.0421      0.0648
## Maximum           0.0634      0.0655      0.0925      0.1587
## SE Mean           0.0049      0.0049      0.0054      0.0085
## LCL Mean (0.95)    0.0012      0.0028      0.0086     -0.0021
## UCL Mean (0.95)    0.0209      0.0226      0.0303      0.0322
## Variance           0.0014      0.0014      0.0017      0.0042
## Stdev              0.0371      0.0371      0.0408      0.0645
## Skewness          -2.8888     -2.8856     -1.4522     -0.5396
## Kurtosis           13.6450     13.6314      4.5872      0.5254
##          VYM.Adjusted
## Observations      57.0000
## NAs                1.0000
## Minimum           -0.1964
## Quartile 1         0.0006
## Median            0.0132
## Arithmetic Mean    0.0081
## Geometric Mean     0.0074
## Quartile 3         0.0242
## Maximum           0.0604
## SE Mean           0.0047
## LCL Mean (0.95)   -0.0013
## UCL Mean (0.95)    0.0175
## Variance           0.0013
## Stdev              0.0354
## Skewness          -3.4312
## Kurtosis           17.7366
```

Comparing distribution using boxplots

- XBI has the greatest variability and highest standard deviation of stock price, 6.45%. VYM has the smallest variability among the 5 investments and this shows that VYM has the lowest volatility in terms of discrete returns.

```
chart.Boxplot(etf1_return_discrete[,c(1:5)], main = "Trailing 36-Month Returns")
```

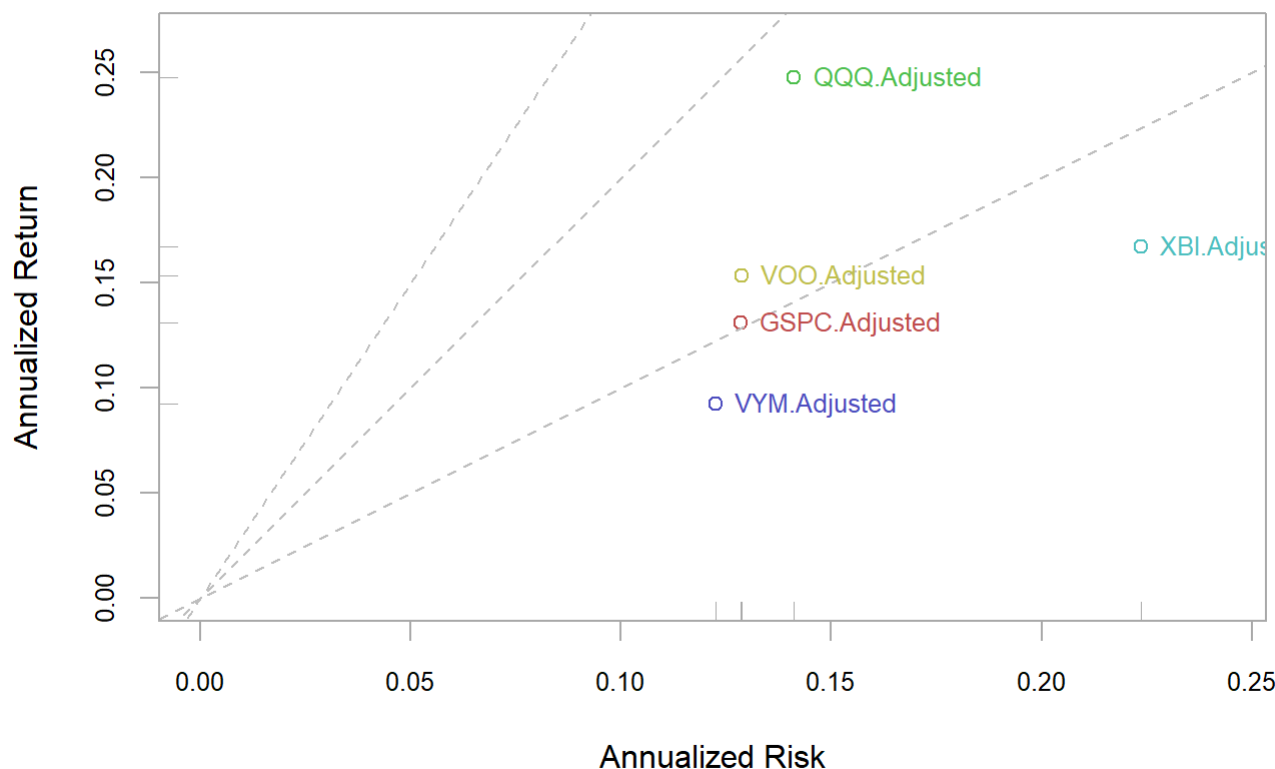

Trailing 36-Month Returns



- The three lines plot show the slope of 1,2,3 for the sharpe ratio.
- VOO offers the higher annualized return for almost the same annualized risk as VYM.
- In addition, QQQ has the strongest return-risk reward while XBI has the highest annualized risk.

```
chart.RiskReturnScatter(etf1_return_discrete[,c(1:5)],
  main = "Trailing Performance",
  colorset=rainbow6equal)
```

Trailing Performance

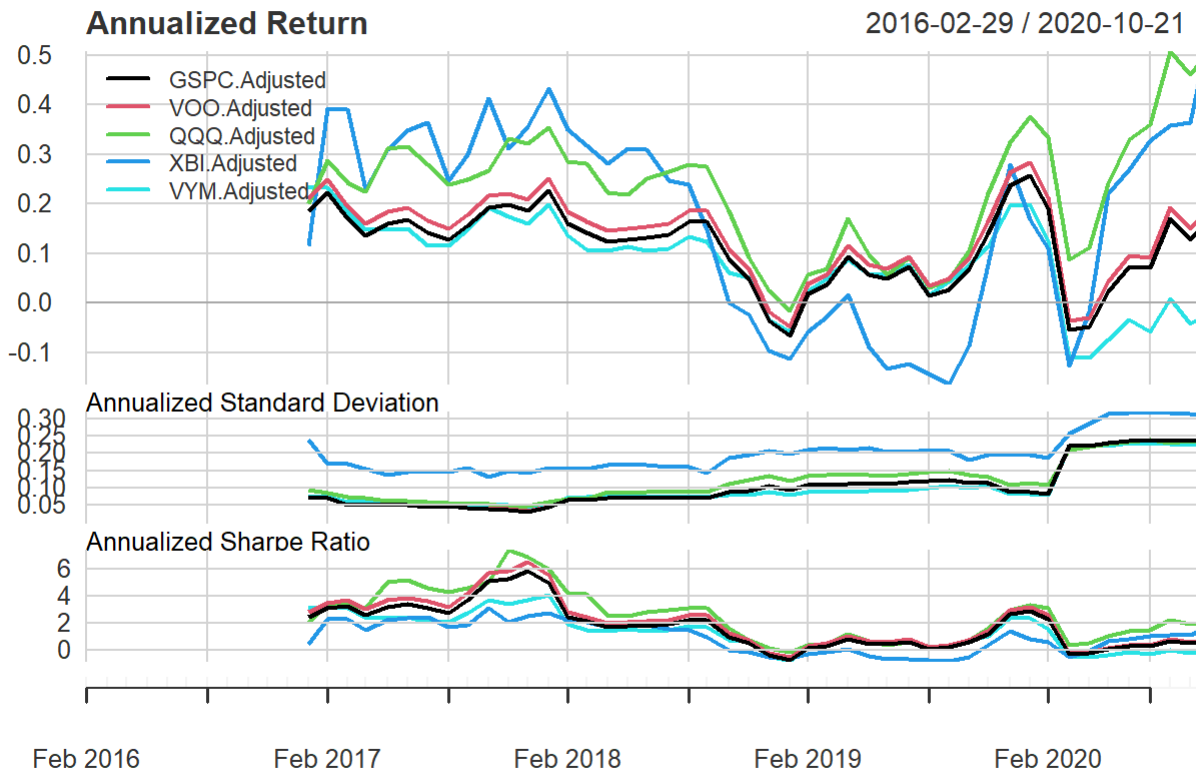


Measuring performance consistency

- Rolling performance is typically used as a way to assess stability of a return stream.
- For XBI, there was a drop from Apr 19 however the index price shot up in Sep 2019 and dipped to a trough again in Feb 2020, eventually peaking due to the coronavirus that has played a huge role in their returns due to the biomedical nature of XBI.
- VYM follows a similar trend as XBI however since Mar 2020 the price have not rebound as quickly as the other indexes.
- Based on annualized sharpe ratio, there is a convergence among the ETFs and the S&P500 and the tend to move in a similar fashion. That can be said for the annualized standard deviation as well. With that said, XBI has the highest annualized standard deviation which is a constant finding in this investigation across various measures.
- In addition, QQQ and XBI have performed exceptionally well in 2020 compared to the other two ETFs and the S&P500.

```
charts.RollingPerformance(etf1_return_discrete[,c(1:5)],
                          width = 12,
                          main="Rolling 12-Month Performance",
                          legend.loc="topleft",
                          lwd=2)
```

Rolling 12-Month Performance



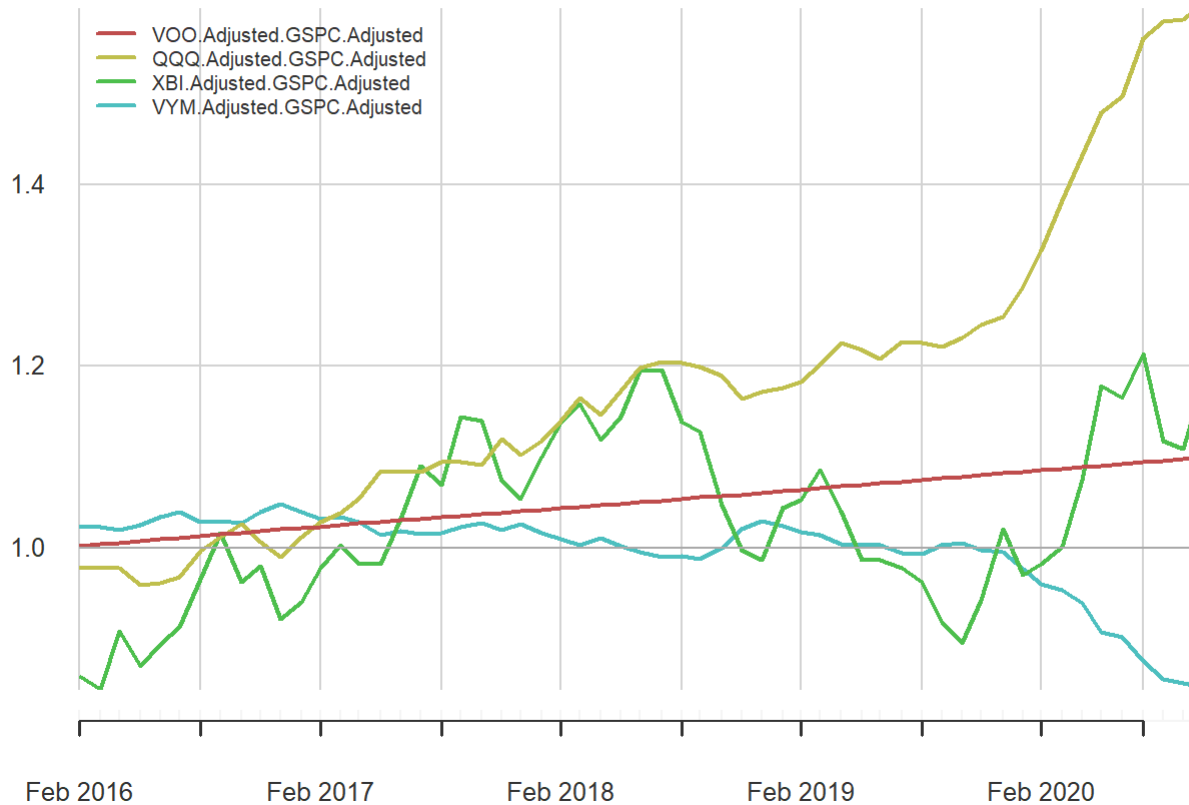
Measuring relative performance

- Identifying and using a benchmark can help us assess and explain how well in terms of meeting investment objectives, in terms of a widely held substitute. In this case, the S&P500 is used as a benchmark.
- QQQ performs very well relative to the benchmark and XBI has managed to climb and reach its peak in 2020 due to the coronavirus effect and the biomedical research industry that XBI captures.

```
chart.RelativePerformance(etf1_return_discrete[,c(2:5), drop =FALSE],
  etf1_return_discrete[,1,drop =FALSE],
  colorset = rainbow6equal,
  lwd = 6,
  legend.loc = "topleft",
  main="Relative performance to S&P500")
```

Relative performance to S&P500

2016-02-29 / 2020-10-21



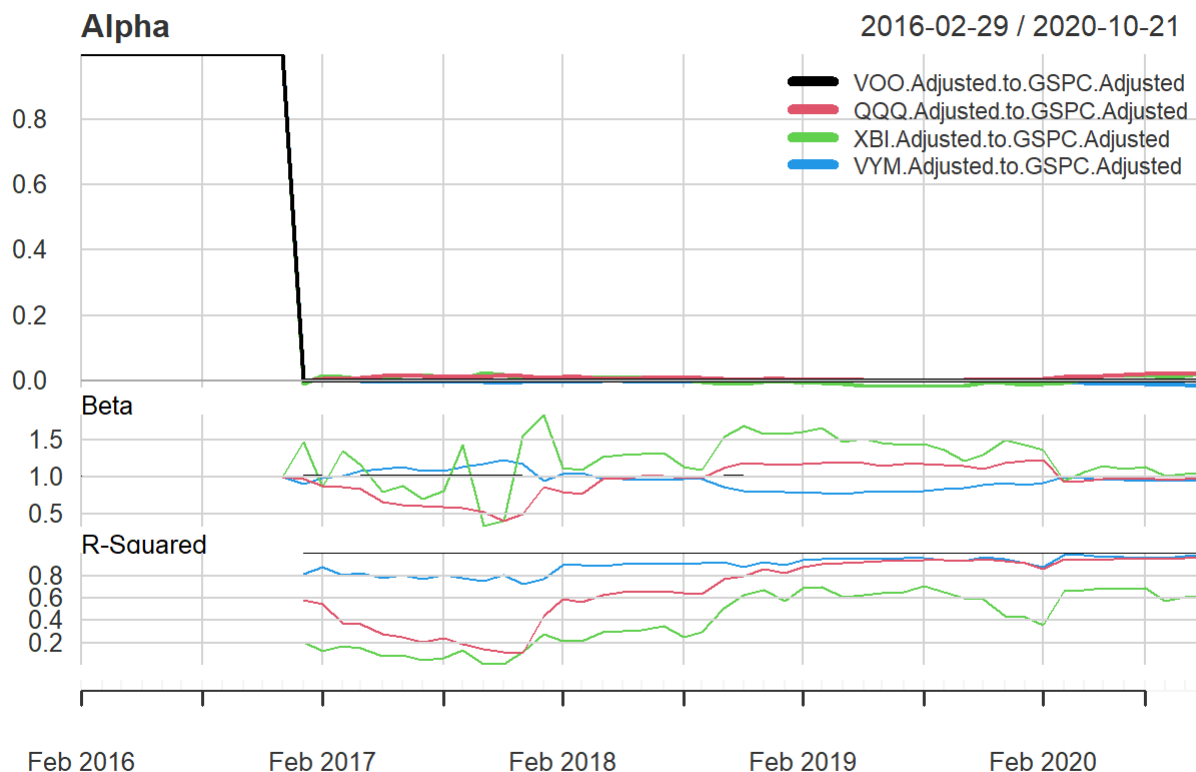
- QQQ has the highest alpha at 0.82% based off a monthly period.
- XBI has the highest beta at 1.1795 which means that XBI is theoretically more volatile than the market. While VYM has the lowest beta at 0.9151 which means that VYM is theoretically less volatile than the market.
- XBI has a relative low R squared at 0.4587 almost two times lower than QQQ and VYM, this effectively means XBI does not move in as much tandem with the S&P500 benchmark as compared to QQQ, VYM and VOO.
- QQQ has the highest Treynor ratio which indicates that it generated 0.24 excess return for each unit of risk taken on.

```
table.CAPM(etf1_return_discrete[,c(2:5), drop =FALSE],
           etf1_return_discrete[,c(1),drop =FALSE],)
```

##	V00.Adjusted to GSPC.Adjusted	QQQ.Adjusted to GSPC.Adjusted
## Alpha	0.0016	0.0082
## Beta	1.0023	1.0139
## Beta+	1.0018	1.1571
## Beta-	1.0019	0.8701
## R-squared	0.9999	0.8494
## Annualized Alpha	0.0199	0.1035
## Correlation	1.0000	0.9216
## Correlation p-value	0.0000	0.0000
## Tracking Error	0.0011	0.0548
## Active Premium	0.0226	0.1164
## Information Ratio	21.0871	2.1232
## Treynor Ratio	0.1533	0.2441
##	XBI.Adjusted to GSPC.Adjusted	VYM.Adjusted to GSPC.Adjusted
## Alpha	0.0020	-0.0020
## Beta	1.1795	0.9151
## Beta+	0.6183	0.7278
## Beta-	0.8049	1.0151
## R-squared	0.4587	0.9177
## Annualized Alpha	0.0245	-0.0242
## Correlation	0.6772	0.9580
## Correlation p-value	0.0000	0.0000
## Tracking Error	0.1661	0.0368
## Active Premium	0.0361	-0.0386
## Information Ratio	0.2172	-1.0477
## Treynor Ratio	0.1417	0.1010

```
charts.RollingRegression(etf1_return_discrete[,c(2:5), drop =FALSE],
                          etf1_return_discrete[,c(1),drop =FALSE],
                          lwd = 5,
                          legend.loc = "topright")
```

Rolling 12-month Regressions

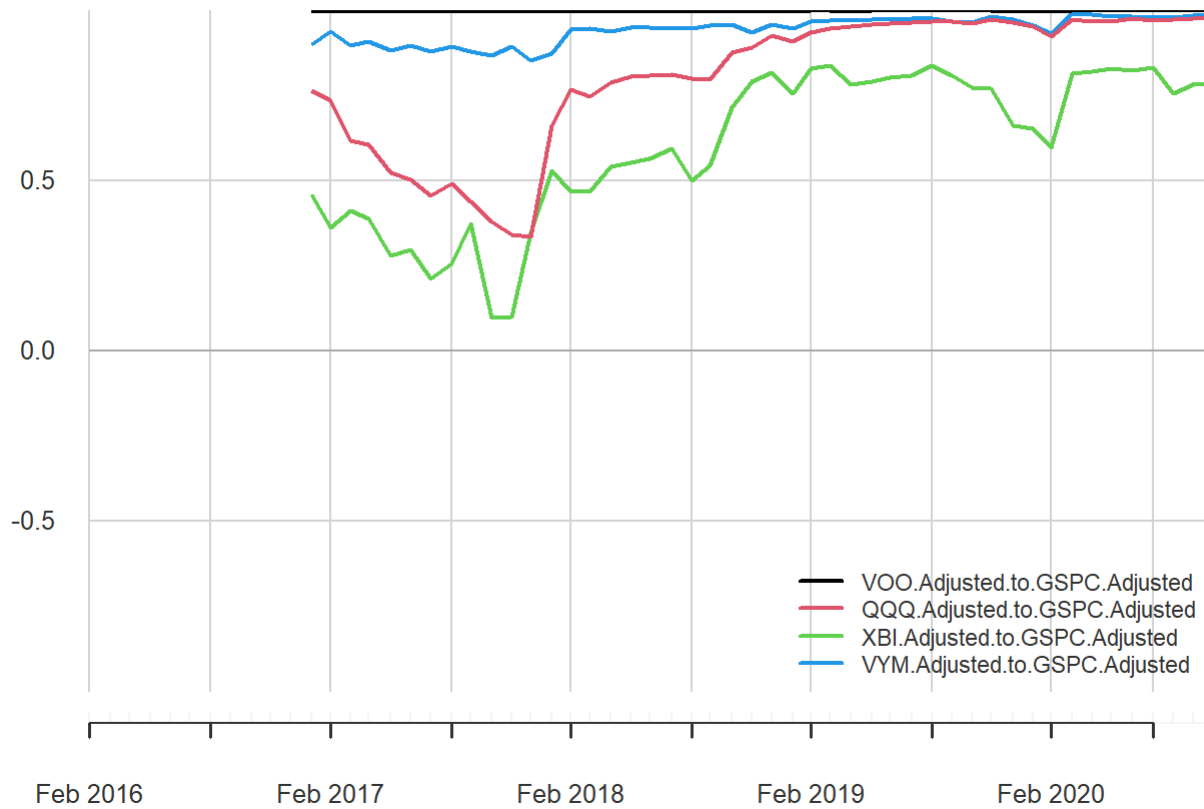


- Through the rolling correlation, XBI, QQQ and VYM are moving close to that of the S&P500 in terms of correlation.
- XBI has a correlation that is not as near that of the S&P500 due to the nature of their industry and the small biomedical and research companies that the index tracks.
- Apart from VOO and GSPC, VYM and VOO has the highest correlation at 0.96 followed by QQQ and VOO at 0.93.
- The correlation between each stock is high because VOO and VYM tracks similar size companies within the same industries such as consumer, healthcare and technology, with VYM tilted more heavily to dividend payers.
- That can be said for VOO and QQQ of which is made up by 34.41% and 62.89% in technology sector respectively.

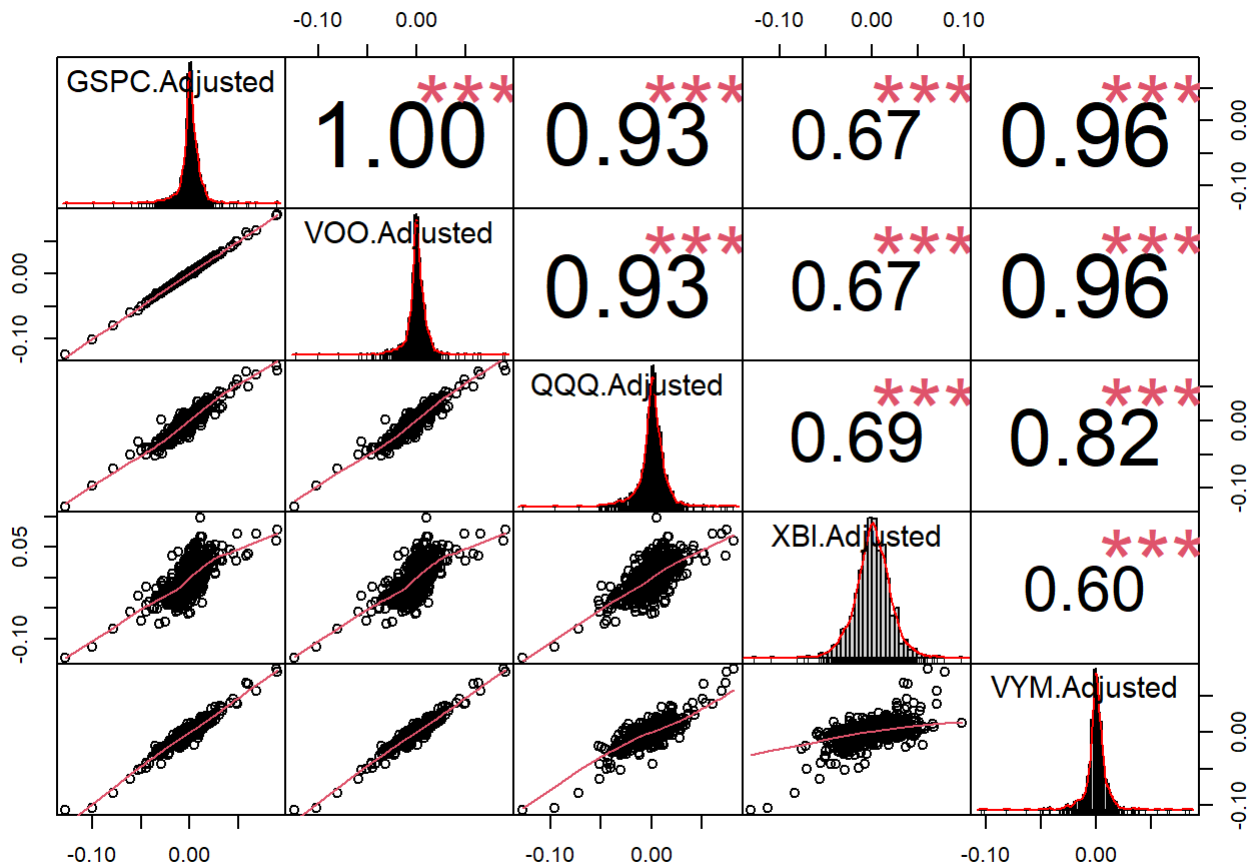
```
chart.RollingCorrelation(etf1_return_discrete[,c(2:5), drop =FALSE],
                        etf1_return_discrete[,c(1),drop =FALSE],
                        lwd = 2,
                        main = "Rolling Correlation",
                        legend.loc = "bottomright")
```

Rolling Correlation

2016-02-29 / 2020-10-21



```
data <- diff(log(etf))  
chart.Correlation(data)
```



Downside Risk

- Semi deviation looks only at negative price fluctuation and is an alternative measurement to standard deviation or variance. It is used to evaluate the downside risk of an investment. A higher number represents greater fluctuations in the negative price of the ETF.
- Gain deviation is similar calculation to standard deviation and the opposite of loss deviation. It calculates the deviation using only up period returns, variances and the number of up periods, a higher number is means there is more deviation in the gains of the ETF.
- Loss deviation is similar to standard deviation but calculates the deviation using only the down period returns, variances and number of down periods.

```
table.DownsideRisk(etf1_return_discrete[,1:5])
```


##	GSPC.Adjusted	V00.Adjusted	QQQ.Adjusted
## Semi Deviation	0.0316	0.0316	0.0324
## Gain Deviation	0.0161	0.0168	0.0230
## Loss Deviation	0.0452	0.0473	0.0399
## Downside Deviation (MAR=10%)	0.0306	0.0301	0.0280
## Downside Deviation (Rf=0%)	0.0280	0.0276	0.0251
## Downside Deviation (0%)	0.0280	0.0276	0.0251
## Maximum Drawdown	0.1909	0.1894	0.1573
## Historical VaR (95%)	-0.0343	-0.0328	-0.0440
## Historical ES (95%)	-0.0960	-0.0946	-0.0852
## Modified VaR (95%)	-0.0637	-0.0622	-0.0584
## Modified ES (95%)	-0.1566	-0.1553	-0.1160
##	XBI.Adjusted	VYM.Adjusted	
## Semi Deviation	0.0483	0.0307	
## Gain Deviation	0.0361	0.0151	
## Loss Deviation	0.0440	0.0494	
## Downside Deviation (MAR=10%)	0.0449	0.0308	
## Downside Deviation (Rf=0%)	0.0409	0.0283	
## Downside Deviation (0%)	0.0409	0.0283	
## Maximum Drawdown	0.2416	0.2101	
## Historical VaR (95%)	-0.0763	-0.0344	
## Historical ES (95%)	-0.1438	-0.0941	
## Modified VaR (95%)	-0.0990	-0.0636	
## Modified ES (95%)	-0.1371	-0.1557	