Financial Data Analysis

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Data

Raw Data

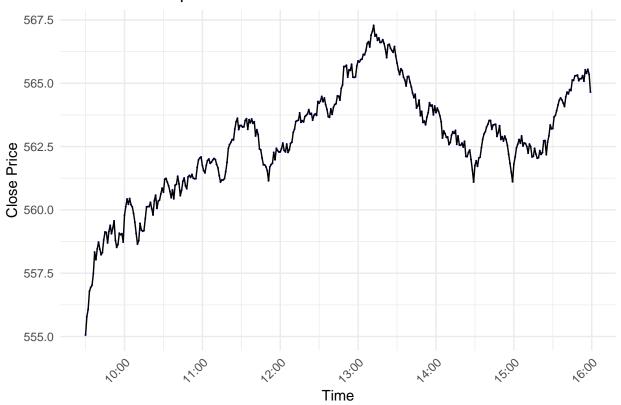
```
#political shocks
#raw_truths <- read.csv(here("data/political_data", "trump_all_truths.csv"))</pre>
#raw_tweets <- read.csv(here("data/political_data", "tweets.csv"))</pre>
#market prices
raw_ONEQ <- read.csv(here("data/market_data", "ONEQ.csv")) #USA</pre>
raw_SMI <- read.csv(here("data/market_data", "SMI.csv")) #CH</pre>
raw_VTHR <- read.csv(here("data/market_data", "VTHR.csv")) #USA</pre>
raw_VTI <- read.csv(here("data/market_data", "VTI.csv")) #USA</pre>
raw_DAX <- read.csv(here("data/market_data", "DAX.csv")) #DE</pre>
raw_ASHR <- read.csv(here("data/market_data", "ASHR.csv")) #CHINA</pre>
data_loader(year=2021,months=1:6,symbol="SPY")
#data_loader(year=2024,months=1:12,symbol="SPY")
data_loader(year=2025,months=1:4,symbol="SPY")
data_loader(year=2024,months=1:12,symbol="VGK") #EU
data_loader(year=2025,months=1:4,symbol="VGK")
data_loader(year=2025,months=1:2,symbol="ASHR")
data_loader2(year=2024,months=1:12,"SPY")
```

Quick Analysis

SPY April 2nd 2025

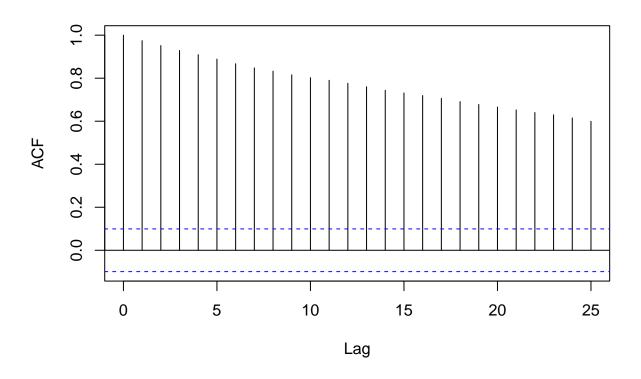
```
#extract a particular day
day_SPY_0402 = day_selector(raw_SPY_2025_04,2025,04,02) #april 2nd 2025
#let's plot it
day_plotter(day_SPY_0402,"SPY Price on April 2nd 2025")
```



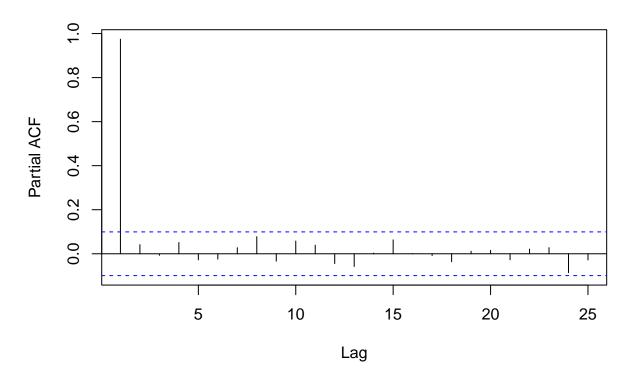


#quickly test some ARMA specifications
quick_arma(day_SPY_0402,1,0,0) #checking AR1,AR2,AR3

Series data\$close



Series data\$close



##		AR Estimations						
##								
##		AR-1	AR-2	AR-3				
##								
##	ar1	0.9975	0.9728	1.4609				
##		(0.0030)	(0.0514)	(NaN)				
##	intercept	561.0971	561.3655	562.5635				
##		(3.2897)	(3.4352)	(22.1897)				
##	ar2		0.0249	0.0770				
##			(0.0515)	(0.0013)				
##	ar3			-0.5386				
##				(0.0007)				
##								
##	nobs	390	390	390				
##	sigma	0.2854	0.2853	0.3414				
##	logLik	-67.0847	-66.9808	-135.4359				
##	AIC	140.1693	141.9615	280.8718				
##	BIC	152.0678	157.8261	300.7025				
##	nobs.1	390.0000	390.0000	390.0000				
##								
##	*** p < 0.001; ** p < 0.01; * p <							
##	0.05.							
##								
##	Column names: names, AR-1,	, AR-2, AR-3						
##	Checking Residuals							
##								

```
##
                            AR-1 Residuals
                                              AR-2 Residuals
                                                                AR-3 Residuals
##
          (Intercept)
##
                                  0.0302 *
                                                     0.0291 *
                                                                    -0.0051
                                                    (0.0145)
                                                                    (0.0171)
##
                                  (0.0145)
##
          REG1res_lagged
                                  -0.0476
##
                                  (0.0510)
##
          REG2res lagged
                                                    -0.0217
##
                                                    (0.0511)
##
          REG3res_lagged
                                                                    -0.1733 ***
##
                                                                    (0.0503)
##
                                 389
##
          N
                                                   389
                                                                   389
##
          R.2
                                   0.0022
                                                     0.0005
                                                                     0.0297
##
##
          *** p < 0.001; ** p < 0.01; * p < 0.05.
##
## Column names: names, AR-1 Residuals, AR-2 Residuals, AR-3 Residuals
#quick_arma(day_SPY_0402,2,0,0) #checking AR2,AR3,AR4
```

Realised Volatility

```
#for a particular day (outputs scalar)
r.vol(day_SPY_0402)

## [1] 0.08152862

#for a month (outputs vector of each day's realised volatility)
r.vol_month(raw_SPY_2025_04)

## [1] 0.07851997 0.08152862 0.13381914 0.44172333 2.05732267 0.43592888

## [7] 1.67344102 0.73881997 0.36128756 0.17859522 0.11181995

#for each day in each month of one year
r.vol_year(raw_SPY_2024)
```

```
[,6]
##
                [,1]
                            [,2]
                                        [,3]
                                                   [,4]
                                                               [,5]
##
                 NA 0.021143900 0.011284204 0.01039484 0.073471057
                                                                             NA
    [1,]
  [2,] 0.012415812 0.026401852
                                         NA 0.01106034 0.032803729
                                         NA 0.01569628 0.028870802 0.029894320
  [3,] 0.020250868
                              NA 0.007255057 0.03713616
##
   [4,] 0.013558265
                                                                NA 0.027115784
   [5,] 0.020836134 0.015481985 0.019712747 0.03546997
##
                                                                NA 0.010274598
##
  [6,]
                 NA 0.015688934 0.040378114
                                                    NA 0.006682677 0.011012847
##
   [7,]
                 NA 0.015511707 0.013316025
                                                    NA 0.010045139 0.022338203
   [8,] 0.011702214 0.007324744 0.032537414 0.01546303 0.008217314
##
## [9,] 0.013565561 0.005462019
                                         NA 0.02976630 0.007987699
                                         NA 0.05831688 0.011167486 0.010134506
## [10,] 0.011226764
## [11,] 0.028957726
                             NA 0.019373235 0.02815141
                                                                NA 0.012942579
## [12,] 0.018413433 0.008574647 0.026763865 0.03218925
                                                                NA 0.034261802
```

```
## [13,] NA 0.019252004 0.015805393 NA 0.005709164 0.017524376
## [14,] NA 0.019557301 0.027563491 NA 0.010022021 0.014908185
## [15,] NA 0.012794836 0.025976894 0.05147409 0.008722512 NA
## [16,] 0.023132659 0.019873140 NA 0.06475655 0.010891813 NA ## [17,] 0.013333943 NA NA 0.05262046 0.006561694 0.013768102
## [18,] 0.016934856 NA 0.011360510 0.03811009 NA 0.009728472
## [19,] 0.016387994 NA 0.013312919 0.05071581 NA NA
## [20,] NA 0.020211411 0.033814023 NA 0.006391525 0.024980110 
## [21,] NA 0.017481851 0.010547097 NA 0.004268227 0.017829988
## [22,] 0.011262476 0.014535184 0.008975325 0.02394531 0.014486274 NA
## [23,] 0.006962253 0.017206658 NA 0.02321498 0.027692195 NA ## [24,] 0.015067410 NA NA 0.02149434 0.015690968 0.018464188
## [25,] 0.014599715 NA 0.005104990 0.02784299 NA 0.012014496
## [26,] 0.011253195 0.010144772 0.005631855 0.01896832
                                                                                                          NA 0.015186078
## [27,] NA 0.007235280 0.010236955 NA NA 0.016284111
## [28,] NA 0.008065767 0.009858483 NA 0.013165048 0.026323009
## [29,] 0.010753872 0.019767203 NA 0.01370803 0.013628710 NA
## [30,] 0.007483591 NA NA 0.02300110 0.013521575 NA
## [31,] 0.034980298 NA NA NA NA 0.022734872 NA
## [1,] 0.024277365 0.08118317 NA 0.056233921 0.034067683 NA
## [2,] 0.014109760 0.11141425 NA 0.025023994 NA 0.007057617
## [3,] 0.026727482 NA 0.03554182 0.053038734 NA 0.007357840
## [4,] NA NA 0.06306683 0.039590995 0.028184648 0.007290868
## [5,] 0.010014673 0.25271880 0.04483728 NA 0.023820010 0.008227533 
## [6,] NA 0.09794829 0.07865960 NA 0.030504596 0.009179047
## [7,] NA 0.09774069 NA 0.023672077 0.016381866 NA ## [8,] 0.006640538 0.06704930 NA 0.017974316 0.009823421 NA
## [9,] 0.007550356 0.05574313 0.02596162 0.012909441 NA 0.011937257
## [10,] 0.006680875 NA 0.03080083 0.020067051 NA 0.010900085
## [11,] 0.029091124 NA 0.06853948 0.013463507 0.013616677 0.006175591
## [12,] 0.020804270 0.04435560 0.04630338 NA 0.017243694 0.013346317 
## [13,] NA 0.02453407 0.02524256 NA 0.027147887 0.017938639
## [14,] NA 0.02887814 NA 0.008339548 0.032959293 NA ## [15,] 0.025983981 0.01767380 NA 0.019469876 0.031111389 NA
## [16,] 0.018021355 0.01808247 0.02271454 0.010670427 NA 0.008383432
## [17,] 0.024554978 NA 0.03173591 0.023053029 NA 0.011695146
## [18,] 0.046885352 NA 0.14493815 0.010288253 0.017144319 0.090883909
## [19,] 0.038526683 0.01496351 0.03160202 NA 0.028431226 0.098734280
## [20,] NA 0.01550493 0.02320854 NA 0.038609709 0.085187673
## [21,] NA 0.03219310 NA 0.014695319 0.054226501 NA
## [21,] NA 0.03219310 NA 0.014695319 0.054226501 NA ## [22,] 0.031365368 0.04686951 NA 0.011858962 0.034440282 NA
## [24,] 0.022507886 0.06811768 0.01822570 0.020819306 NA 0.047487339
## [24,] 0.023376791 NA 0.01616798 0.018362657 NA 0.013891701
## [25,] 0.069171155 NA 0.01071128 0.025115688 0.023451588 NA
## [26,] 0.054778597 0.02239555 0.01843709 NA 0.014858419 0.020061213
## [27,] NA 0.01471526 0.01466890 NA 0.013608355 0.059920090
## [28,] NA 0.02758580 NA 0.016578544 NA NA
## [29,] 0.033572140 0.04919462 NA 0.017842154 0.005804139 NA
## [30,] 0.041303625 0.03523880 0.02055323 0.018900435 NA 0.059037359 
## [31,] 0.063912209 NA NA 0.035515594 NA 0.064453573
```

#for each hour in a day (outputs a vector of each hour's realised volatility)
r.vol_hour(day_SPY_0402)

[1] 0.15760939 0.08701794 0.06571201 0.06303564 0.06319524 0.08271313 0.06726031

#for each hour in a day for each day in a month (outputs a matrix)
month_hour = r.vol_month_hour(raw_SPY_2025_04)
huxtable(data.frame(month_hour))

X1	X2	X3	X4	X5	X6	X7	X8	X9	X10	X11
0.141	0.158	0.278	0.812	1.35	0.578	1.27	0.825	0.812	0.353	0.178
0.13	0.087	0.148	0.445	7.3	0.278	0.526	0.492	0.625	0.205	0.222
0.0766	0.0657	0.137	0.815	2.38	0.29	0.401	0.288	0.314	0.165	0.131
0.0657	0.063	0.16	0.319	0.531	0.452	0.189	0.958	0.256	0.222	0.0745
0.0492	0.0632	0.0927	0.284	0.918	0.439	7.33	1.12	0.355	0.182	0.0878
0.0596	0.0827	0.0745	0.216	0.582	0.496	1.22	0.72	0.153	0.0414	0.0669
0.0613	0.0673	0.112	0.419	0.6	0.61	0.618	0.796	0.17	0.137	0.0517

#plots
#hours in a day

plot(r.vol_hour(day_SPY_0402))

