# Financial Data Analysis

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### Data

#### Raw Data

```
#political shocks
#raw_truths <- read.csv(here("data/political_data", "trump_all_truths.csv"))
#raw_tweets <- read.csv(here("data/political_data", "tweets.csv"))

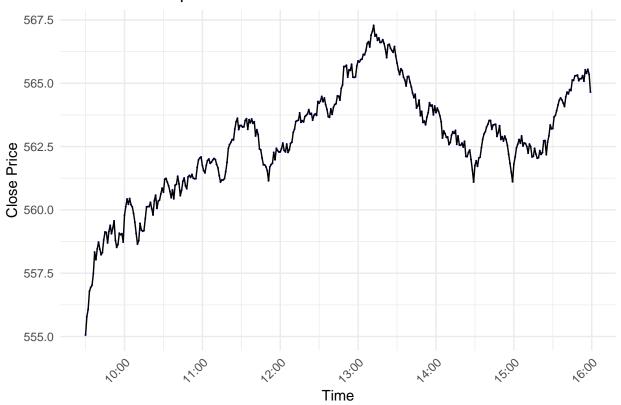
#market prices
raw_ONEQ <- read.csv(here("data/market_data", "ONEQ.csv")) #USA
raw_SMI <- read.csv(here("data/market_data", "SMI.csv")) #CH
raw_VTHR <- read.csv(here("data/market_data", "VTHR.csv")) #USA
raw_DAX <- read.csv(here("data/market_data", "VTI.csv")) #USA
raw_DAX <- read.csv(here("data/market_data", "DAX.csv")) #DE
raw_ASHR <- read.csv(here("data/market_data", "ASHR.csv")) #CHINA
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=2025,months=1:12,symbol="VGK") #EU
data_loader(year=2025,months=1:4,symbol="VGK")
data_loader(year=2025,months=1:2,symbol="VGK")
data_loader(year=2025,months=1:2,symbol="VGK")
data_loader(year=2025,months=1:2,symbol="VGK")
data_loader(year=2025,months=1:2,symbol="VGK")</pre>
```

## **Quick Analysis**

### SPY April 2nd 2025

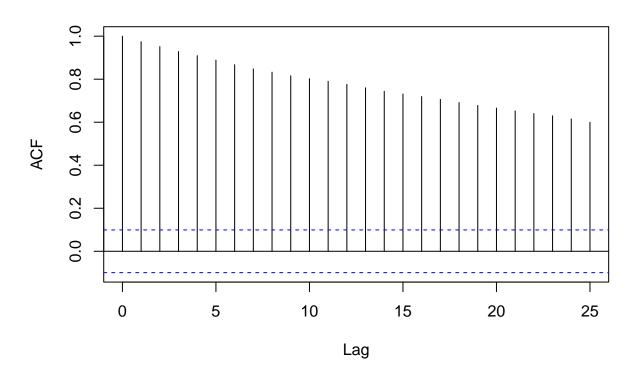
```
#extract a particular day
day_SPY_0402 = day_selector(raw_SPY_2025_04,2025,04,02) #april 2nd 2025
#let's plot it
day_plotter(day_SPY_0402,"SPY Price on April 2nd 2025")
```



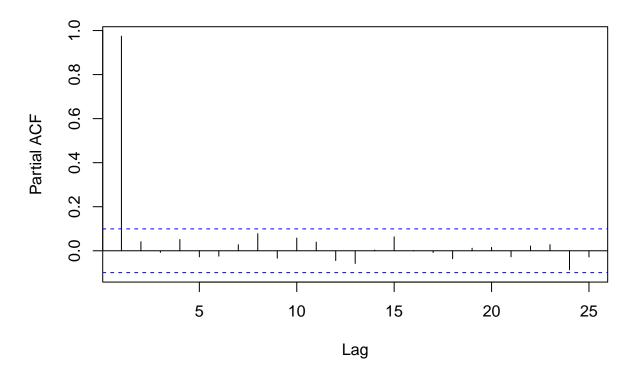


#quickly test some ARMA specifications
quick\_arma(day\_SPY\_0402,1,0,0) #checking AR1,AR2,AR3

## 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							
##			J					

```
##
                            AR-1 Residuals
                                              AR-2 Residuals
                                                                AR-3 Residuals
##
                                                    0.0291 *
                                                                   -0.0051
##
          (Intercept)
                                  0.0302 *
##
                                  (0.0145)
                                                    (0.0145)
                                                                    (0.0171)
##
          REG1res_lagged
                                  -0.0476
##
                                  (0.0510)
##
          REG2res lagged
                                                    -0.0217
                                                    (0.0511)
##
##
          REG3res_lagged
                                                                    -0.1733 ***
                                                                    (0.0503)
##
##
          N
                                 389
                                                                  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

plot(r.vol\_hour(day\_SPY\_0402))

```
## [1] 0.08152862

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

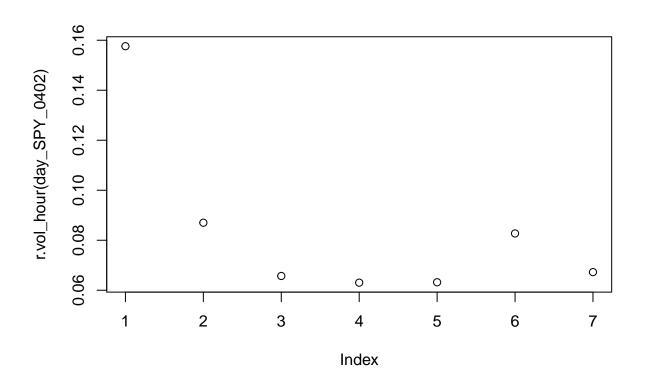
## [1] 0.15760939 0.08701794 0.06571201 0.06303564 0.06319524 0.08271313 0.06726031

#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))

#plots

#hours in a day
```

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



#days in a month
plot(r.vol\_month(raw\_SPY\_2025\_04))

