

# Who are the financial profiteers of war?

University of Applied Sciences Lucerne

Master of Science in Applied Information and Data Science

Time Series in Finance (TSA01)

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TODO: \* Ticker Section Mentioning in Charts and Code -> in Einleitung erwähnt und in Grafik angepasst  
\* Nach differencing nochmals Stationarity testen -> ergänzt \* Causality testing andere Seite -> Einfluss Ticker auf Trends ausschliessen -> in Code ergänzt \* Add Inflation as Controll Variable

## Potential structure

- Context and Goal (Literature?)
- Methods
- Data
  - Import data (indices & Google words)
  - Describe data
  - Preprocessing
- VAR & Causality Testing
- Results & Discussion

## 1. Context and Goal

With recent outbreaks of war (e.g., in Israel or in the Ukraine), the question arises of how stock markets react to this and if there are sectors that profit from such crises. Specifically, this case study examines if several indices from specific economic sectors perform well in situations of crisis. The following six indices are included in the analysis:

- **XLK**: Technology Select Sector
- **XLF**: Financial Select Sector
- **XLV**: Health Care Select Sector
- **XLY**: Consumer Discretionary Select Sector
- **XLE**: Energy Select Sector
- **XLU**: Utilities Select Sector

Essentially our research question is the following: *Who are the financial profiteers of war?*. To answer this question, the performance of the sector-specific indices should be compared to an overall crisis measure. As an overall crisis measure the Google Trend data for the search word “war” is used. We postulate, that an increased search for this term points to a higher crisis. So in conclusion, we want to examine if there are positive correlations between our crisis measure and stock market profits in the different indices.

## 2. Methods

Our research question is of an exploratory nature. Therefore we don’t have a hypothesis going in, as to which indices will perform better under situations of crisis. The method we will use is vector autoregression (VAR). This allows us to measure the influence of our crisis measure on the performance of the different indices. For each index we will calculate a separate model. Furthermore, to address potential spurious correlations a Granger causality test will be conducted for every ticker. However, before a VAR and causality test can be carried out the data in form of different time series has to be imported, cleaned and pre-processed. Concretely, potential seasonal effects and trends in time series have to be identified and adjusted for to achieve stationarity, which characterizes a stabilized mean and variance for the time series over time and will ultimately allow more reliable and accurate modeling. To test the stationarity of a time series the “*Augmented Dickey-Fuller Test*” will be carried out. If a time series is non-stationary, lagged differences will be calculated. Lastly, to deal with right skewness and achieve normal distributions in the data, the time series will first be logarithmized before applying the difference.

### 3. Data

#### Financial Data

The financial data used for the paper at hand comes from Yahoo Finance (2023)(**SOURCE**). The data can be retrieved using the `getSymbols.yahoo()` function from the `quantmod` package. The data represents the weekly adjusted closing prices for the tickers introduced in chapter 1 for the time frame from December 2021 until December 2023. Mentioned time frame was chosen as it captures the start of the Russian invasion of Ukraine as well as the start of the Israel-Hamas war. Additionally, the time frame and periodicity is considered of good balance between having enough data points as well as keeping the data set manageable and focused. The decision to use the adjusted closing prices instead of the closing price can be justified that adjusted prices are adapted to address any splits, dividends or capital gain distributions. Lastly, the weekly periodicity of the ticker data was chosen as the Google Trends data comes in weekly measurements.

#### Google Trends Data

To import the Google Trends data the package `gtrendsR` is used. The package allows to query the data of the interest over time for a keyword. In the case of this study the keyword *war* in web searches was chosen, as it is considered to capture all conflicts alike. When it comes to the geographical origin of the searches the whole world was considered as the study aims to give a generalized and non location specific view. Returned data from Google Trends is normalized and reflects the search volume for a keyword on a scale from 0-100, where 100 means very high interest and 0 no interest. (**SOURCE: <https://newsinitiative.withgoogle.com/en-gb/resources/trainings/fundamentals/google-trends-understanding-the-data/#::~text=The%20percentages%20are%20based%20on,for%20the%2030%20days%20prior.>**)

<https://cran.r-project.org/web/packages/gtrendsR/gtrendsR.pdf>

#### Inflation Data

In order to handle potential confounding variables that could cause spurious associations due to influence on the dependent and independent variable, it is important to also include a control variable that represents the general economic state. Therefore, the worldwide inflation rate is included as well, which was retrieved from the OECD. To allow a simple aggregation the mean inflation rate per month for every available country in the data set was calculated.

<https://data.oecd.org/price/inflation-cpi.htm>

#### Import Time Series Data

TODO: - Justify why Adjusted values and not close values - Timeframe: December 2021 until December 2023

```
# Ticker Data
ticker_data <- NULL
tickers_index <- c("XLK", "XLF", "XLV", "XLY", "XLE", "XLU")

for (Ticker in tickers_index){
  ticker_data <- cbind(ticker_data,
                      getSymbols.yahoo(Ticker, from="2021-12-01", to="2023-12-01",
                                       periodicity = "weekly", auto.assign=FALSE)[,6])
}

# Check length and min max date
length(ticker_data$XLK.Adjusted)
```

```
## [1] 105
```

```
summary(ticker_data)[c(1, 6)]
```

```
## [1] "Min.      :2021-11-29  " "Max.      :2023-11-27  "
```

```
head(ticker_data)
```

```
##           XLK.Adjusted XLF.Adjusted XLV.Adjusted XLY.Adjusted XLE.Adjusted
## 2021-11-29      162.1687      36.56124      126.3468      196.2840      50.84305
## 2021-12-06      171.7438      37.54185      130.4269      201.5340      52.77941
## 2021-12-13      164.8694      37.09962      133.6638      191.9975      50.13306
## 2021-12-20      169.8386      37.32073      135.0013      199.7643      50.62176
## 2021-12-27      171.0750      37.72171      137.0669      201.2943      51.83764
## 2022-01-03      163.2626      39.76960      130.7044      196.4106      57.29226
##           XLU.Adjusted
## 2021-11-29      63.42516
## 2021-12-06      65.07197
## 2021-12-13      65.88124
## 2021-12-20      65.60835
## 2021-12-27      67.88811
## 2022-01-03      66.77844
```

```
# Google Trends data
gtrends_war_web <- gtrends(
  keyword = "war",
  time = "2021-11-30 2023-12-01",
  gprop = "web"
)$interest_over_time
gtrends_war_web
```

```
# Too many requests exclude?
gtrends_war_news <- gtrends(
  keyword = "war",
  time = "2021-12-01 2023-12-01",
  gprop = "news"
)$interest_over_time
```

```
# If gtrends does not work --> response 429
library(lubridate)

gtrends_war_web <- read.csv('gtrends_war_web.csv')
gtrends_war_web <- gtrends_war_web %>%
  mutate(date = ymd(date))
head(gtrends_war_web)
```

```
##    X      date hits keyword   geo           time gprop category
## 1 1 2021-12-05  33    war world 2021-11-30 2023-12-01  web        0
## 2 2 2021-12-12  35    war world 2021-11-30 2023-12-01  web        0
## 3 3 2021-12-19  31    war world 2021-11-30 2023-12-01  web        0
## 4 4 2021-12-26  31    war world 2021-11-30 2023-12-01  web        0
## 5 5 2022-01-02  34    war world 2021-11-30 2023-12-01  web        0
## 6 6 2022-01-09  39    war world 2021-11-30 2023-12-01  web        0
```

Google Trends delivers data always from Sunday weekly. This has to be put to the weekly data from the financial tickers, which always starts at Monday.

```
# Increase date by one day to match ticker data
gtrends_war_web$date <- gtrends_war_web$date + days(1)

# Check length and min max date
length(gtrends_war_web$date)
```

```
## [1] 104
```

```
min(gtrends_war_web$date)
```

```
## [1] "2021-12-06"
```

```
max(gtrends_war_web$date)
```

```
## [1] "2023-11-27"
```

```
# Remove first date of ticker data to match gtrends data
ticker_data <- ticker_data[-1,]
ticker_data
```

##	XLK.Adjusted	XLF.Adjusted	XLV.Adjusted	XLY.Adjusted	XLE.Adjusted
## 2021-12-06	171.7438	37.54185	130.4269	201.5340	52.77941
## 2021-12-13	164.8694	37.09962	133.6638	191.9975	50.13306
## 2021-12-20	169.8386	37.32073	135.0013	199.7643	50.62176
## 2021-12-27	171.0750	37.72171	137.0669	201.2943	51.83764
## 2022-01-03	163.2626	39.76960	130.7044	196.4106	57.29226
## 2022-01-10	163.0954	39.44117	130.4319	193.4961	60.25308
## 2022-01-17	151.8491	36.90063	125.9568	177.6438	58.32901
## 2022-01-24	155.4601	37.40294	126.9394	175.3595	61.29917
## 2022-01-31	157.0344	38.73599	128.7100	181.5626	64.35338
## 2022-02-07	152.4001	38.74566	126.7156	177.7029	65.76374
## 2022-02-14	149.9895	37.88593	124.0500	177.2303	63.55947
## 2022-02-21	151.8688	37.77001	127.3966	173.4789	64.33469
## 2022-02-28	147.3428	35.96362	128.9240	168.9596	70.26568
## 2022-03-07	141.7442	35.16185	125.4411	164.1546	71.77877
## 2022-03-14	152.5772	37.67341	133.2727	179.0814	69.00476
## 2022-03-21	155.6766	38.17572	132.5723	180.5386	73.55339
## 2022-03-28	156.2008	37.06440	134.7424	182.6600	72.66340
## 2022-04-04	150.2371	36.74438	139.3877	176.9898	74.99246
## 2022-04-11	144.5493	35.77461	135.2987	175.9248	75.29420
## 2022-04-18	141.0401	35.07639	130.4485	173.3806	71.85245
## 2022-04-25	139.4038	33.46657	127.1499	160.6201	70.86237
## 2022-05-02	138.5659	33.68962	126.6522	155.9557	78.18905
## 2022-05-09	133.9230	32.51620	125.5007	150.2066	76.17114
## 2022-05-16	129.1915	31.92464	126.6522	138.4619	77.10466
## 2022-05-23	139.4432	34.49453	130.7900	151.6266	83.47898
## 2022-05-30	137.9942	33.81569	126.6912	150.8772	84.40306
## 2022-06-06	129.2210	31.54644	122.4656	142.4655	83.64870

## 2022-06-13	122.8433	29.99482	117.0006	134.2117	69.29707
## 2022-06-20	131.5474	31.36218	126.0862	144.9506	67.50547
## 2022-06-27	125.9199	31.08089	127.0889	138.3867	69.20649
## 2022-07-04	131.3160	31.27588	128.1277	144.6730	67.65224
## 2022-07-11	130.8318	30.98340	127.5985	143.1607	65.40193
## 2022-07-18	135.6151	31.92909	127.2359	152.9558	67.84296
## 2022-07-25	142.5333	32.86502	129.7545	160.9421	74.77505
## 2022-08-01	145.3005	32.84552	128.8235	162.4050	69.68325
## 2022-08-08	148.8485	34.66865	130.9501	167.9400	74.82272
## 2022-08-15	146.3778	34.08369	130.2641	165.5876	75.76670
## 2022-08-22	138.2440	32.87477	124.7466	157.8286	78.98959
## 2022-08-29	131.2963	32.06557	122.4926	153.6872	76.24346
## 2022-09-05	135.5954	33.49873	127.8925	162.5730	76.85371
## 2022-09-12	127.2047	32.23132	124.9034	156.1088	74.79412
## 2022-09-19	122.2928	30.27170	120.2778	144.7224	67.20409
## 2022-09-26	117.6806	29.75470	119.1632	141.0932	69.42124
## 2022-10-03	119.6522	30.27413	120.6490	139.3104	78.84836
## 2022-10-10	115.7190	30.39174	121.9084	133.9816	77.38319
## 2022-10-17	123.1892	31.54821	124.5059	141.0437	83.79324
## 2022-10-24	128.3807	33.50834	130.7243	143.5793	86.02953
## 2022-10-31	119.8702	33.23392	128.7073	136.1904	88.12122
## 2022-11-07	131.9077	35.15484	130.9900	143.7081	89.76952
## 2022-11-14	130.8675	34.66481	132.4069	139.6471	88.32364
## 2022-11-21	132.3635	35.39006	134.9454	141.2715	88.55499
## 2022-11-28	134.1072	35.21364	137.5331	144.5698	87.05128
## 2022-12-05	129.6984	33.84156	135.7621	138.1020	79.69659
## 2022-12-12	126.3001	33.02811	133.3514	132.6346	81.31598
## 2022-12-19	123.4468	33.27312	133.9123	128.1972	83.91855
## 2022-12-26	123.6423	33.72831	134.2169	128.2804	85.18214
## 2023-01-02	123.9304	34.89204	134.0391	131.1706	85.26978
## 2023-01-09	129.6535	35.62183	133.8217	138.7486	87.59727
## 2023-01-16	130.4980	34.86245	132.3002	138.0335	88.11341
## 2023-01-23	135.8237	35.75004	131.2628	146.8829	88.84380
## 2023-01-30	140.8612	36.08535	131.0949	150.3193	83.71163
## 2023-02-06	139.4801	35.98673	130.8973	147.1411	87.85046
## 2023-02-13	138.9237	35.88811	130.4033	149.5446	82.28008
## 2023-02-20	135.2176	35.16818	126.9651	142.8803	82.44564
## 2023-02-27	139.2416	35.49363	127.6172	145.3136	84.97763
## 2023-03-06	134.8996	32.47583	122.6575	137.2589	80.46873
## 2023-03-13	142.5403	30.55272	124.3470	140.3775	74.95677
## 2023-03-20	145.0243	30.56259	125.7697	140.6158	75.85271
## 2023-03-27	150.3732	31.86395	128.4026	148.9748	81.50535
## 2023-04-03	148.4413	31.70538	132.4294	144.3823	83.62096
## 2023-04-10	148.0231	32.58746	133.4709	146.3448	85.83498
## 2023-04-17	147.1369	32.92443	133.2031	146.8130	83.62096
## 2023-04-24	150.1939	32.87487	132.4393	147.3012	83.76856
## 2023-05-01	150.6022	32.04235	132.4989	146.6337	78.94692
## 2023-05-08	150.3134	31.61618	131.1103	147.2713	77.26427
## 2023-05-15	156.8159	32.30995	130.2375	150.9772	78.36636
## 2023-05-22	164.0851	31.82431	126.4884	151.3957	77.51027
## 2023-05-29	166.1961	32.50817	129.2556	156.4066	78.62221
## 2023-06-05	165.3099	32.85506	129.3548	160.5908	80.02934
## 2023-06-12	172.4596	33.28123	131.1301	165.7512	79.56685
## 2023-06-19	167.9388	32.44871	130.9219	164.8247	76.12283

##	2023-06-26	173.4894	33.55998	132.1878	169.5085	80.57540
##	2023-07-03	170.8451	33.44052	128.4730	169.0693	80.19819
##	2023-07-10	175.6149	34.09758	131.1620	174.6094	80.85336
##	2023-07-17	175.7546	35.10308	135.6835	170.6764	83.68248
##	2023-07-24	177.5607	35.04335	134.6378	172.4533	85.22112
##	2023-07-31	170.7952	34.75464	131.8591	171.2454	86.28329
##	2023-08-07	166.5443	34.76460	135.1158	169.4187	89.24146
##	2023-08-14	164.6084	33.82878	132.9845	162.3712	88.15944
##	2023-08-21	168.3504	33.83874	132.8650	164.3677	86.94837
##	2023-08-28	175.8244	34.53562	133.0244	169.3688	90.07530
##	2023-09-04	172.4217	34.14736	131.5205	168.4504	91.37571
##	2023-09-11	168.5400	34.67500	131.6500	171.4850	91.33600
##	2023-09-18	164.0596	33.51020	129.6183	160.6443	88.65578
##	2023-09-25	163.9300	33.17000	128.7400	160.9800	90.39000
##	2023-10-02	168.2200	33.04000	130.0100	160.6100	85.73000
##	2023-10-09	168.6300	33.21000	130.1900	158.9700	89.59000
##	2023-10-16	163.9400	32.20000	128.0600	151.6900	90.26000
##	2023-10-23	161.1200	31.45000	123.1400	149.5900	84.63000
##	2023-10-30	171.7600	33.78000	127.4100	160.2200	86.68000
##	2023-11-06	179.5200	33.91000	126.2700	161.4000	83.41000
##	2023-11-13	182.8000	35.01000	128.2500	167.1400	84.69000
##	2023-11-20	184.4100	35.38000	131.1300	168.2400	84.93000
##	2023-11-27	185.1600	35.90000	131.3100	168.7800	84.58000
##	XLU.Adjusted					
##	2021-12-06	65.07197				
##	2021-12-13	65.88124				
##	2021-12-20	65.60835				
##	2021-12-27	67.88811				
##	2022-01-03	66.77844				
##	2022-01-10	65.82055				
##	2022-01-17	65.28943				
##	2022-01-24	64.45482				
##	2022-01-31	64.94800				
##	2022-02-07	63.57278				
##	2022-02-14	62.80456				
##	2022-02-21	64.09441				
##	2022-02-28	67.23370				
##	2022-03-07	66.79741				
##	2022-03-14	67.15781				
##	2022-03-21	69.02621				
##	2022-03-28	72.08438				
##	2022-04-04	73.47834				
##	2022-04-11	72.64769				
##	2022-04-18	70.89095				
##	2022-04-25	68.03621				
##	2022-05-02	68.94323				
##	2022-05-09	68.19850				
##	2022-05-16	68.49449				
##	2022-05-23	71.96027				
##	2022-05-30	70.98640				
##	2022-06-06	68.11257				
##	2022-06-13	61.93529				
##	2022-06-20	65.88799				
##	2022-06-27	69.14848				



##	2022-07-04	67.19480
##	2022-07-11	67.13706
##	2022-07-18	66.81947
##	2022-07-25	71.16952
##	2022-08-01	71.43900
##	2022-08-08	73.83536
##	2022-08-15	74.77852
##	2022-08-22	72.86333
##	2022-08-29	71.79507
##	2022-09-05	74.47055
##	2022-09-12	71.66998
##	2022-09-19	69.06187
##	2022-09-26	63.47388
##	2022-10-03	61.77827
##	2022-10-10	60.21831
##	2022-10-17	61.34226
##	2022-10-24	65.31483
##	2022-10-31	65.00476
##	2022-11-07	65.96400
##	2022-11-14	66.71007
##	2022-11-21	68.73512
##	2022-11-28	68.82231
##	2022-12-05	68.62853
##	2022-12-12	68.28941
##	2022-12-19	68.70604
##	2022-12-26	68.86770
##	2023-01-02	69.35612
##	2023-01-09	69.67847
##	2023-01-16	67.62710
##	2023-01-23	67.29497
##	2023-01-30	66.33767
##	2023-02-06	66.14230
##	2023-02-13	66.89447
##	2023-02-20	65.06777
##	2023-02-27	64.71609
##	2023-03-06	62.92846
##	2023-03-13	65.41943
##	2023-03-20	64.16907
##	2023-03-27	66.61012
##	2023-04-03	68.69630
##	2023-04-10	67.79098
##	2023-04-17	68.50934
##	2023-04-24	67.87955
##	2023-05-01	67.92875
##	2023-05-08	67.92875
##	2023-05-15	65.05534
##	2023-05-22	63.55957
##	2023-05-29	64.08112
##	2023-06-05	65.33086
##	2023-06-12	66.22635
##	2023-06-19	63.99256
##	2023-06-26	64.91399
##	2023-07-03	64.86439
##	2023-07-10	66.31264

```
## 2023-07-17      67.90971
## 2023-07-24      66.50112
## 2023-07-31      63.47564
## 2023-08-07      64.05097
## 2023-08-14      62.97966
## 2023-08-21      63.11853
## 2023-08-28      62.13649
## 2023-09-04      62.67215
## 2023-09-11      64.41801
## 2023-09-18      62.78127
## 2023-09-25      58.93000
## 2023-10-02      57.25000
## 2023-10-09      59.30000
## 2023-10-16      58.03000
## 2023-10-23      58.73000
## 2023-10-30      61.86000
## 2023-11-06      60.32000
## 2023-11-13      62.29000
## 2023-11-20      62.69000
## 2023-11-27      62.76000
```

## Visualize Ticker Data

```
# Visualization
ticker_data_df <- fortify.zoo(ticker_data)
colnames(ticker_data_df)[1] <- "Date"
ticker_data_df <- gather(ticker_data_df, key="Ticker", value="Value",
  c("XLK.Adjusted", "XLF.Adjusted", "XLV.Adjusted",
    "XLY.Adjusted", "XLE.Adjusted", "XLU.Adjusted"), -Date)

ticker_data_df
```

```
##           Date      Ticker      Value
## 1  2021-12-06  XLK.Adjusted 171.74384
## 2  2021-12-13  XLK.Adjusted 164.86938
## 3  2021-12-20  XLK.Adjusted 169.83864
## 4  2021-12-27  XLK.Adjusted 171.07500
## 5  2022-01-03  XLK.Adjusted 163.26263
## 6  2022-01-10  XLK.Adjusted 163.09538
## 7  2022-01-17  XLK.Adjusted 151.84909
## 8  2022-01-24  XLK.Adjusted 155.46013
## 9  2022-01-31  XLK.Adjusted 157.03439
## 10 2022-02-07  XLK.Adjusted 152.40012
## 11 2022-02-14  XLK.Adjusted 149.98950
## 12 2022-02-21  XLK.Adjusted 151.86882
## 13 2022-02-28  XLK.Adjusted 147.34276
## 14 2022-03-07  XLK.Adjusted 141.74422
## 15 2022-03-14  XLK.Adjusted 152.57724
## 16 2022-03-21  XLK.Adjusted 155.67657
## 17 2022-03-28  XLK.Adjusted 156.20082
## 18 2022-04-04  XLK.Adjusted 150.23706
## 19 2022-04-11  XLK.Adjusted 144.54935
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##	21	2022-04-25	XLK.Adjusted	139.40376
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##	23	2022-05-09	XLK.Adjusted	133.92300
##	24	2022-05-16	XLK.Adjusted	129.19145
##	25	2022-05-23	XLK.Adjusted	139.44318
##	26	2022-05-30	XLK.Adjusted	137.99416
##	27	2022-06-06	XLK.Adjusted	129.22102
##	28	2022-06-13	XLK.Adjusted	122.84327
##	29	2022-06-20	XLK.Adjusted	131.54738
##	30	2022-06-27	XLK.Adjusted	125.91988
##	31	2022-07-04	XLK.Adjusted	131.31599
##	32	2022-07-11	XLK.Adjusted	130.83176
##	33	2022-07-18	XLK.Adjusted	135.61513
##	34	2022-07-25	XLK.Adjusted	142.53326
##	35	2022-08-01	XLK.Adjusted	145.30052
##	36	2022-08-08	XLK.Adjusted	148.84853
##	37	2022-08-15	XLK.Adjusted	146.37778
##	38	2022-08-22	XLK.Adjusted	138.24402
##	39	2022-08-29	XLK.Adjusted	131.29626
##	40	2022-09-05	XLK.Adjusted	135.59538
##	41	2022-09-12	XLK.Adjusted	127.20467
##	42	2022-09-19	XLK.Adjusted	122.29279
##	43	2022-09-26	XLK.Adjusted	117.68064
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##	45	2022-10-10	XLK.Adjusted	115.71896
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##	47	2022-10-24	XLK.Adjusted	128.38069
##	48	2022-10-31	XLK.Adjusted	119.87019
##	49	2022-11-07	XLK.Adjusted	131.90773
##	50	2022-11-14	XLK.Adjusted	130.86746
##	51	2022-11-21	XLK.Adjusted	132.36346
##	52	2022-11-28	XLK.Adjusted	134.10719
##	53	2022-12-05	XLK.Adjusted	129.69838
##	54	2022-12-12	XLK.Adjusted	126.30012
##	55	2022-12-19	XLK.Adjusted	123.44678
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##	57	2023-01-02	XLK.Adjusted	123.93041
##	58	2023-01-09	XLK.Adjusted	129.65347
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##	62	2023-02-06	XLK.Adjusted	139.48009
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##	64	2023-02-20	XLK.Adjusted	135.21759
##	65	2023-02-27	XLK.Adjusted	139.24161
##	66	2023-03-06	XLK.Adjusted	134.89964
##	67	2023-03-13	XLK.Adjusted	142.54033
##	68	2023-03-20	XLK.Adjusted	145.02431
##	69	2023-03-27	XLK.Adjusted	150.37315
##	70	2023-04-03	XLK.Adjusted	148.44135
##	71	2023-04-10	XLK.Adjusted	148.02310
##	72	2023-04-17	XLK.Adjusted	147.13686
##	73	2023-04-24	XLK.Adjusted	150.19392

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## 77	2023-05-22	XLK.Adjusted	164.08508
## 78	2023-05-29	XLK.Adjusted	166.19614
## 79	2023-06-05	XLK.Adjusted	165.30989
## 80	2023-06-12	XLK.Adjusted	172.45963
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## 82	2023-06-26	XLK.Adjusted	173.48943
## 83	2023-07-03	XLK.Adjusted	170.84508
## 84	2023-07-10	XLK.Adjusted	175.61488
## 85	2023-07-17	XLK.Adjusted	175.75459
## 86	2023-07-24	XLK.Adjusted	177.56073
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## 88	2023-08-07	XLK.Adjusted	166.54425
## 89	2023-08-14	XLK.Adjusted	164.60840
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## 91	2023-08-28	XLK.Adjusted	175.82443
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## 93	2023-09-11	XLK.Adjusted	168.53999
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## 95	2023-09-25	XLK.Adjusted	163.92999
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## 97	2023-10-09	XLK.Adjusted	168.63000
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## 99	2023-10-23	XLK.Adjusted	161.12000
## 100	2023-10-30	XLK.Adjusted	171.75999
## 101	2023-11-06	XLK.Adjusted	179.52000
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## 103	2023-11-20	XLK.Adjusted	184.41000
## 104	2023-11-27	XLK.Adjusted	185.16000
## 105	2021-12-06	XLF.Adjusted	37.54185
## 106	2021-12-13	XLF.Adjusted	37.09962
## 107	2021-12-20	XLF.Adjusted	37.32073
## 108	2021-12-27	XLF.Adjusted	37.72171
## 109	2022-01-03	XLF.Adjusted	39.76960
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## 111	2022-01-17	XLF.Adjusted	36.90063
## 112	2022-01-24	XLF.Adjusted	37.40294
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## 115	2022-02-14	XLF.Adjusted	37.88593
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## 123	2022-04-11	XLF.Adjusted	35.77461
## 124	2022-04-18	XLF.Adjusted	35.07639
## 125	2022-04-25	XLF.Adjusted	33.46657
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## 127	2022-05-09	XLF.Adjusted	32.51620

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##	131	2022-06-06	XLF.Adjusted	31.54644
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##	133	2022-06-20	XLF.Adjusted	31.36218
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##	135	2022-07-04	XLF.Adjusted	31.27588
##	136	2022-07-11	XLF.Adjusted	30.98340
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##	138	2022-07-25	XLF.Adjusted	32.86502
##	139	2022-08-01	XLF.Adjusted	32.84552
##	140	2022-08-08	XLF.Adjusted	34.66865
##	141	2022-08-15	XLF.Adjusted	34.08369
##	142	2022-08-22	XLF.Adjusted	32.87477
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##	144	2022-09-05	XLF.Adjusted	33.49873
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##	146	2022-09-19	XLF.Adjusted	30.27170
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##	150	2022-10-17	XLF.Adjusted	31.54821
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##	156	2022-11-28	XLF.Adjusted	35.21364
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## 245 2022-08-15 XLV.Adjusted 130.26414
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## 253 2022-10-10 XLV.Adjusted 121.90838
## 254 2022-10-17 XLV.Adjusted 124.50594
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## 258 2022-11-14 XLV.Adjusted 132.40688
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## 261 2022-12-05 XLV.Adjusted 135.76205
## 262 2022-12-12 XLV.Adjusted 133.35144
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## 270 2023-02-06 XLV.Adjusted 130.89729
## 271 2023-02-13 XLV.Adjusted 130.40331
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## 282 2023-05-01 XLV.Adjusted 132.49886
## 283 2023-05-08 XLV.Adjusted 131.11031
## 284 2023-05-15 XLV.Adjusted 130.23747
## 285 2023-05-22 XLV.Adjusted 126.48836
## 286 2023-05-29 XLV.Adjusted 129.25557
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## 289 2023-06-19 XLV.Adjusted 130.92186

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## 328 2022-03-21 XLY.Adjusted 180.53859
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## 338 2022-05-30 XLY.Adjusted 150.87723
## 339 2022-06-06 XLY.Adjusted 142.46555
## 340 2022-06-13 XLY.Adjusted 134.21169
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## 342 2022-06-27 XLY.Adjusted 138.38669
## 343 2022-07-04 XLY.Adjusted 144.67297

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## 344 2022-07-11 XLY.Adjusted 143.16069  
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## 382 2023-04-03 XLY.Adjusted 144.38226  
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## 384 2023-04-17 XLY.Adjusted 146.81305  
## 385 2023-04-24 XLY.Adjusted 147.30119  
## 386 2023-05-01 XLY.Adjusted 146.63373  
## 387 2023-05-08 XLY.Adjusted 147.27130  
## 388 2023-05-15 XLY.Adjusted 150.97725  
## 389 2023-05-22 XLY.Adjusted 151.39566  
## 390 2023-05-29 XLY.Adjusted 156.40665  
## 391 2023-06-05 XLY.Adjusted 160.59076  
## 392 2023-06-12 XLY.Adjusted 165.75121  
## 393 2023-06-19 XLY.Adjusted 164.82471  
## 394 2023-06-26 XLY.Adjusted 169.50851  
## 395 2023-07-03 XLY.Adjusted 169.06929  
## 396 2023-07-10 XLY.Adjusted 174.60944  
## 397 2023-07-17 XLY.Adjusted 170.67644

```

## 398 2023-07-24 XLY.Adjusted 172.45326
## 399 2023-07-31 XLY.Adjusted 171.24542
## 400 2023-08-07 XLY.Adjusted 169.41867
## 401 2023-08-14 XLY.Adjusted 162.37122
## 402 2023-08-21 XLY.Adjusted 164.36766
## 403 2023-08-28 XLY.Adjusted 169.36876
## 404 2023-09-04 XLY.Adjusted 168.45039
## 405 2023-09-11 XLY.Adjusted 171.48499
## 406 2023-09-18 XLY.Adjusted 160.64427
## 407 2023-09-25 XLY.Adjusted 160.98000
## 408 2023-10-02 XLY.Adjusted 160.61000
## 409 2023-10-09 XLY.Adjusted 158.97000
## 410 2023-10-16 XLY.Adjusted 151.69000
## 411 2023-10-23 XLY.Adjusted 149.59000
## 412 2023-10-30 XLY.Adjusted 160.22000
## 413 2023-11-06 XLY.Adjusted 161.39999
## 414 2023-11-13 XLY.Adjusted 167.14000
## 415 2023-11-20 XLY.Adjusted 168.24001
## 416 2023-11-27 XLY.Adjusted 168.78000
## 417 2021-12-06 XLE.Adjusted 52.77941
## 418 2021-12-13 XLE.Adjusted 50.13306
## 419 2021-12-20 XLE.Adjusted 50.62176
## 420 2021-12-27 XLE.Adjusted 51.83764
## 421 2022-01-03 XLE.Adjusted 57.29226
## 422 2022-01-10 XLE.Adjusted 60.25308
## 423 2022-01-17 XLE.Adjusted 58.32901
## 424 2022-01-24 XLE.Adjusted 61.29917
## 425 2022-01-31 XLE.Adjusted 64.35338
## 426 2022-02-07 XLE.Adjusted 65.76374
## 427 2022-02-14 XLE.Adjusted 63.55947
## 428 2022-02-21 XLE.Adjusted 64.33469
## 429 2022-02-28 XLE.Adjusted 70.26568
## 430 2022-03-07 XLE.Adjusted 71.77877
## 431 2022-03-14 XLE.Adjusted 69.00476
## 432 2022-03-21 XLE.Adjusted 73.55339
## 433 2022-03-28 XLE.Adjusted 72.66340
## 434 2022-04-04 XLE.Adjusted 74.99246
## 435 2022-04-11 XLE.Adjusted 75.29420
## 436 2022-04-18 XLE.Adjusted 71.85245
## 437 2022-04-25 XLE.Adjusted 70.86237
## 438 2022-05-02 XLE.Adjusted 78.18905
## 439 2022-05-09 XLE.Adjusted 76.17114
## 440 2022-05-16 XLE.Adjusted 77.10466
## 441 2022-05-23 XLE.Adjusted 83.47898
## 442 2022-05-30 XLE.Adjusted 84.40306
## 443 2022-06-06 XLE.Adjusted 83.64870
## 444 2022-06-13 XLE.Adjusted 69.29707
## 445 2022-06-20 XLE.Adjusted 67.50547
## 446 2022-06-27 XLE.Adjusted 69.20649
## 447 2022-07-04 XLE.Adjusted 67.65224
## 448 2022-07-11 XLE.Adjusted 65.40193
## 449 2022-07-18 XLE.Adjusted 67.84296
## 450 2022-07-25 XLE.Adjusted 74.77505
## 451 2022-08-01 XLE.Adjusted 69.68325

```

##	452	2022-08-08	XLE.Adjusted	74.82272
##	453	2022-08-15	XLE.Adjusted	75.76670
##	454	2022-08-22	XLE.Adjusted	78.98959
##	455	2022-08-29	XLE.Adjusted	76.24346
##	456	2022-09-05	XLE.Adjusted	76.85371
##	457	2022-09-12	XLE.Adjusted	74.79412
##	458	2022-09-19	XLE.Adjusted	67.20409
##	459	2022-09-26	XLE.Adjusted	69.42124
##	460	2022-10-03	XLE.Adjusted	78.84836
##	461	2022-10-10	XLE.Adjusted	77.38319
##	462	2022-10-17	XLE.Adjusted	83.79324
##	463	2022-10-24	XLE.Adjusted	86.02953
##	464	2022-10-31	XLE.Adjusted	88.12122
##	465	2022-11-07	XLE.Adjusted	89.76952
##	466	2022-11-14	XLE.Adjusted	88.32364
##	467	2022-11-21	XLE.Adjusted	88.55499
##	468	2022-11-28	XLE.Adjusted	87.05128
##	469	2022-12-05	XLE.Adjusted	79.69659
##	470	2022-12-12	XLE.Adjusted	81.31598
##	471	2022-12-19	XLE.Adjusted	83.91855
##	472	2022-12-26	XLE.Adjusted	85.18214
##	473	2023-01-02	XLE.Adjusted	85.26978
##	474	2023-01-09	XLE.Adjusted	87.59727
##	475	2023-01-16	XLE.Adjusted	88.11341
##	476	2023-01-23	XLE.Adjusted	88.84380
##	477	2023-01-30	XLE.Adjusted	83.71163
##	478	2023-02-06	XLE.Adjusted	87.85046
##	479	2023-02-13	XLE.Adjusted	82.28008
##	480	2023-02-20	XLE.Adjusted	82.44564
##	481	2023-02-27	XLE.Adjusted	84.97763
##	482	2023-03-06	XLE.Adjusted	80.46873
##	483	2023-03-13	XLE.Adjusted	74.95677
##	484	2023-03-20	XLE.Adjusted	75.85271
##	485	2023-03-27	XLE.Adjusted	81.50535
##	486	2023-04-03	XLE.Adjusted	83.62096
##	487	2023-04-10	XLE.Adjusted	85.83498
##	488	2023-04-17	XLE.Adjusted	83.62096
##	489	2023-04-24	XLE.Adjusted	83.76856
##	490	2023-05-01	XLE.Adjusted	78.94692
##	491	2023-05-08	XLE.Adjusted	77.26427
##	492	2023-05-15	XLE.Adjusted	78.36636
##	493	2023-05-22	XLE.Adjusted	77.51027
##	494	2023-05-29	XLE.Adjusted	78.62221
##	495	2023-06-05	XLE.Adjusted	80.02934
##	496	2023-06-12	XLE.Adjusted	79.56685
##	497	2023-06-19	XLE.Adjusted	76.12283
##	498	2023-06-26	XLE.Adjusted	80.57540
##	499	2023-07-03	XLE.Adjusted	80.19819
##	500	2023-07-10	XLE.Adjusted	80.85336
##	501	2023-07-17	XLE.Adjusted	83.68248
##	502	2023-07-24	XLE.Adjusted	85.22112
##	503	2023-07-31	XLE.Adjusted	86.28329
##	504	2023-08-07	XLE.Adjusted	89.24146
##	505	2023-08-14	XLE.Adjusted	88.15944

##	506	2023-08-21	XLE.Adjusted	86.94837
##	507	2023-08-28	XLE.Adjusted	90.07530
##	508	2023-09-04	XLE.Adjusted	91.37571
##	509	2023-09-11	XLE.Adjusted	91.33600
##	510	2023-09-18	XLE.Adjusted	88.65578
##	511	2023-09-25	XLE.Adjusted	90.39000
##	512	2023-10-02	XLE.Adjusted	85.73000
##	513	2023-10-09	XLE.Adjusted	89.59000
##	514	2023-10-16	XLE.Adjusted	90.26000
##	515	2023-10-23	XLE.Adjusted	84.63000
##	516	2023-10-30	XLE.Adjusted	86.68000
##	517	2023-11-06	XLE.Adjusted	83.41000
##	518	2023-11-13	XLE.Adjusted	84.69000
##	519	2023-11-20	XLE.Adjusted	84.93000
##	520	2023-11-27	XLE.Adjusted	84.58000
##	521	2021-12-06	XLU.Adjusted	65.07197
##	522	2021-12-13	XLU.Adjusted	65.88124
##	523	2021-12-20	XLU.Adjusted	65.60835
##	524	2021-12-27	XLU.Adjusted	67.88811
##	525	2022-01-03	XLU.Adjusted	66.77844
##	526	2022-01-10	XLU.Adjusted	65.82055
##	527	2022-01-17	XLU.Adjusted	65.28943
##	528	2022-01-24	XLU.Adjusted	64.45482
##	529	2022-01-31	XLU.Adjusted	64.94800
##	530	2022-02-07	XLU.Adjusted	63.57278
##	531	2022-02-14	XLU.Adjusted	62.80456
##	532	2022-02-21	XLU.Adjusted	64.09441
##	533	2022-02-28	XLU.Adjusted	67.23370
##	534	2022-03-07	XLU.Adjusted	66.79741
##	535	2022-03-14	XLU.Adjusted	67.15781
##	536	2022-03-21	XLU.Adjusted	69.02621
##	537	2022-03-28	XLU.Adjusted	72.08438
##	538	2022-04-04	XLU.Adjusted	73.47834
##	539	2022-04-11	XLU.Adjusted	72.64769
##	540	2022-04-18	XLU.Adjusted	70.89095
##	541	2022-04-25	XLU.Adjusted	68.03621
##	542	2022-05-02	XLU.Adjusted	68.94323
##	543	2022-05-09	XLU.Adjusted	68.19850
##	544	2022-05-16	XLU.Adjusted	68.49449
##	545	2022-05-23	XLU.Adjusted	71.96027
##	546	2022-05-30	XLU.Adjusted	70.98640
##	547	2022-06-06	XLU.Adjusted	68.11257
##	548	2022-06-13	XLU.Adjusted	61.93529
##	549	2022-06-20	XLU.Adjusted	65.88799
##	550	2022-06-27	XLU.Adjusted	69.14848
##	551	2022-07-04	XLU.Adjusted	67.19480
##	552	2022-07-11	XLU.Adjusted	67.13706
##	553	2022-07-18	XLU.Adjusted	66.81947
##	554	2022-07-25	XLU.Adjusted	71.16952
##	555	2022-08-01	XLU.Adjusted	71.43900
##	556	2022-08-08	XLU.Adjusted	73.83536
##	557	2022-08-15	XLU.Adjusted	74.77852
##	558	2022-08-22	XLU.Adjusted	72.86333
##	559	2022-08-29	XLU.Adjusted	71.79507

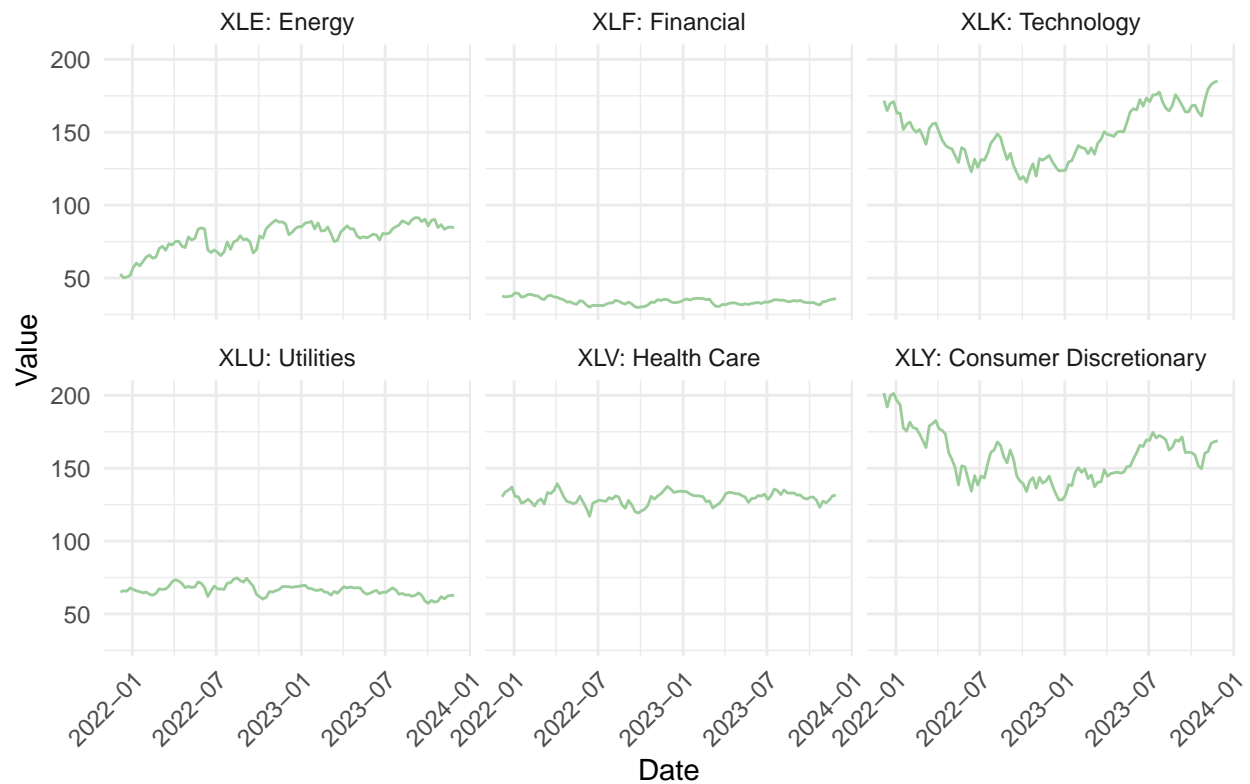
##	560	2022-09-05	XLU.Adjusted	74.47055
##	561	2022-09-12	XLU.Adjusted	71.66998
##	562	2022-09-19	XLU.Adjusted	69.06187
##	563	2022-09-26	XLU.Adjusted	63.47388
##	564	2022-10-03	XLU.Adjusted	61.77827
##	565	2022-10-10	XLU.Adjusted	60.21831
##	566	2022-10-17	XLU.Adjusted	61.34226
##	567	2022-10-24	XLU.Adjusted	65.31483
##	568	2022-10-31	XLU.Adjusted	65.00476
##	569	2022-11-07	XLU.Adjusted	65.96400
##	570	2022-11-14	XLU.Adjusted	66.71007
##	571	2022-11-21	XLU.Adjusted	68.73512
##	572	2022-11-28	XLU.Adjusted	68.82231
##	573	2022-12-05	XLU.Adjusted	68.62853
##	574	2022-12-12	XLU.Adjusted	68.28941
##	575	2022-12-19	XLU.Adjusted	68.70604
##	576	2022-12-26	XLU.Adjusted	68.86770
##	577	2023-01-02	XLU.Adjusted	69.35612
##	578	2023-01-09	XLU.Adjusted	69.67847
##	579	2023-01-16	XLU.Adjusted	67.62710
##	580	2023-01-23	XLU.Adjusted	67.29497
##	581	2023-01-30	XLU.Adjusted	66.33767
##	582	2023-02-06	XLU.Adjusted	66.14230
##	583	2023-02-13	XLU.Adjusted	66.89447
##	584	2023-02-20	XLU.Adjusted	65.06777
##	585	2023-02-27	XLU.Adjusted	64.71609
##	586	2023-03-06	XLU.Adjusted	62.92846
##	587	2023-03-13	XLU.Adjusted	65.41943
##	588	2023-03-20	XLU.Adjusted	64.16907
##	589	2023-03-27	XLU.Adjusted	66.61012
##	590	2023-04-03	XLU.Adjusted	68.69630
##	591	2023-04-10	XLU.Adjusted	67.79098
##	592	2023-04-17	XLU.Adjusted	68.50934
##	593	2023-04-24	XLU.Adjusted	67.87955
##	594	2023-05-01	XLU.Adjusted	67.92875
##	595	2023-05-08	XLU.Adjusted	67.92875
##	596	2023-05-15	XLU.Adjusted	65.05534
##	597	2023-05-22	XLU.Adjusted	63.55957
##	598	2023-05-29	XLU.Adjusted	64.08112
##	599	2023-06-05	XLU.Adjusted	65.33086
##	600	2023-06-12	XLU.Adjusted	66.22635
##	601	2023-06-19	XLU.Adjusted	63.99256
##	602	2023-06-26	XLU.Adjusted	64.91399
##	603	2023-07-03	XLU.Adjusted	64.86439
##	604	2023-07-10	XLU.Adjusted	66.31264
##	605	2023-07-17	XLU.Adjusted	67.90971
##	606	2023-07-24	XLU.Adjusted	66.50112
##	607	2023-07-31	XLU.Adjusted	63.47564
##	608	2023-08-07	XLU.Adjusted	64.05097
##	609	2023-08-14	XLU.Adjusted	62.97966
##	610	2023-08-21	XLU.Adjusted	63.11853
##	611	2023-08-28	XLU.Adjusted	62.13649
##	612	2023-09-04	XLU.Adjusted	62.67215
##	613	2023-09-11	XLU.Adjusted	64.41801

```
## 614 2023-09-18 XLU.Adjusted 62.78127
## 615 2023-09-25 XLU.Adjusted 58.93000
## 616 2023-10-02 XLU.Adjusted 57.25000
## 617 2023-10-09 XLU.Adjusted 59.30000
## 618 2023-10-16 XLU.Adjusted 58.03000
## 619 2023-10-23 XLU.Adjusted 58.73000
## 620 2023-10-30 XLU.Adjusted 61.86000
## 621 2023-11-06 XLU.Adjusted 60.32000
## 622 2023-11-13 XLU.Adjusted 62.29000
## 623 2023-11-20 XLU.Adjusted 62.69000
## 624 2023-11-27 XLU.Adjusted 62.76000
```

```
# Line Chart
# New facet label names for ticker_data
ticker.labs <- c("XLK: Technology", "XLF: Financial", "XLV: Health Care",
                 "XLY: Consumer Discretionary", "XLE: Energy", "XLU: Utilities")
names(ticker.labs) <- c("XLK.Adjusted", "XLF.Adjusted", "XLV.Adjusted",
                       "XLY.Adjusted", "XLE.Adjusted", "XLU.Adjusted")

# Create the plot
ggplot(ticker_data_df, aes(Date, Value)) +
  geom_line(color='darkseagreen3') +
  facet_wrap(~Ticker, labeller = labeller(Ticker = ticker.labs)) +
  ggtitle("Ticker Time Series") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        plot.title = element_text(hjust = 0.5))
```

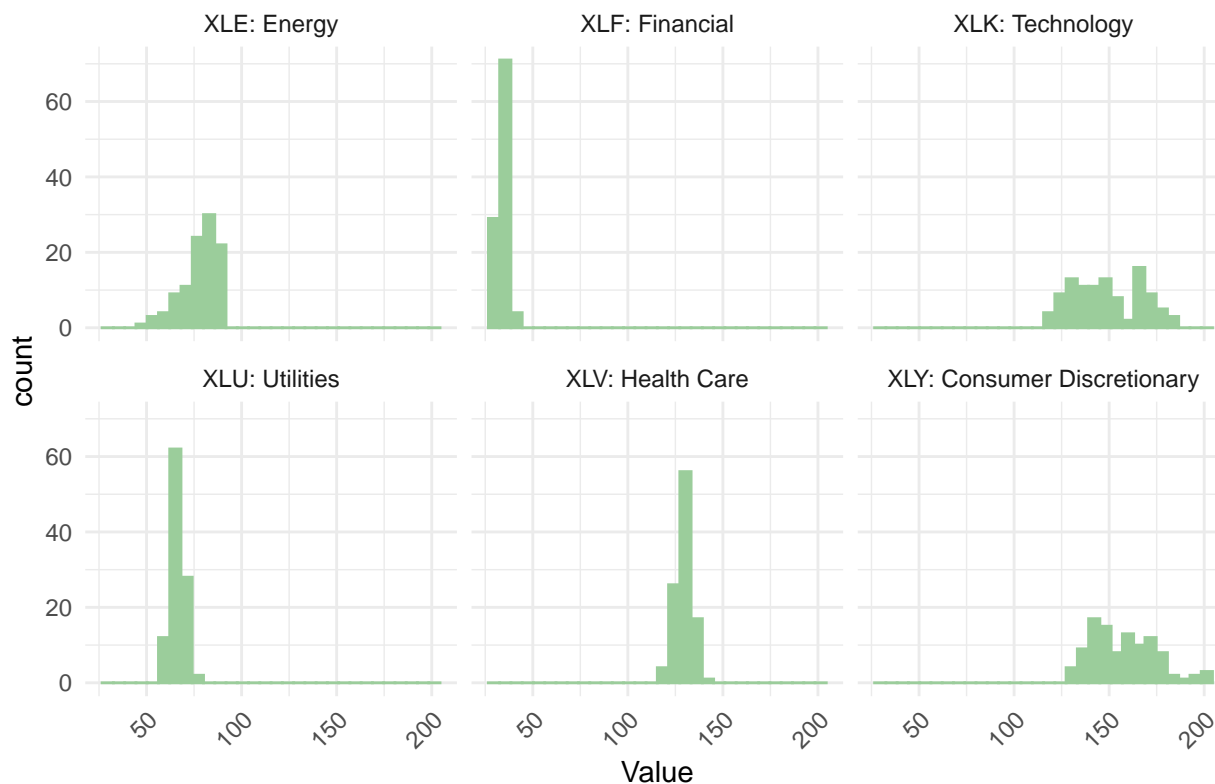
## Ticker Time Series



```
# Histogram
ggplot(ticker_data_df, aes(Value)) +
  geom_histogram(color='darkseagreen3', fill='darkseagreen3') +
  facet_wrap(~Ticker, labeller = labeller(Ticker = ticker.labs)) +
  ggtitle("Ticker Distributions") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        plot.title = element_text(hjust = 0.5))
```

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

## Ticker Distributions



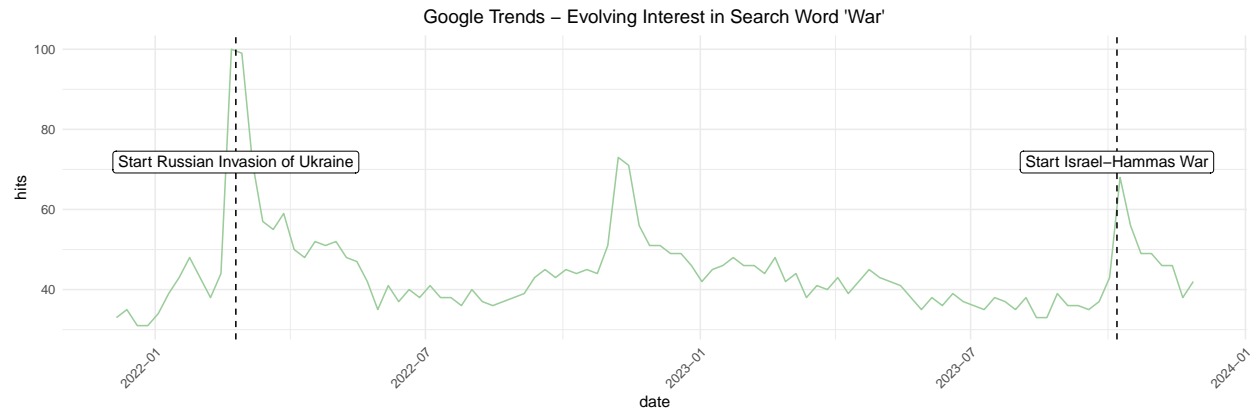
## Visualize Google Trends Data

```
# Line chart

# Convert 'date' column to POSIXct format
gtrends_war_web$date <- as.POSIXct(gtrends_war_web$date, format = "%Y-%m-%d")

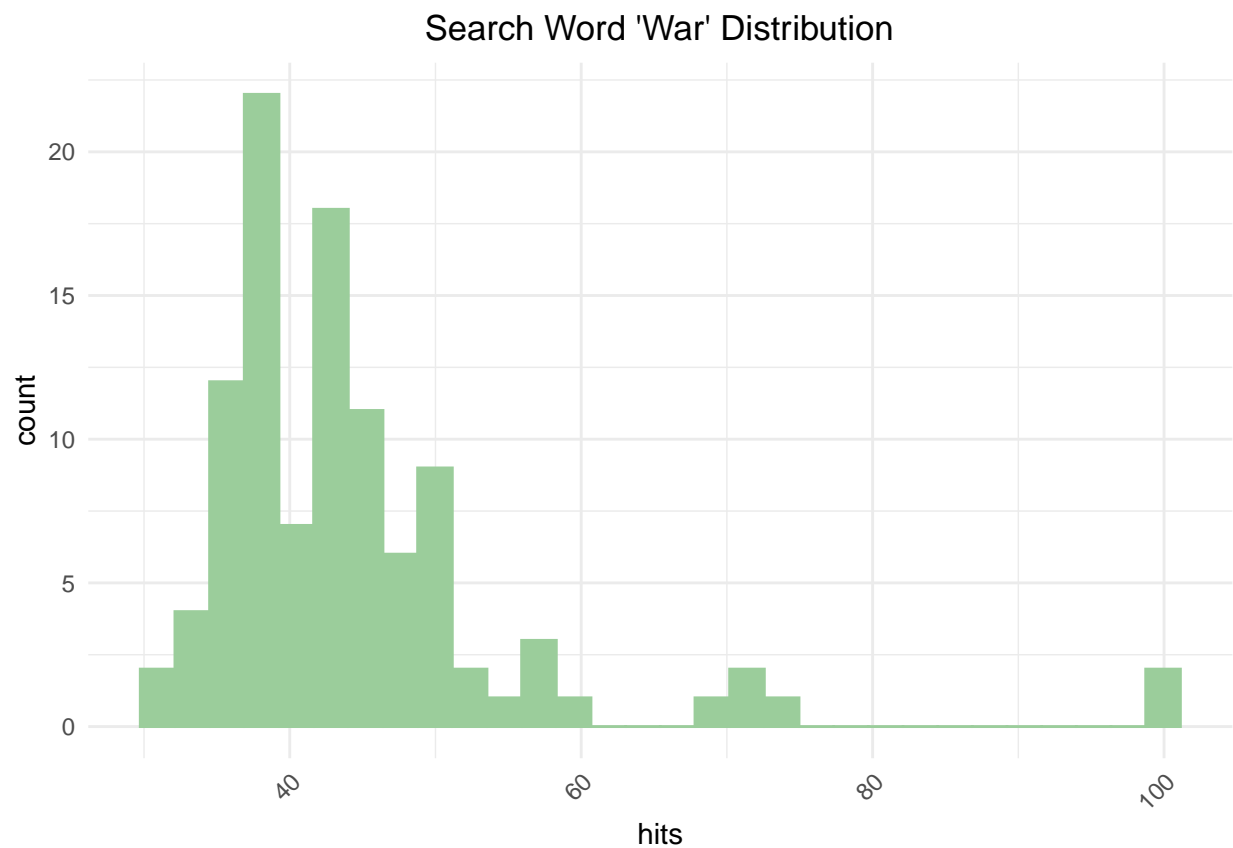
ggplot(gtrends_war_web, aes(date, hits)) +
  geom_line(color='darkseagreen3') +
  theme_minimal() +
  ggtitle("Google Trends - Evolving Interest in Search Word 'War'") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        plot.title = element_text(hjust = 0.5)) +
  geom_vline(xintercept = as.POSIXct("2022-02-24"), linetype="dashed",
            color = "black", size=0.5) +
  annotate('label', x=as.POSIXct("2022-02-24"), y=80,
          label="Start Russian Invasion of Ukraine", vjust=2, color="black") +
  geom_vline(xintercept = as.POSIXct("2023-10-07"), linetype="dashed",
            color = "black", size=0.5) +
  annotate('label', x=as.POSIXct("2023-10-07"), y=80,
          label="Start Israel-Hamas War", vjust=2, color="black")
```





```
# Histogram
ggplot(gtrends_war_web, aes(hits)) +
  geom_histogram(color='darkseagreen3', fill='darkseagreen3') +
  ggtitle("Search Word 'War' Distribution") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        plot.title = element_text(hjust = 0.5))
```

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



## Stationarity test and adjustment

```
# Check stationarity
sapply(ticker_data, adf.test)
```

### Ticker Data

```
##           XLK.Adjusted           XLF.Adjusted
## statistic    -1.952864          -2.839779
## parameter     4                4
## alternative "stationary"        "stationary"
## p.value      0.5962105          0.228463
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"          "X[[i]]"
##           XLV.Adjusted           XLY.Adjusted
## statistic    -3.792237          -2.573107
## parameter     4                4
## alternative "stationary"        "stationary"
## p.value      0.02194783          0.339035
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"          "X[[i]]"
##           XLE.Adjusted           XLU.Adjusted
## statistic    -3.42728           -3.526364
## parameter     4                4
## alternative "stationary"        "stationary"
## p.value      0.05372003          0.04313627
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"          "X[[i]]"
```

Non-stationary are XLK, XLF, XLY, XLE. Make them stationary. Only log the others.

```
ticker_data_differenced <- NULL
ticker_data_differenced$XLK.Adjusted <- na.omit(diff(log(ticker_data$XLK.Adjusted)))
ticker_data_differenced$XLF.Adjusted <- na.omit(diff(log(ticker_data$XLF.Adjusted)))
ticker_data_differenced$XLY.Adjusted <- na.omit(diff(log(ticker_data$XLY.Adjusted)))
ticker_data_differenced$XLE.Adjusted <- na.omit(diff(log(ticker_data$XLE.Adjusted)))
```

```
# Already stationary ticker series
```

```
ticker_data_differenced$XLV.Adjusted <- log(ticker_data$XLV.Adjusted)
ticker_data_differenced$XLU.Adjusted <- log(ticker_data$XLU.Adjusted)
```

```
# For non differenced time series remove first row
```

```
ticker_data_differenced$XLV.Adjusted <- ticker_data_differenced$XLV.Adjusted[-1,]
ticker_data_differenced$XLU.Adjusted <- ticker_data_differenced$XLU.Adjusted[-1,]
```

```
# Check stationarity again after diff-log transformation
```

```
sapply(ticker_data_differenced, adf.test)
```

```
## Warning in FUN(X[[i]], ...): p-value smaller than printed p-value
```

```
## Warning in FUN(X[[i]], ...): p-value smaller than printed p-value

## Warning in FUN(X[[i]], ...): p-value smaller than printed p-value

## Warning in FUN(X[[i]], ...): p-value smaller than printed p-value

##           XLK.Adjusted           XLF.Adjusted
## statistic  -4.849782           -5.398199
## parameter   4                  4
## alternative "stationary"       "stationary"
## p.value     0.01              0.01
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"         "X[[i]]"
##           XLY.Adjusted           XLE.Adjusted
## statistic  -4.882529           -5.159398
## parameter   4                  4
## alternative "stationary"       "stationary"
## p.value     0.01              0.01
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"         "X[[i]]"
##           XLV.Adjusted           XLU.Adjusted
## statistic  -3.793158           -3.567445
## parameter   4                  4
## alternative "stationary"       "stationary"
## p.value     0.02191681         0.03947997
## method      "Augmented Dickey-Fuller Test" "Augmented Dickey-Fuller Test"
## data.name    "X[[i]]"         "X[[i]]"
```

After the diff-log-transformation all the time series are stationary.

```
adf.test(gtrends_war_web$hits)
```

## Google Trends Data

```
##
## Augmented Dickey-Fuller Test
##
## data:  gtrends_war_web$hits
## Dickey-Fuller = -3.5584, Lag order = 4, p-value = 0.04027
## alternative hypothesis: stationary
```

Not stationary and distribution has long tail on right -> Log Diff. Although p value is slightly below 0.05.

```
logged_diff_hits <- na.omit(diff(log(gtrends_war_web$hits)))

corresponding_dates <- gtrends_war_web[-1,]$date # +1 due to diff reduction

gtrends_war_web_differenced <- data.frame(
  date = corresponding_dates,
  hits = logged_diff_hits
```

```
)
# Check stationarity after diff-log-transformation
adf.test(gtrends_war_web_differenced$hits)

## Warning in adf.test(gtrends_war_web_differenced$hits): p-value smaller than
## printed p-value

##
## Augmented Dickey-Fuller Test
##
## data: gtrends_war_web_differenced$hits
## Dickey-Fuller = -5.2254, Lag order = 4, p-value = 0.01
## alternative hypothesis: stationary
```

Check that differenced time series have same length.

```
# Check length of time series
length(gtrends_war_web_differenced$date) == length(ticker_data_differenced$XLK.Adjusted)

## [1] TRUE

length(gtrends_war_web_differenced$date)

## [1] 103

length(ticker_data_differenced$XLK.Adjusted)

## [1] 103

min(gtrends_war_web_differenced$date)

## [1] "2021-12-13 01:00:00 CET"

max(gtrends_war_web_differenced$date)

## [1] "2023-11-27 01:00:00 CET"
```

Create final dataframe

```
# Make tickers in list to dataframe
ticker_data_differenced_df <- do.call(cbind.data.frame, ticker_data_differenced)

# Make date index to column for merging
ticker_data_differenced_df$date <- rownames(ticker_data_differenced_df)

# Make date in gtrends to character in order to merge
gtrends_war_web_differenced$date <- format(gtrends_war_web_differenced$date, format = "%Y-%m-%d")

# Merge
war_ticker_df <- left_join(gtrends_war_web_differenced, ticker_data_differenced_df)

## Joining, by = "date"
```

## 4. VAR & Causality Testing

TODO: - Check lag.max what to apply - Control variable?

```
# Create ticker list
ticker_cols <- colnames(war_ticker_df)[3:8]

for (Ticker in ticker_cols) {

  # Create data for VAR
  print(Ticker)
  data_for_var <- cbind(war_interest=war_ticker_df$hits, ticker=war_ticker_df[[Ticker]])

  # Run VAR model
  VAR_est <- VAR(data_for_var, ic = "AIC", lag.max = 24)
  coefs <- coefest(VAR_est)
  summ <- summary(VAR_est)

  print(coefs)
  print(summ)

  # Run Granger Causality Test
  causal <- causality(VAR_est, cause="war_interest")["Granger"]
  print(causal)

  # Run Granger Causality Test other way (check if index has influence on crisis measure)
  causal_ticker <- causality(VAR_est, cause="ticker")["Granger"]
  print(causal_ticker)

  # Impulse response functions
  plot(irf(VAR_est, impulse="war_interest", response="ticker"))

}
```

```
## [1] "XLK.Adjusted"
##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    0.0018754  0.0136893   0.1370   0.8913
## war_interest:war_interest.l1 0.0590052  0.1003184   0.5882   0.5578
## war_interest:ticker.l1     -0.2429632  0.3888822  -0.6248   0.5336
## ticker:(Intercept)         0.0012100  0.0034886   0.3469   0.7294
## ticker:war_interest.l1     -0.0279391  0.0255650  -1.0929   0.2771
## ticker:ticker.l1          -0.0479635  0.0991024  -0.4840   0.6295
##
##
## VAR Estimation Results:
## =====
## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 256.754
## Roots of the characteristic polynomial:
```

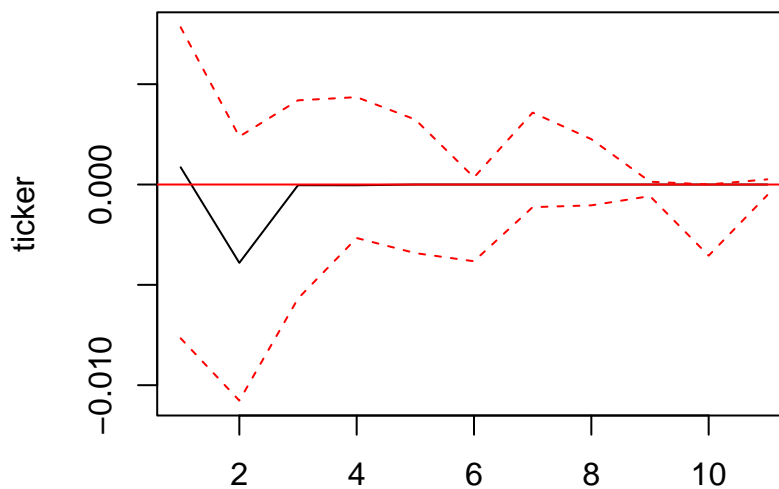
```

## 0.1037 0.09271
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.059005   0.100318   0.588   0.558
## ticker.l1        -0.242963   0.388882  -0.625   0.534
## const            0.001875   0.013689   0.137   0.891
##
##
## Residual standard error: 0.1382 on 99 degrees of freedom
## Multiple R-Squared: 0.007273, Adjusted R-squared: -0.01278
## F-statistic: 0.3627 on 2 and 99 DF, p-value: 0.6967
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 -0.027939   0.025565  -1.093   0.277
## ticker.l1        -0.047964   0.099102  -0.484   0.629
## const            0.001210   0.003489   0.347   0.729
##
##
## Residual standard error: 0.03522 on 99 degrees of freedom
## Multiple R-Squared: 0.01439, Adjusted R-squared: -0.005526
## F-statistic: 0.7225 on 2 and 99 DF, p-value: 0.4881
##
##
##
## Covariance matrix of residuals:
##           war_interest  ticker
## war_interest  0.0191051 0.0001187
## ticker        0.0001187 0.0012407
##
## Correlation matrix of residuals:
##           war_interest  ticker
## war_interest  1.00000 0.02438
## ticker        0.02438 1.00000
##
##
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data: VAR object VAR_est
## F-Test = 1.1944, df1 = 1, df2 = 198, p-value = 0.2758
##

```

```
##
## $Granger
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data: VAR object VAR_est
## F-Test = 0.39034, df1 = 1, df2 = 198, p-value = 0.5328
```

Orthogonal Impulse Response from war\_interest



95 % Bootstrap CI, 100 runs

```
## [1] "XLF.Adjusted"
##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    0.00170637  0.01371596  0.1244  0.9012
## war_interest:war_interest.l1  0.05805078  0.10051706  0.5775  0.5649
## war_interest:ticker.l1    -0.00137969  0.44072266 -0.0031  0.9975
## ticker:(Intercept)    -0.00022179  0.00306846 -0.0723  0.9425
## ticker:war_interest.l1    -0.04134063  0.02248710 -1.8384  0.0690
## ticker:ticker.l1         0.07442982  0.09859594  0.7549  0.4521
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## VAR Estimation Results:
## =====
```

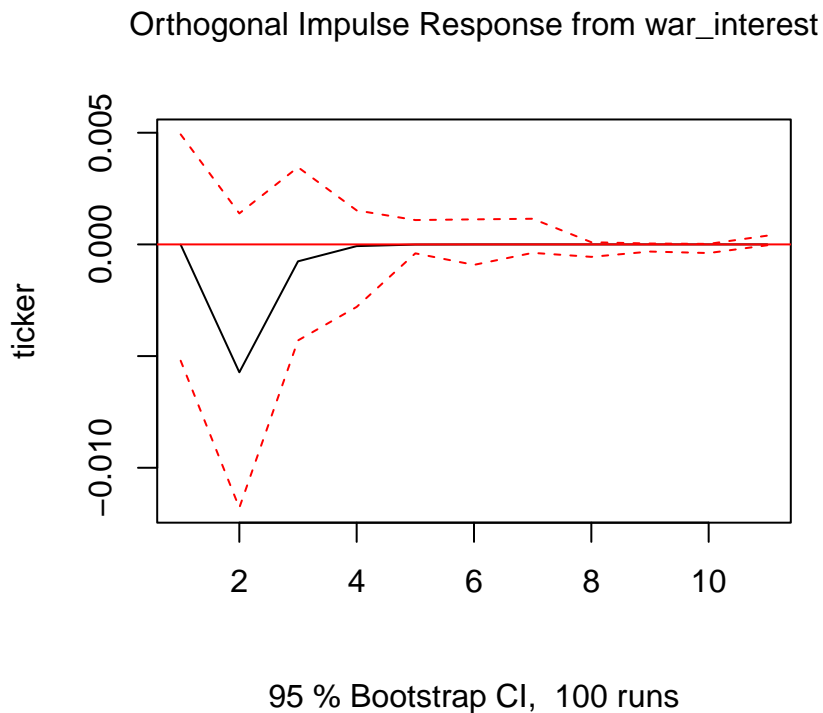
```

## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 269.609
## Roots of the characteristic polynomial:
## 0.07738 0.0551
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.058051   0.100517   0.578   0.565
## ticker.l1        -0.001380   0.440723  -0.003   0.998
## const            0.001706   0.013716   0.124   0.901
##
##
## Residual standard error: 0.1385 on 99 degrees of freedom
## Multiple R-Squared: 0.003359, Adjusted R-squared: -0.01677
## F-statistic: 0.1668 on 2 and 99 DF, p-value: 0.8466
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 -0.0413406  0.0224871  -1.838   0.069 .
## ticker.l1        0.0744298  0.0985959   0.755   0.452
## const           -0.0002218  0.0030685  -0.072   0.943
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## Residual standard error: 0.03098 on 99 degrees of freedom
## Multiple R-Squared: 0.03879, Adjusted R-squared: 0.01937
## F-statistic: 1.997 on 2 and 99 DF, p-value: 0.1411
##
##
##
## Covariance matrix of residuals:
##           war_interest      ticker
## war_interest  1.918e-02 -1.409e-06
## ticker        -1.409e-06  9.599e-04
##
## Correlation matrix of residuals:
##           war_interest      ticker
## war_interest  1.0000000 -0.0003284
## ticker        -0.0003284  1.0000000
##
##

```



```
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data: VAR object VAR_est
## F-Test = 3.3798, df1 = 1, df2 = 198, p-value = 0.0675
##
##
## $Granger
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data: VAR object VAR_est
## F-Test = 9.8001e-06, df1 = 1, df2 = 198, p-value = 0.9975
```



```
## [1] "XLY.Adjusted"
##
## t test of coefficients:
##
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    0.0017293  0.0137281   0.1260  0.90001
## war_interest:war_interest.l1  0.0583668  0.1008952   0.5785  0.56425
## war_interest:ticker.l1       0.0127693  0.3645435   0.0350  0.97213
## ticker:(Intercept)        -0.0013305  0.0036677  -0.3628  0.71756
## ticker:war_interest.l1     -0.0613923  0.0269556  -2.2775  0.02491 *
```

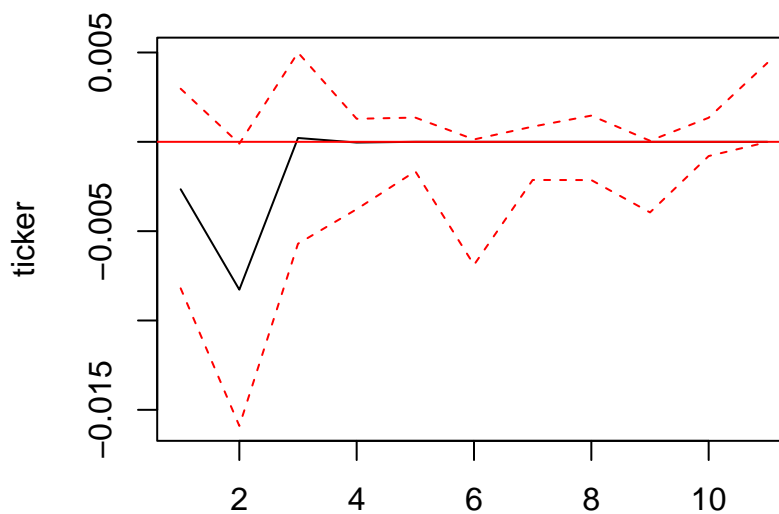
```

## ticker:ticker.l1          -0.0857372  0.0973932 -0.8803  0.38082
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## VAR Estimation Results:
## =====
## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 251.769
## Roots of the characteristic polynomial:
## 0.08007 0.0527
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##              Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 0.058367  0.100895  0.578  0.564
## ticker.l1       0.012769  0.364544  0.035  0.972
## const          0.001729  0.013728  0.126  0.900
##
##
## Residual standard error: 0.1385 on 99 degrees of freedom
## Multiple R-Squared: 0.003371, Adjusted R-squared: -0.01676
## F-statistic: 0.1675 on 2 and 99 DF, p-value: 0.8461
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##              Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 -0.061392  0.026956 -2.278  0.0249 *
## ticker.l1       -0.085737  0.097393 -0.880  0.3808
## const          -0.001330  0.003668 -0.363  0.7176
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## Residual standard error: 0.037 on 99 degrees of freedom
## Multiple R-Squared: 0.05402, Adjusted R-squared: 0.03491
## F-statistic: 2.827 on 2 and 99 DF, p-value: 0.06401
##
##
## Covariance matrix of residuals:
##              war_interest      ticker
## war_interest  0.0191802 -0.0003676
## ticker       -0.0003676  0.0013690

```

```
##
## Correlation matrix of residuals:
##           war_interest  ticker
## war_interest      1.00000 -0.07174
## ticker            -0.07174  1.00000
##
##
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data:  VAR object VAR_est
## F-Test = 5.1871, df1 = 1, df2 = 198, p-value = 0.02382
##
##
## $Granger
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data:  VAR object VAR_est
## F-Test = 0.001227, df1 = 1, df2 = 198, p-value = 0.9721
```

Orthogonal Impulse Response from war\_interest



95 % Bootstrap CI, 100 runs

```
## [1] "XLE.Adjusted"
##
## t test of coefficients:
```

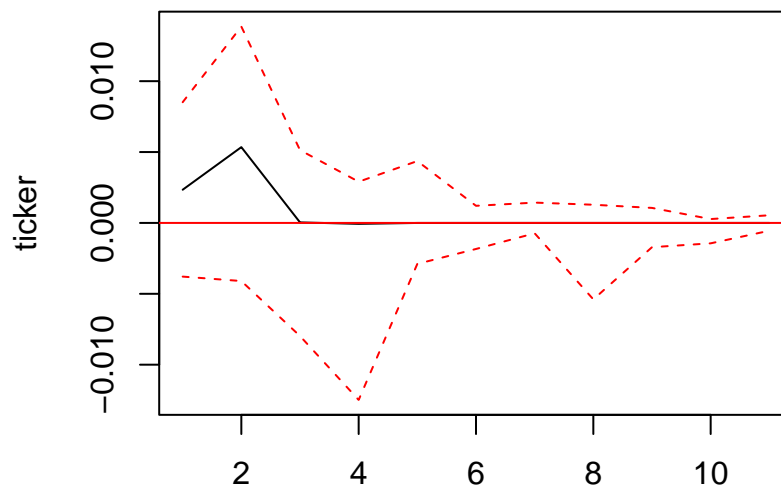
```

##
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    0.0035701  0.0136526  0.2615  0.7943
## war_interest:war_interest.l1 0.0661948  0.0997379  0.6637  0.5084
## war_interest:ticker.l1      -0.4018675  0.2927685 -1.3726  0.1730
## ticker:(Intercept)          0.0053230  0.0046244  1.1511  0.2525
## ticker:war_interest.l1      0.0398860  0.0337831  1.1806  0.2406
## ticker:ticker.l1            -0.0537284  0.0991663 -0.5418  0.5892
##
##
## VAR Estimation Results:
## =====
## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 229.352
## Roots of the characteristic polynomial:
## 0.1117 0.1117
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.06620    0.09974  0.664  0.508
## ticker.l1        -0.40187    0.29277 -1.373  0.173
## const            0.00357    0.01365  0.261  0.794
##
##
## Residual standard error: 0.1372 on 99 degrees of freedom
## Multiple R-Squared: 0.02197, Adjusted R-squared: 0.002215
## F-statistic: 1.112 on 2 and 99 DF,  p-value: 0.3329
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.039886  0.033783  1.181  0.241
## ticker.l1        -0.053728  0.099166 -0.542  0.589
## const            0.005323  0.004624  1.151  0.252
##
##
## Residual standard error: 0.04647 on 99 degrees of freedom
## Multiple R-Squared: 0.01607, Adjusted R-squared: -0.003805
## F-statistic: 0.8086 on 2 and 99 DF,  p-value: 0.4484
##
##
## Covariance matrix of residuals:

```

```
##           war_interest  ticker
## war_interest    0.018822 0.000321
## ticker          0.000321 0.002159
##
## Correlation matrix of residuals:
##           war_interest  ticker
## war_interest    1.00000 0.05034
## ticker          0.05034 1.00000
##
##
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data:  VAR object VAR_est
## F-Test = 1.3939, df1 = 1, df2 = 198, p-value = 0.2392
##
##
## $Granger
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data:  VAR object VAR_est
## F-Test = 1.8842, df1 = 1, df2 = 198, p-value = 0.1714
```

### Orthogonal Impulse Response from war\_interest



95 % Bootstrap CI, 100 runs

```
## [1] "XLV.Adjusted"
```

```

##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    1.8501368  2.0648758  0.8960 0.3724242
## war_interest:war_interest.l1  0.0515319  0.1003650  0.5134 0.6087841
## war_interest:ticker.l1      -0.3800757  0.4245723 -0.8952 0.3728533
## ticker:(Intercept)          1.2078766  0.3191858  3.7842 0.0002642 ***
## ticker:war_interest.l1      -0.0055966  0.0155143 -0.3607 0.7190633
## ticker:ticker.l1            0.7515999  0.0656298 11.4521 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## VAR Estimation Results:
## =====
## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 308.138
## Roots of the characteristic polynomial:
## 0.7546 0.04851
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.05153  0.10037  0.513  0.609
## ticker.l1        -0.38008  0.42457 -0.895  0.373
## const            1.85014  2.06488  0.896  0.372
##
##
## Residual standard error: 0.1379 on 99 degrees of freedom
## Multiple R-Squared: 0.01136, Adjusted R-squared: -0.008611
## F-statistic: 0.5689 on 2 and 99 DF, p-value: 0.568
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 -0.005597  0.015514 -0.361 0.719063
## ticker.l1        0.751600  0.065630 11.452 < 2e-16 ***
## const            1.207877  0.319186  3.784 0.000264 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## Residual standard error: 0.02132 on 99 degrees of freedom

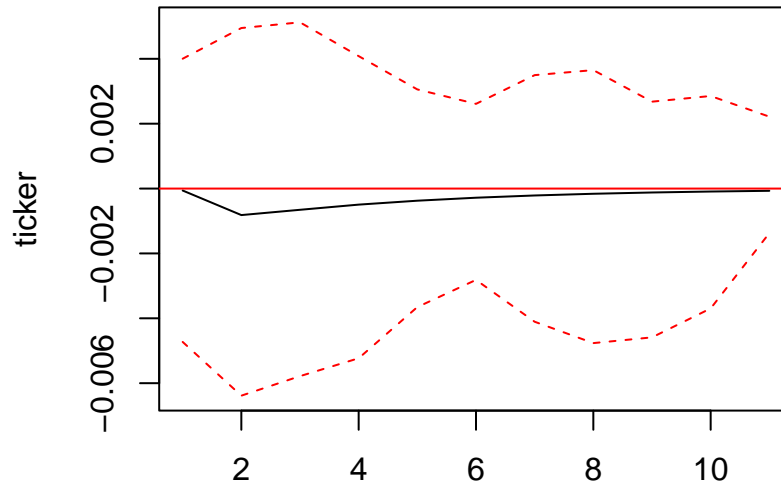
```

```

## Multiple R-Squared: 0.5725, Adjusted R-squared: 0.5639
## F-statistic: 66.29 on 2 and 99 DF, p-value: < 2.2e-16
##
##
##
## Covariance matrix of residuals:
##           war_interest    ticker
## war_interest    1.903e-02 -8.358e-06
## ticker          -8.358e-06  4.546e-04
##
## Correlation matrix of residuals:
##           war_interest    ticker
## war_interest    1.000000 -0.002842
## ticker          -0.002842  1.000000
##
##
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data: VAR object VAR_est
## F-Test = 0.13013, df1 = 1, df2 = 198, p-value = 0.7187
##
##
## $Granger
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data: VAR object VAR_est
## F-Test = 0.80138, df1 = 1, df2 = 198, p-value = 0.3718

```

# Orthogonal Impulse Response from war\_interest



95 % Bootstrap CI, 100 runs

```
## [1] "XLU.Adjusted"
##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## war_interest:(Intercept)    1.29705075  1.04473333  1.2415  0.21735
## war_interest:war_interest.l1  0.04383159  0.10039046  0.4366  0.66334
## war_interest:ticker.l1    -0.30890066  0.24911648 -1.2400  0.21791
## ticker:(Intercept)         0.55600538  0.21504017  2.5856  0.01118 *
## ticker:war_interest.l1    -0.00039069  0.02066363 -0.0189  0.98495
## ticker:ticker.l1          0.86729400  0.05127629 16.9141 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## VAR Estimation Results:
## =====
## Endogenous variables: war_interest, ticker
## Deterministic variables: const
## Sample size: 102
## Log Likelihood: 279.692
## Roots of the characteristic polynomial:
## 0.8674 0.04369
## Call:
## VAR(y = data_for_var, lag.max = 24, ic = "AIC")
##
```

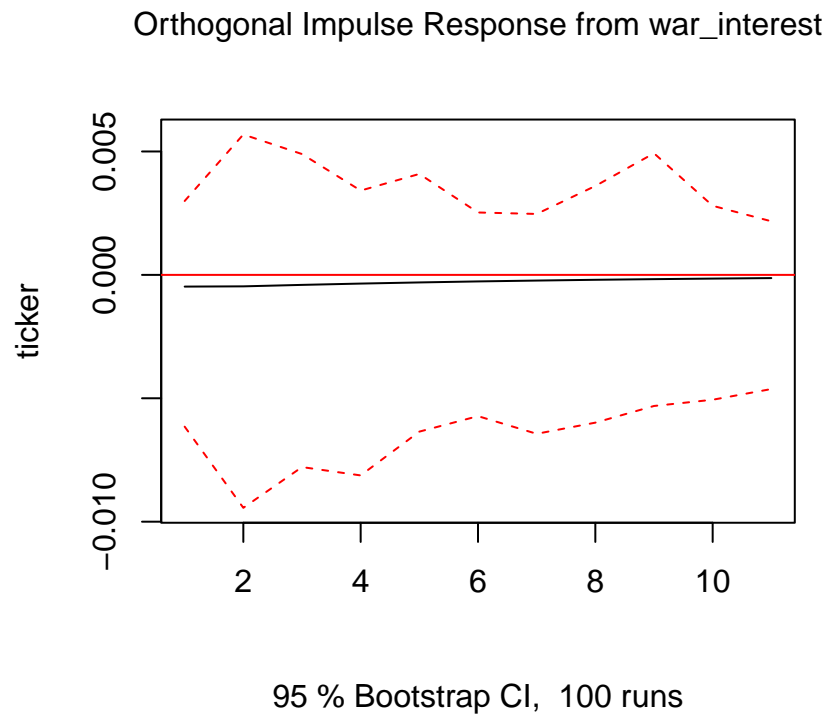


```

##
## Estimation results for equation war_interest:
## =====
## war_interest = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1  0.04383    0.10039   0.437   0.663
## ticker.l1        -0.30890    0.24912  -1.240   0.218
## const           1.29705    1.04473   1.242   0.217
##
##
## Residual standard error: 0.1374 on 99 degrees of freedom
## Multiple R-Squared: 0.0186, Adjusted R-squared: -0.001225
## F-statistic: 0.9382 on 2 and 99 DF, p-value: 0.3948
##
##
## Estimation results for equation ticker:
## =====
## ticker = war_interest.l1 + ticker.l1 + const
##
##           Estimate Std. Error t value Pr(>|t|)
## war_interest.l1 -0.0003907  0.0206636  -0.019   0.9850
## ticker.l1        0.8672940  0.0512763  16.914 <2e-16 ***
## const           0.5560054  0.2150402   2.586  0.0112 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## Residual standard error: 0.02829 on 99 degrees of freedom
## Multiple R-Squared: 0.7455, Adjusted R-squared: 0.7403
## F-statistic: 145 on 2 and 99 DF, p-value: < 2.2e-16
##
##
## Covariance matrix of residuals:
##           war_interest    ticker
## war_interest  1.889e-02 -6.518e-05
## ticker       -6.518e-05  8.002e-04
##
## Correlation matrix of residuals:
##           war_interest    ticker
## war_interest  1.00000 -0.01677
## ticker       -0.01677  1.00000
##
##
## $Granger
##
## Granger causality H0: war_interest do not Granger-cause ticker
##
## data:  VAR object VAR_est
## F-Test = 0.00035747, df1 = 1, df2 = 198, p-value = 0.9849
##
##
## $Granger

```

```
##
## Granger causality H0: ticker do not Granger-cause war_interest
##
## data: VAR object VAR_est
## F-Test = 1.5376, df1 = 1, df2 = 198, p-value = 0.2164
```



## 5. Results & Discussion