

stocks15_analysis.R

pw

2022-01-24

```
# This code is an analysis of the data at 15 minute intervals.
# This code includes the following steps:
## STEP 0: LOAD LIBRARIES AND DATA
## STEP 1: PREPROCESSING
### 1.1: Difference and log-transform stock prices
### 1.2: Create morality index
### 1.3: Add lags
### 1.4: Reorder and make time series
### 1.5: Check serial dependencies; ACF & PACF
## STEP 2: DATA EXPLORATION
### 2.1: Histograms
### 2.2: Box & Violine plots
### 2.3: Scatterplots
## STEP 3: FITTING AND EVALUATING MODEL
### 3.1: Removing outliers
### 3.2: Fitting and summarizing models
### 3.3: Model comparisons
### 3.4: Standardized betas
### 3.6: Interaction plot for morality model
### 3.7: Interaction plots for foundation models

knitr::opts_chunk$set(echo = TRUE,
                      collapse = FALSE,
                      warning = FALSE,
                      tidy = TRUE)

options(width=120)

#####
# STEP 0: LOAD DATA

## Read data
setwd("/home/pw/Projects/mfstocks/data/csv/60shift")
stocks15 <- read.table("data15_60shift.csv", header=TRUE, stringsAsFactors=TRUE, sep=",", na.strings="N")
## Convert datetime variable to POSIX, which will allow us to create a time series later
stocks15$dt_15min <- as.POSIXct(stocks15$dt_15min, tz="EST")
## Factors are used to represent categorical data in statistical analysis. They are stored as unique in
col15 <- c("season_workday_15min", "season_month_15min",
          "tf1_15min", "tf2_15min", "tf3_15min", "tf4_15min", "tf5_15min", "tf6_15min")
stocks15[col15] <- lapply(stocks15[col15], as.factor)

#####
```

```
# STEP 1: PREPROCESSING
```

```
library(RcmdrMisc)
```

```
## Loading required package: car
```

```
## Loading required package: carData
```

```
## Loading required package: sandwich
```

```
library(lubridate) # date()
```

```
##
```

```
## Attaching package: 'lubridate'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      date, intersect, setdiff, union
```

```
library(dplyr) # %>% function
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:car':
```

```
##
```

```
##      recode
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
library(reshape2) # melt()
```

```
## 1.1: Difference and log-transform stock prices
```

```
### Difference for moving averages
```

```
stocks15$stocks_15min_diff = stocks15$stocks_15min %>% diff() %>% append(NA, 0)
```

```
### Log then difference for variance stabilization
```

```
stocks15$stocks_15min_diff_ln = stocks15$stocks_15min %>% log() %>% diff() %>% append(NA, 0)
```

```
### Create variable for counting the days
```

```
stocks15$day_count =
```

```
  stocks15$dt_15min %>% date() %>% as.integer() -
```

```
  stocks15$dt_15min %>% date() %>% as.integer() %>% min() + 1
```

```
## 1.2: Create morality index from the sentiment-weighted probability of each foundation
```

```
stocks15$morality =  
  stocks15$care_p_15min*stocks15$care_sent_15min +  
  stocks15$fairness_p_15min*stocks15$fairness_sent_15min +  
  stocks15$loyalty_p_15min*stocks15$loyalty_sent_15min +  
  stocks15$authority_p_15min*stocks15$authority_sent_15min +  
  stocks15$sanctity_p_15min*stocks15$sanctity_sent_15min
```

```
## 1.3: Add lags
```

```
### Lag morality index
```

```
stocks15$morality_lag = lag(stocks15$morality)
```

```
### Lag foundation probabilities
```

```
stocks15$care_p_15min_lag = lag(stocks15$care_p_15min)
```

```
stocks15$fairness_p_15min_lag = lag(stocks15$fairness_p_15min)
```

```
stocks15$loyalty_p_15min_lag = lag(stocks15$loyalty_p_15min)
```

```
stocks15$authority_p_15min_lag = lag(stocks15$authority_p_15min)
```

```
stocks15$sanctity_p_15min_lag = lag(stocks15$sanctity_p_15min)
```

```
### Lag foundation sentiments
```

```
stocks15$care_sent_15min_lag = lag(stocks15$care_sent_15min)
```

```
stocks15$fairness_sent_15min_lag = lag(stocks15$fairness_sent_15min)
```

```
stocks15$loyalty_sent_15min_lag = lag(stocks15$loyalty_sent_15min)
```

```
stocks15$authority_sent_15min_lag = lag(stocks15$authority_sent_15min)
```

```
stocks15$sanctity_sent_15min_lag = lag(stocks15$sanctity_sent_15min)
```

```
### Lag sentiment-weighted probabilities of foundations
```

```
stocks15$care_lag = stocks15$care_p_15min_lag * stocks15$care_sent_15min_lag
```

```
stocks15$fairness_lag = stocks15$fairness_p_15min_lag * stocks15$fairness_sent_15min_lag
```

```
stocks15$loyalty_lag = stocks15$loyalty_p_15min_lag * stocks15$loyalty_sent_15min_lag
```

```
stocks15$authority_lag = stocks15$authority_p_15min_lag * stocks15$authority_sent_15min_lag
```

```
stocks15$sanctity_lag = stocks15$sanctity_p_15min_lag * stocks15$sanctity_sent_15min_lag
```

```
## 1.4: Reorder and make time series
```

```
### Reorder columns
```

```
stocks15_ordered = stocks15[, c(1, 24, 2:4, 16:21, 5, 22, 23, 25, 6:15, 27:36, 26, 37:41)]
```

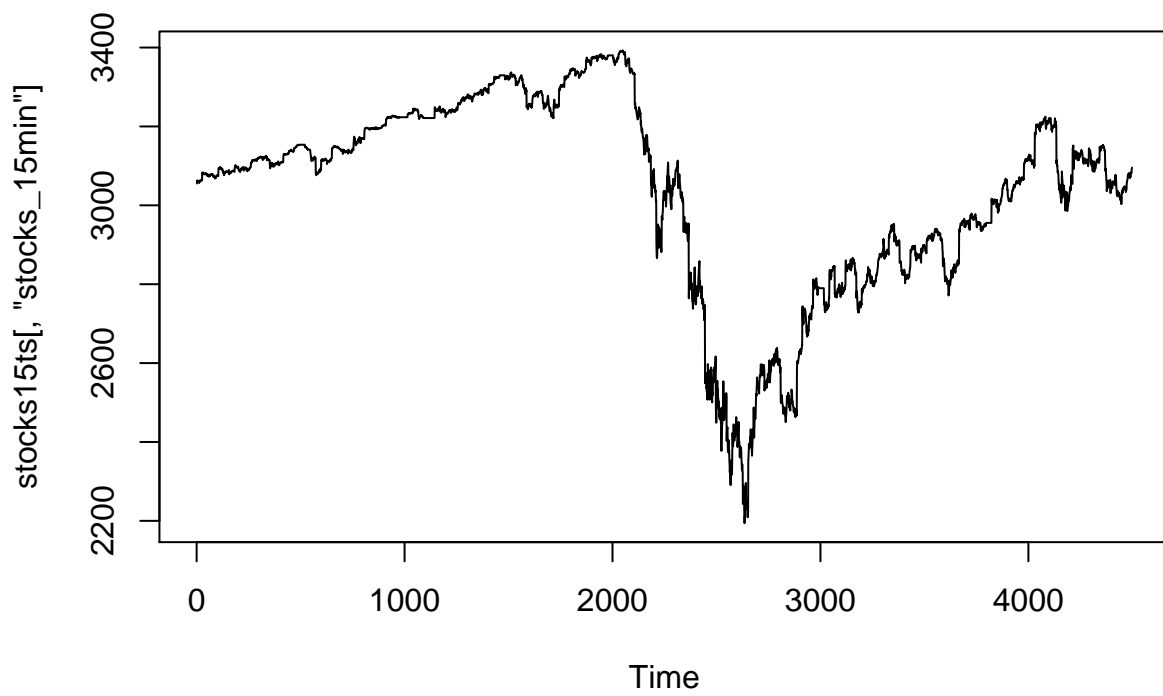
```
### Make time series
```

```
stocks15ts = ts(stocks15_ordered)
```

```
## 1.5: Check serial dependencies; ACF & PACF
```

```
### Plot time series of raw stock prices
```

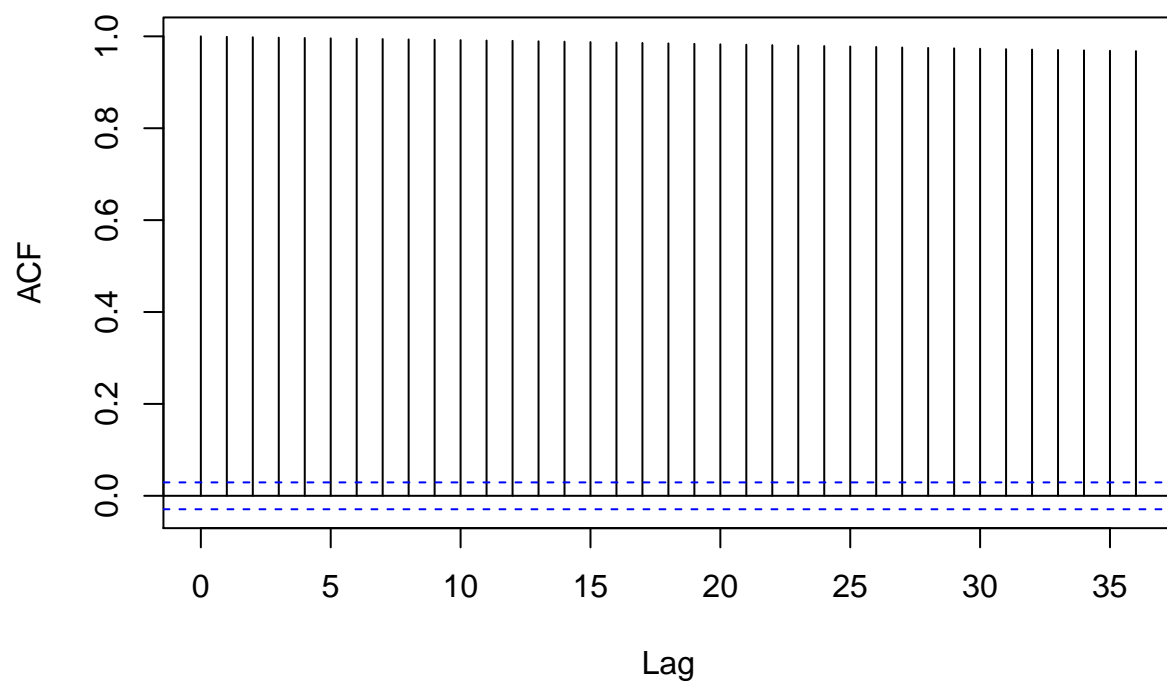
```
plot(stocks15ts[, "stocks_15min"])
```



```
### Plot ACF and PACF for non-transformed data
```

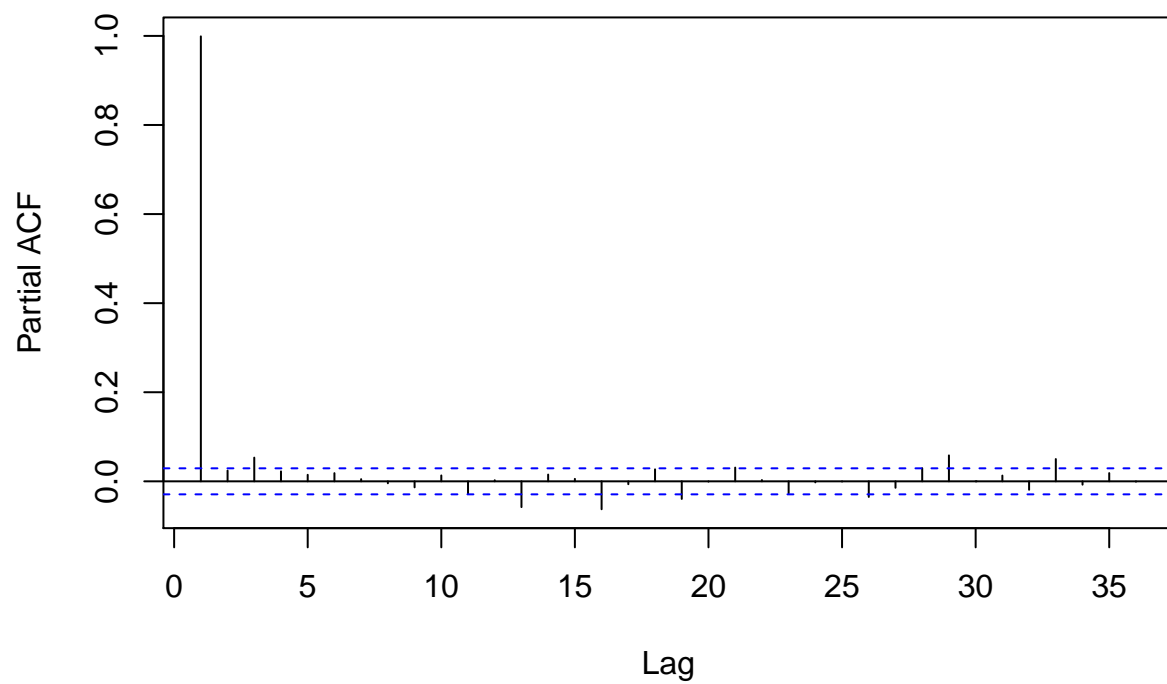
```
acf(stocks15_ordered$stocks_15min, lag.max = NULL, type = c("correlation"), plot = TRUE, na.action = na
```

Series stocks15_ordered\$stocks_15min



```
acf(stocks15_ordered$stocks_15min, lag.max = NULL, type = c("partial"), plot = TRUE, na.action = na.pass)
```

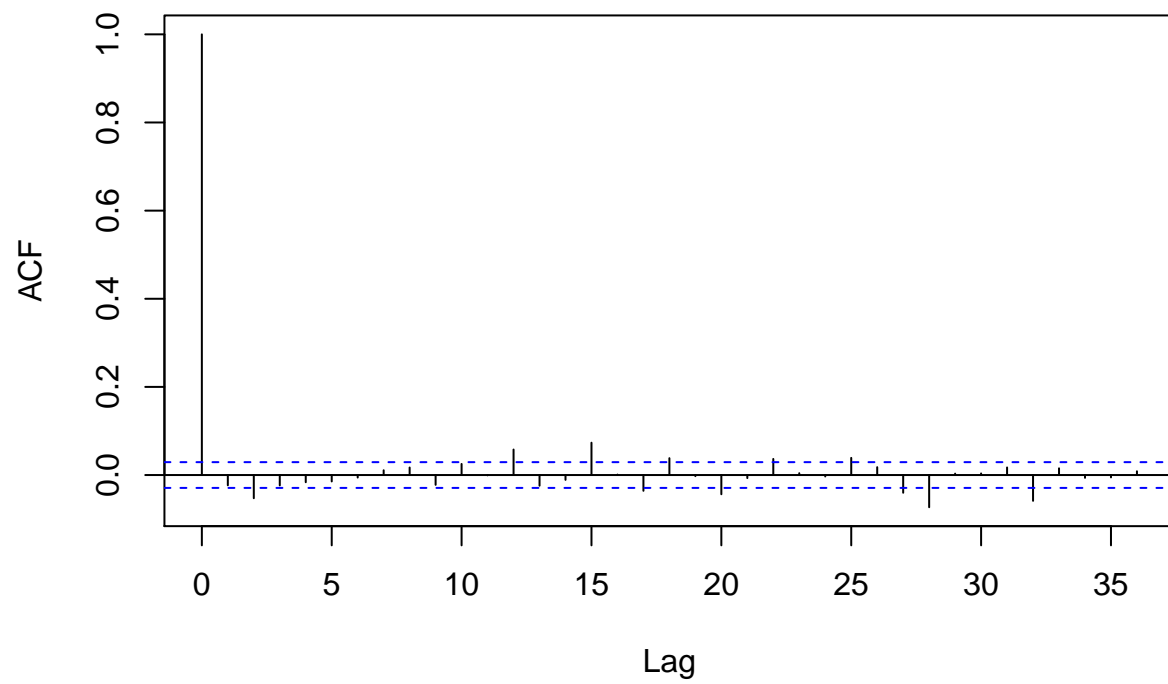
Series stocks15_ordered\$stocks_15min



```
### Plot ACF and PACF for transformed data
```

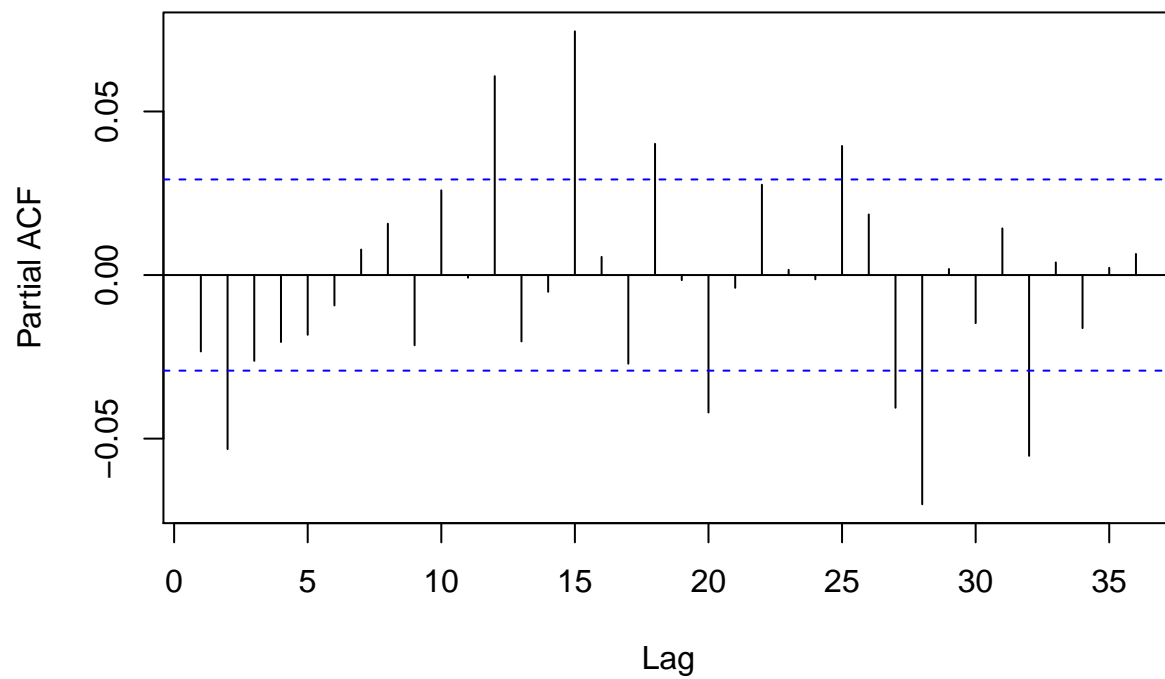
```
acf(stocks15_ordered$stocks_15min_diff_ln, lag.max = NULL, type = c("correlation"), plot = TRUE, na.act
```

Series stocks15_ordered\$stocks_15min_diff_ln

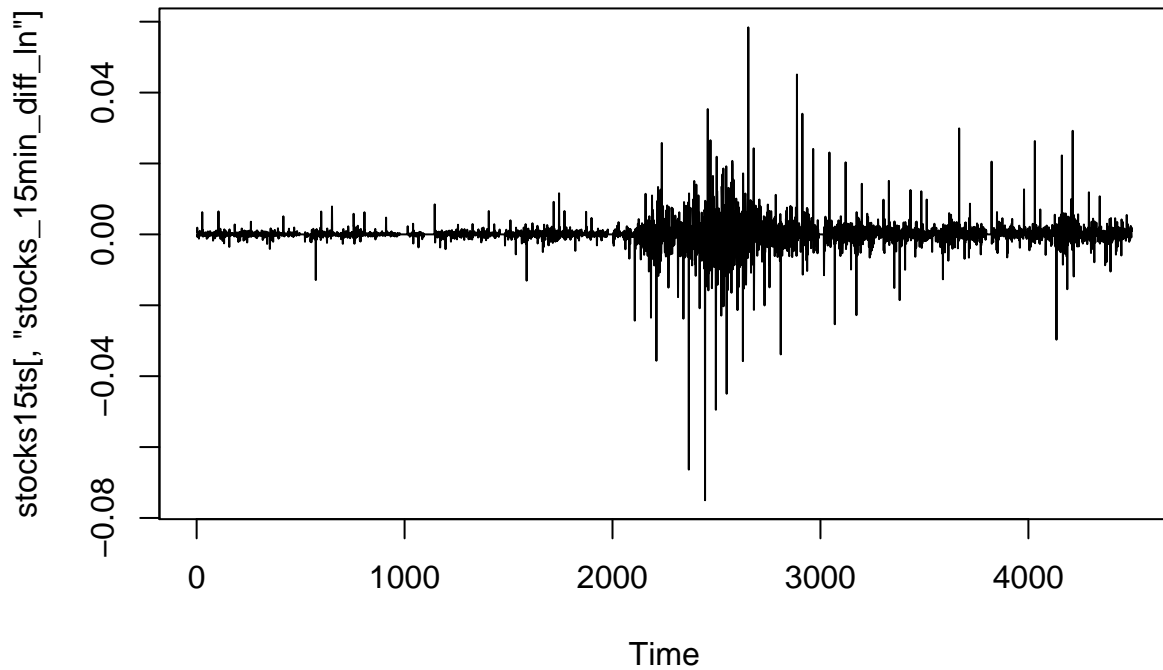


```
acf(stocks15_ordered$stocks_15min_diff_ln, lag.max = NULL, type = c("partial"), plot = TRUE, na.action = na.omit)
```

Series stocks15_ordered\$stocks_15min_diff_ln



```
### Plot time series of stock prices after differencing and log-transformation  
### Mildly unstable variance in contraction and recovery period  
plot(stocks15ts[, "stocks_15min_diff_ln"])
```

```
#####
# STEP 2: DATA EXPLORATION
library(ggplot2)
library(gridExtra) # gridExtra

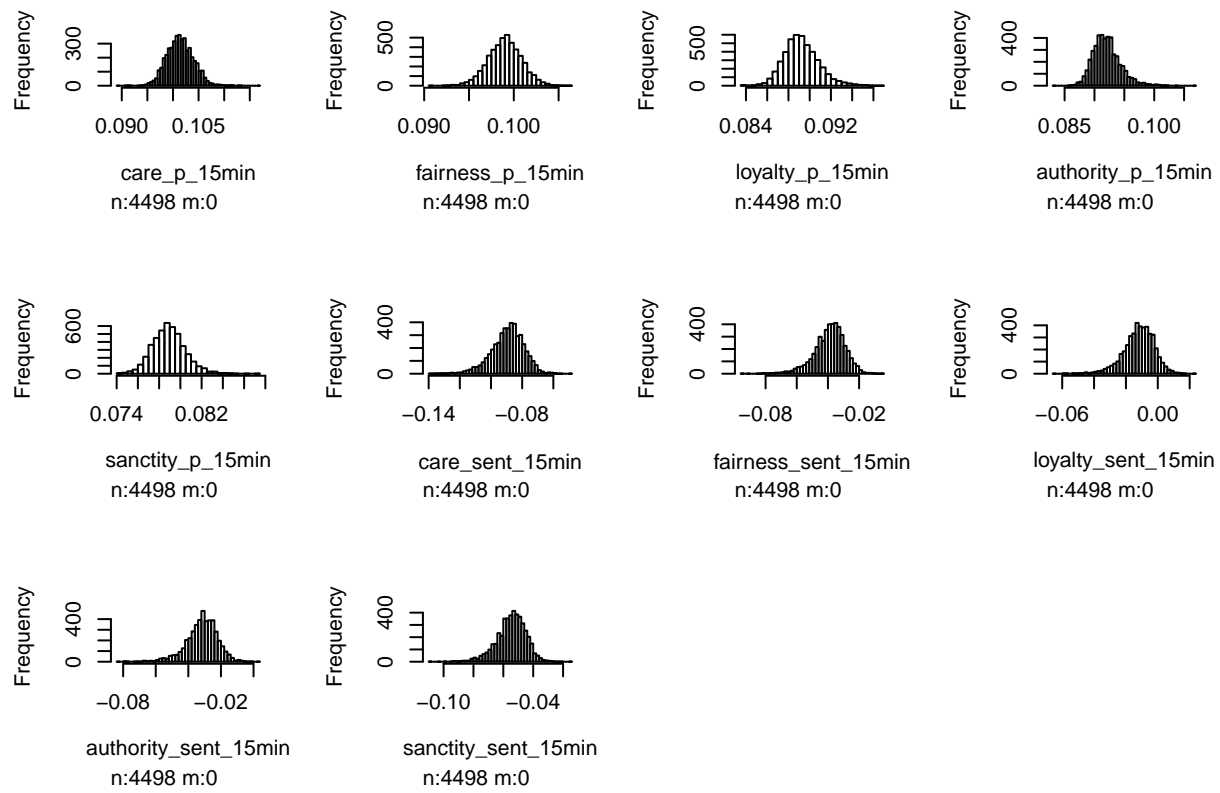
##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':
##
##   combine

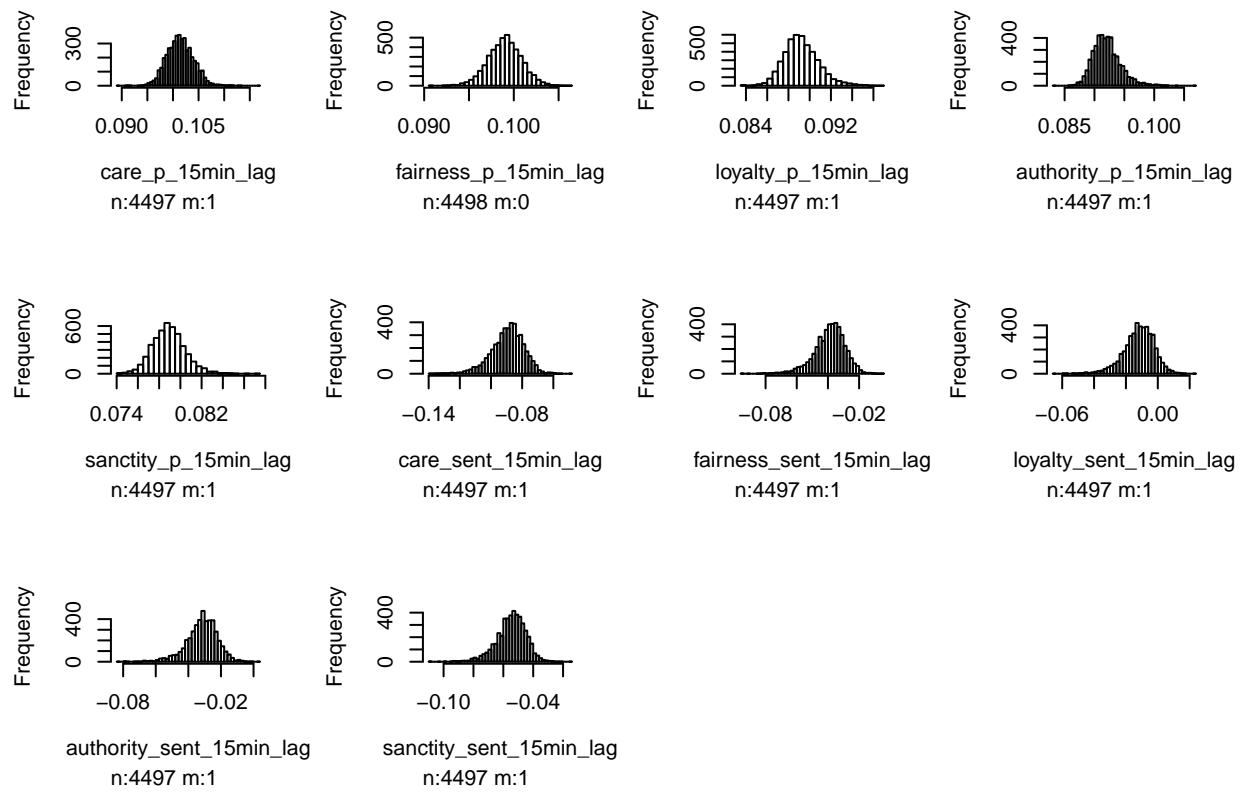
library(ggpubr)
theme_set(theme_pubclean())

## 2.1: Frequency histograms

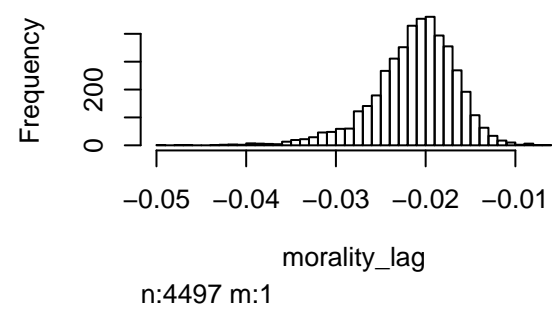
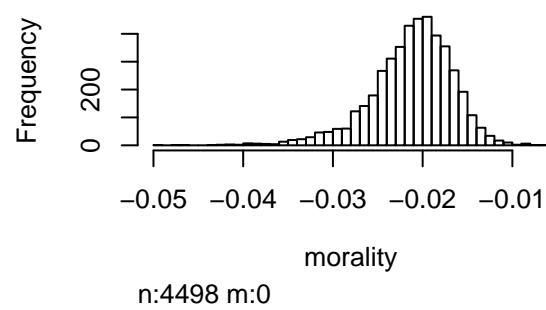
### Probabilities and sentiments of foundations
hist(stocks15_ordered[16:25])
```



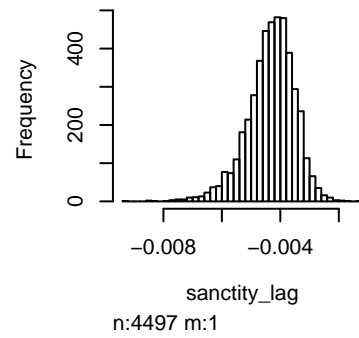
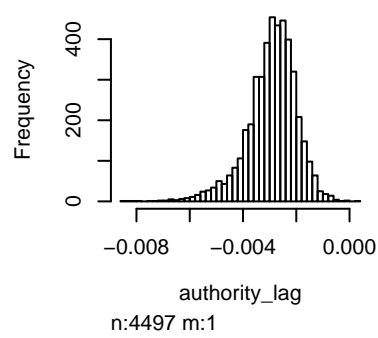
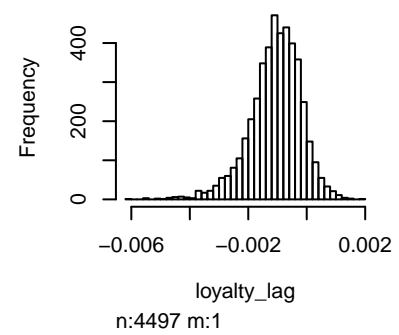
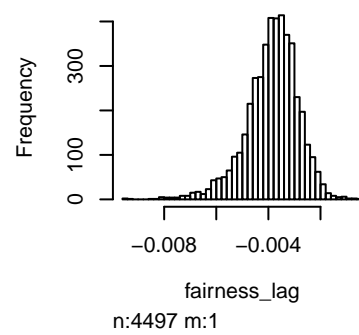
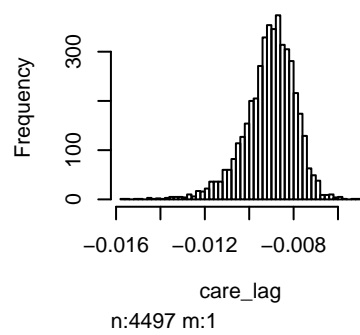
```
### Lagged probabilities and sentiments of foundations
hist(stocks15_ordered[26:35])
```



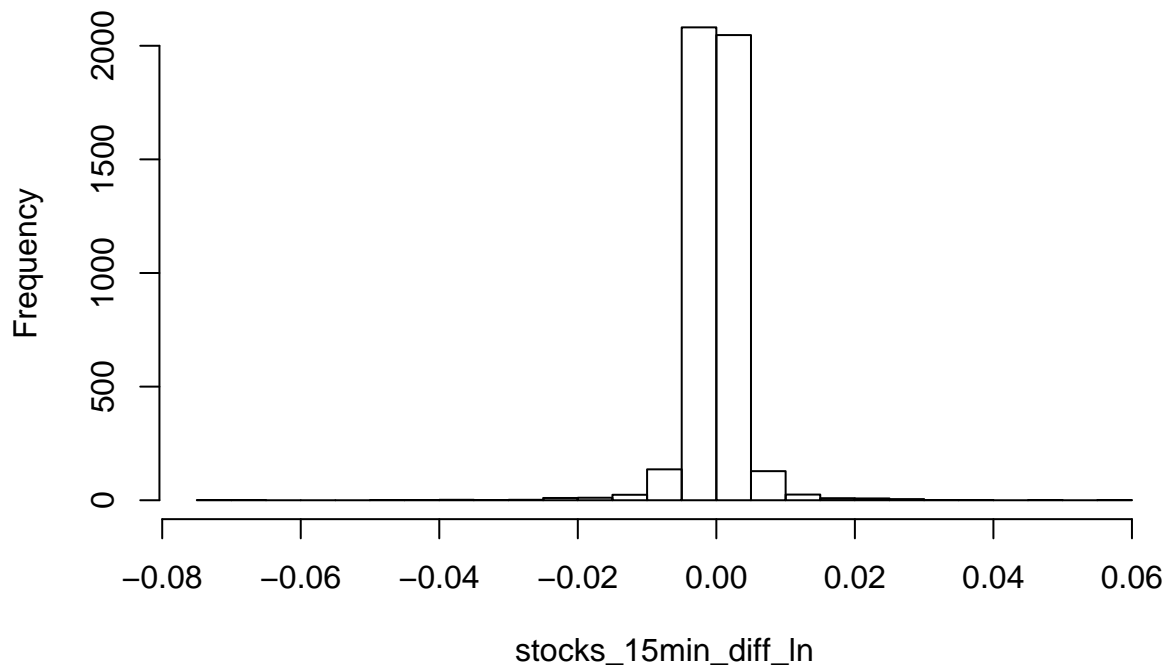
```
### Morality and lagged morality
hist(stocks15_ordered[, c(15,36)])
```



```
### Sentiment-weighted probabilities of foundations lagged  
hist(stocks15_ordered[37:41])
```



```
### Transformed stock prices
hist(stocks15_ordered[14])
```



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2.2: Box & Violine plots

Define plot functions

BOXPLOT

```
plot_boxplot <- function(input, x_labels) {
  return(input +
    geom_boxplot(notch = TRUE, fill = "lightgray") +
    stat_summary(fun.y = mean, geom = "point", shape = 18, size = 2.5, color = "#FC4E07") +
    labs(x="Economic Period", y = "Morality Score") +
    scale_x_discrete(labels=x_labels) )}
```

VIOLIN

```
plot_violin <- function(input, x_labels) {
  return(input +
    geom_violin(trim = FALSE) +
    stat_summary(fun.data = "mean_sdl", fun.args = list(mult = 1), geom = "pointrange", color = "#FC4E07") +
    labs(x="Economic Period", y = "Morality Score") +
    scale_x_discrete(labels=x_labels) )}
```

DOUBLE VIOLIN

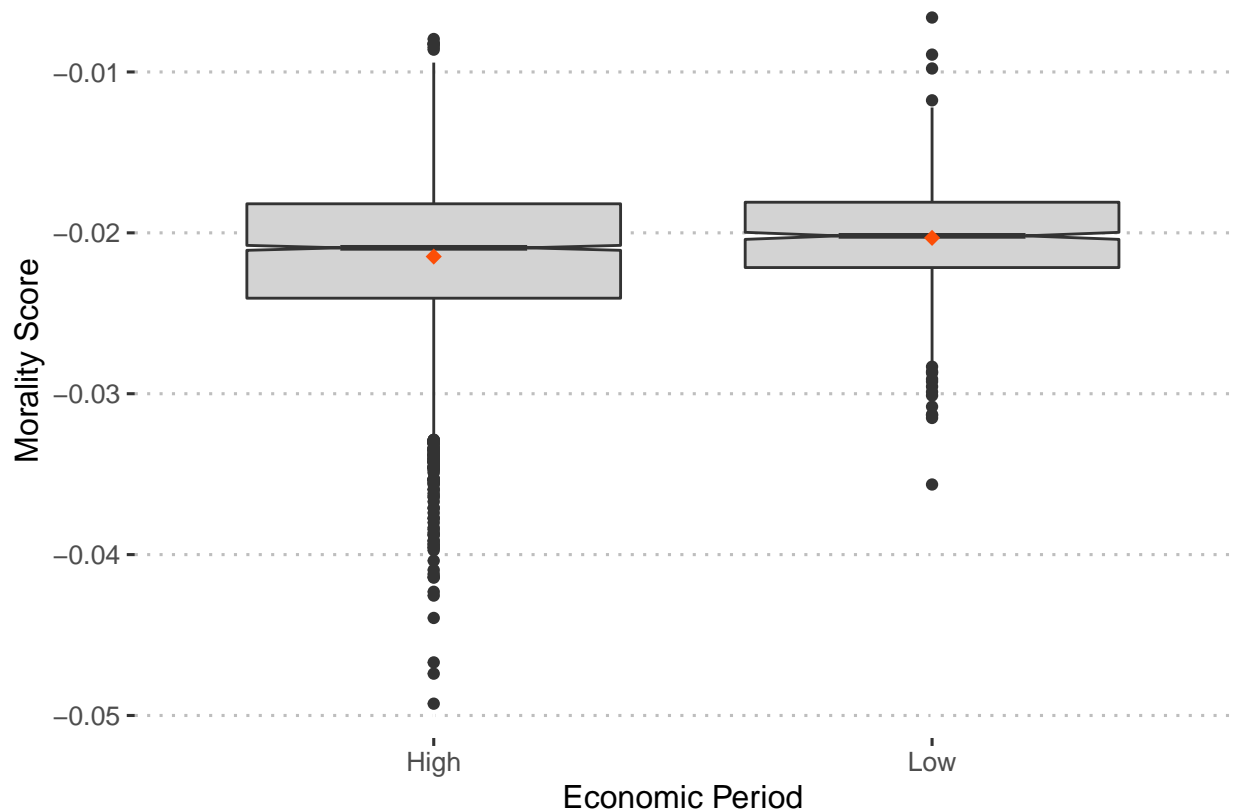
```
plot_violin2 <- function(input, x_labels) {
  return(input +
    geom_violin(aes(color = tf2_15min), trim = FALSE, position = position_dodge(0.9)) +
    geom_boxplot(aes(color = tf2_15min), width = 0.15, position = position_dodge(0.9)) +
    scale_color_manual(labels = c("High", "Low"), values = c("#00AFBB", "#E7B800")) +
    scale_x_discrete(labels=x_labels) +
    theme(legend.position="right") +
    labs(x="Moral Foundations", y = "Foundation Score", colour="Economic Period") )}
```

```
### Plot functions
### MORALITY
e <- ggplot(stocks15, aes(x = tf2_15min, y = morality_lag))
plot_boxplot(e, c("High", "Low"))
```

```
## Warning: 'fun.y' is deprecated. Use 'fun' instead.
```

```
## Warning: Removed 1 rows containing non-finite values (stat_boxplot).
```

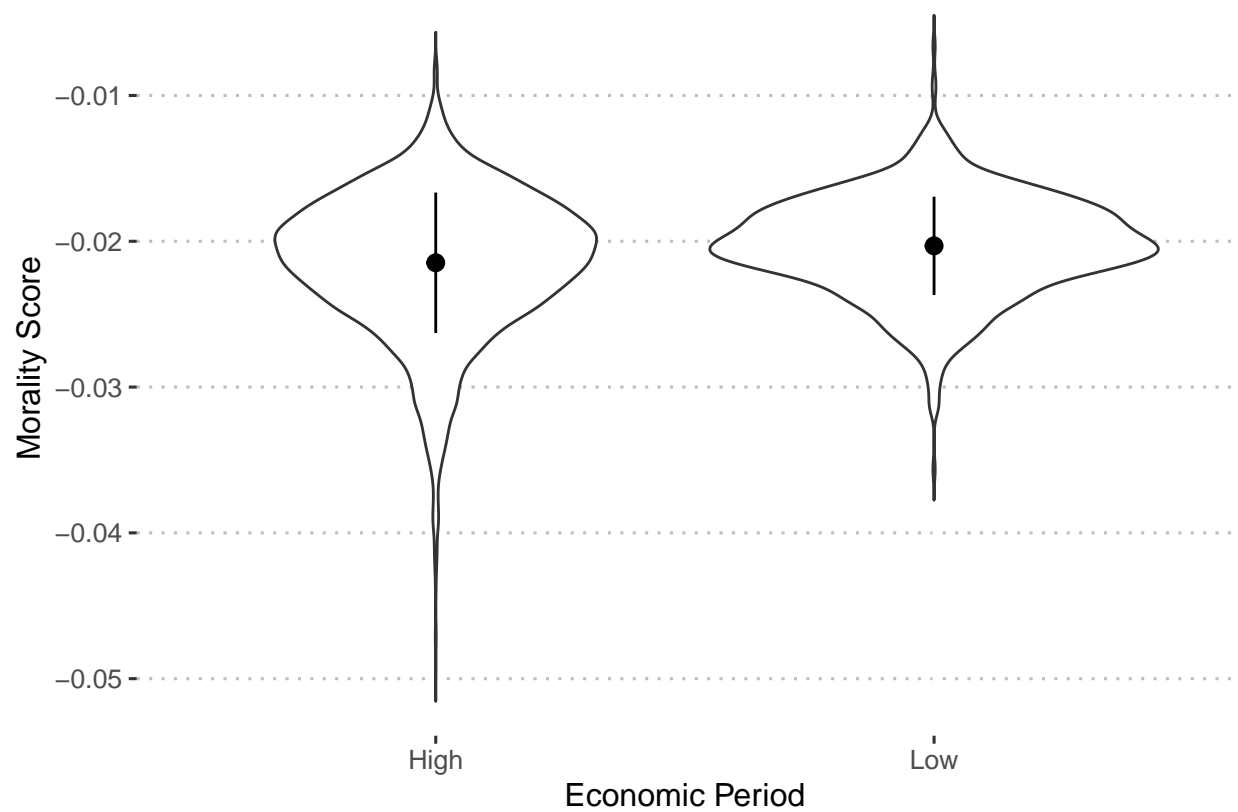
```
## Warning: Removed 1 rows containing non-finite values (stat_summary).
```



```
plot_violin(e, c("High", "Low"))
```

```
## Warning: Removed 1 rows containing non-finite values (stat_ydensity).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_summary).
```



```
### FOUNDATIONS
```

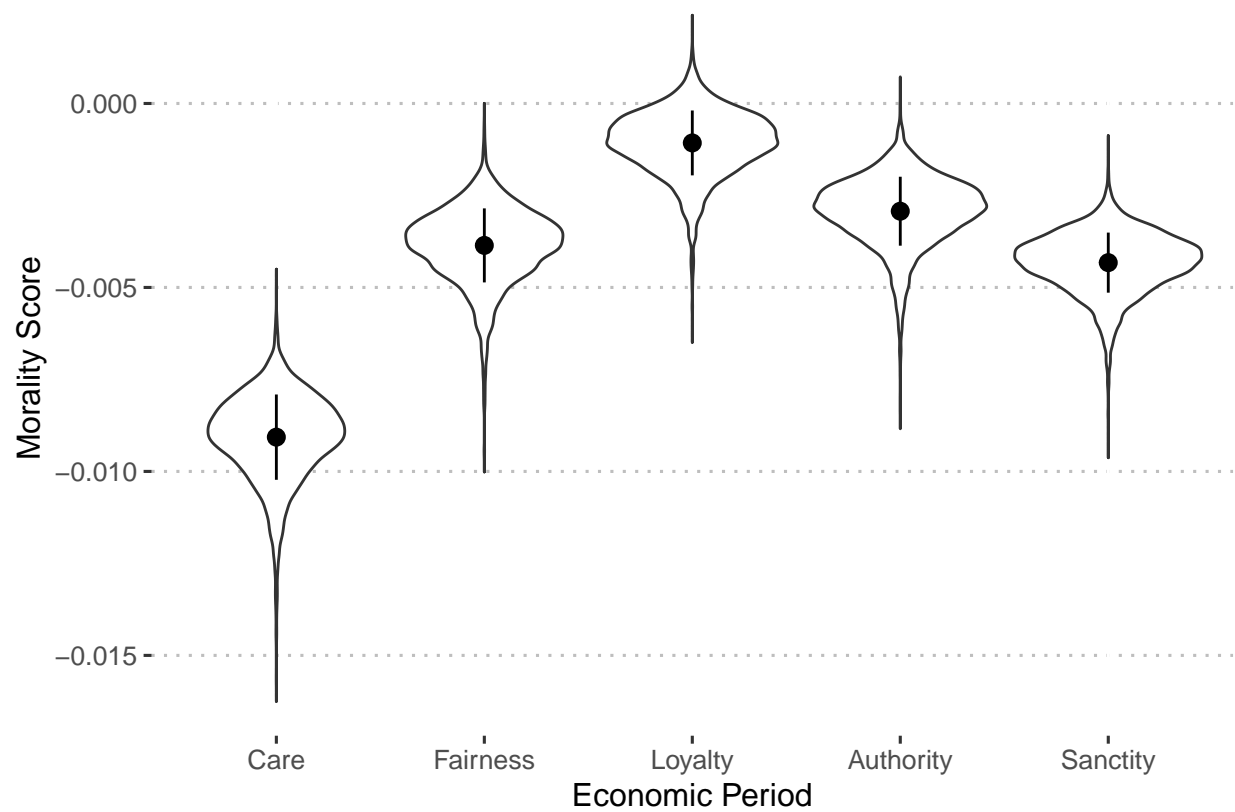
```
stocks15_foundations_long = melt(stocks15_ordered[, c(7,37:41)])
```

```
## Using tf2_15min as id variables
```

```
e1 <- ggplot(stocks15_foundations_long, aes(x = variable, y = value))
plot_violin(e1, c("Care", "Fairness", "Loyalty", "Authority", "Sanctity"))
```

```
## Warning: Removed 5 rows containing non-finite values (stat_ydensity).
```

```
## Warning: Removed 5 rows containing non-finite values (stat_summary).
```

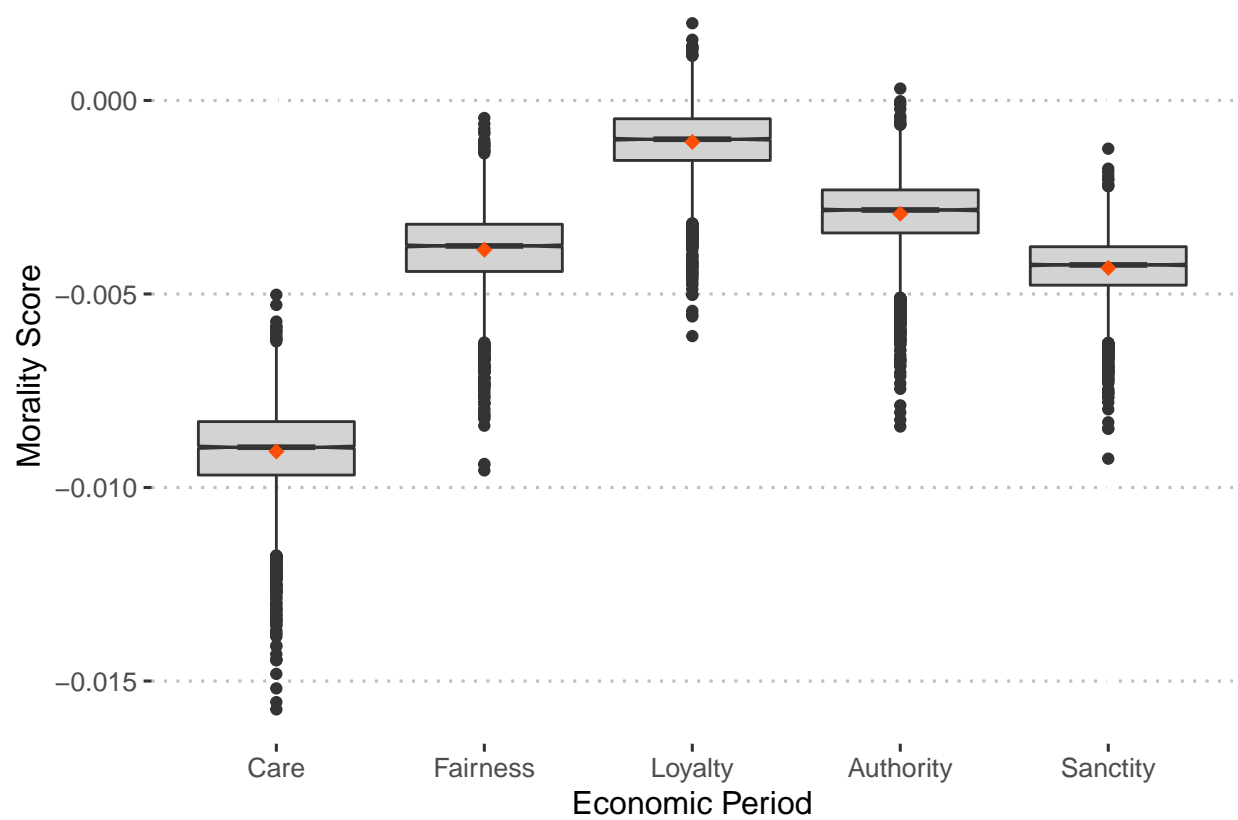



```
plot_boxplot(e1, c("Care","Fairness","Loyalty","Authority", "Sanctity"))
```

```
## Warning: 'fun.y' is deprecated. Use 'fun' instead.
```

```
## Warning: Removed 5 rows containing non-finite values (stat_boxplot).
```

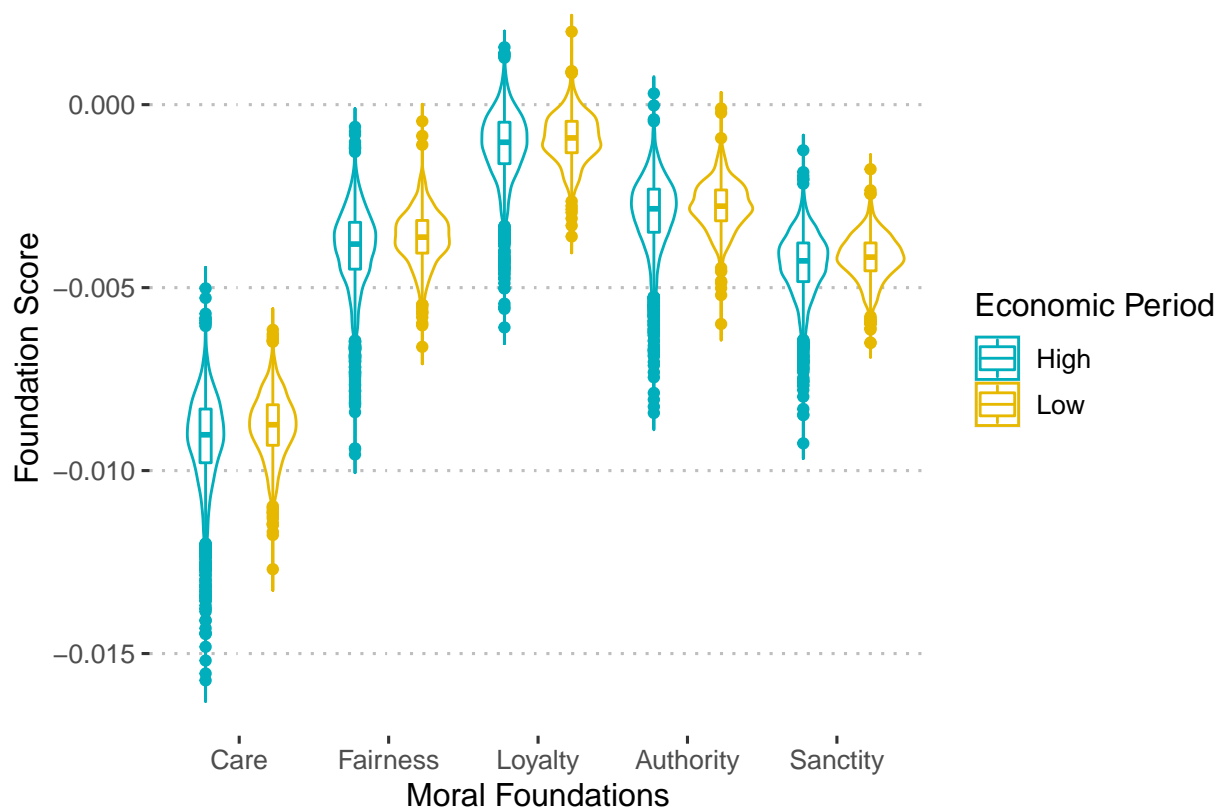
```
## Warning: Removed 5 rows containing non-finite values (stat_summary).
```



```
plot_violin2(e1, c("Care", "Fairness", "Loyalty", "Authority", "Sanctity"))
```

```
## Warning: Removed 5 rows containing non-finite values (stat_ydensity).
```

```
## Warning: Removed 5 rows containing non-finite values (stat_boxplot).
```



2.3: Scatterplots

Define plot functions

SCATTERPLOT

```
plot_scatter <- function(input, x_label) {
  return(ggplot(stocks15_ordered, aes(input, stocks_15min_diff_ln, color=tf2_15min)) +
    geom_point() +
    geom_smooth(method=lm) +
    scale_color_manual(labels = c("High", "Low"), values = c('#00AFBB', '#E7B800')) +
    theme(legend.position=c(0,1), legend.justification=c(0,1)) +
    labs(x=x_label, y = "Market Movement", color = "Economic Period") )}
```

X DENSITY

```
plot_xdensity <- function(input) {
  return(ggplot(stocks15_ordered, aes(input, fill=tf2_15min)) +
    geom_density(alpha=.5) +
    scale_fill_manual(values = c('#00AFBB', '#E7B800')) +
    theme(legend.position = "none") +
    labs(x = "") )}
```

Y DENSITY

```
plot_ydensity <- function() {
  return(ggplot(stocks15_ordered, aes(stocks_15min_diff_ln, fill=tf2_15min)) +
    geom_density(alpha=.5) +
    scale_fill_manual(values = c('#00AFBB', '#E7B800')) +
    theme(legend.position = "none") +
    labs(x = "") )}
```

Plot functions

MORALITY

```
scatterPlot <- plot_scatter(stocks15_ordered$morality_lag, "Morality")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$morality_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

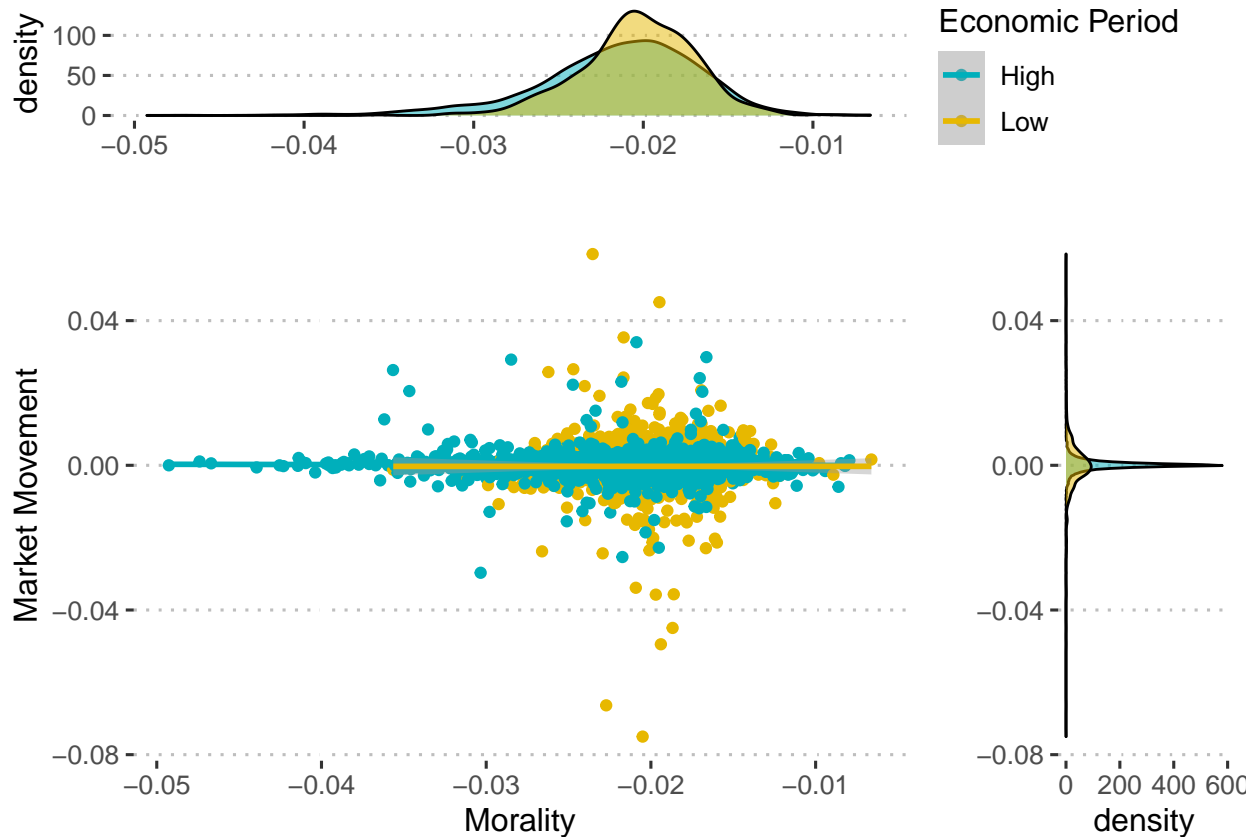
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```



FOUNDATIONS

Care

```
scatterPlot <- plot_scatter(stocks15_ordered$care_lag, "Care")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$care_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

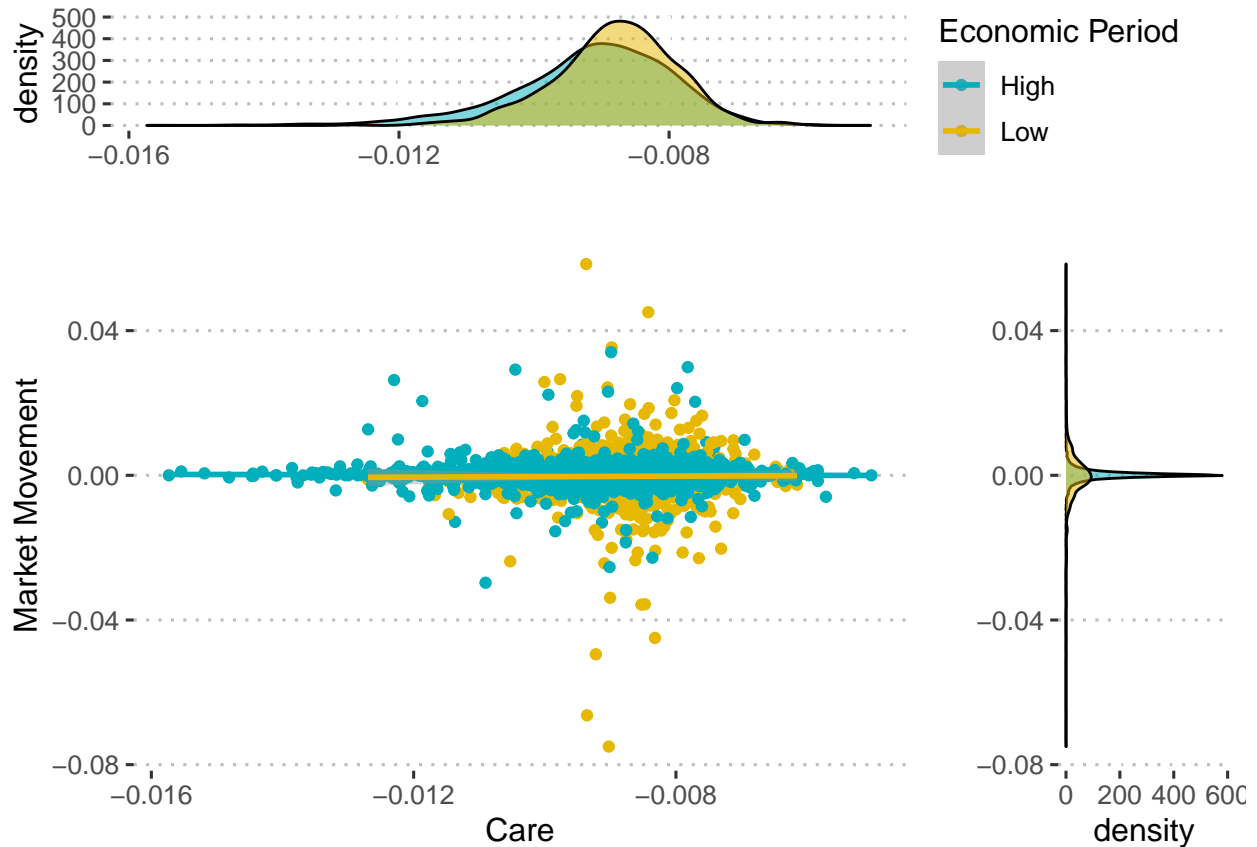
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```



```
#### Fairness
```

```
scatterPlot <- plot_scatter(stocks15_ordered$fairness_lag, "Fairness")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$fairness_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

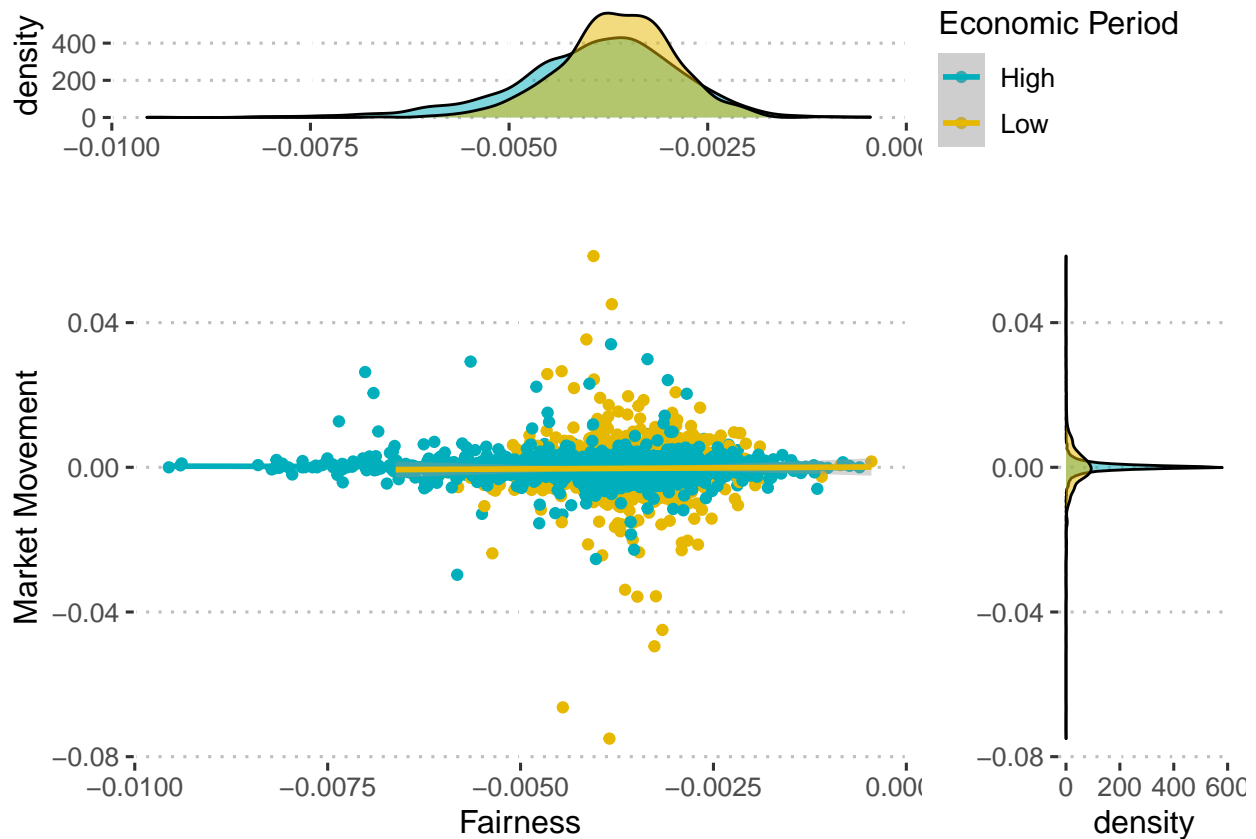
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```



Loyalty

```
scatterPlot <- plot_scatter(stocks15_ordered$loyalty_lag, "Loyalty")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$loyalty_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

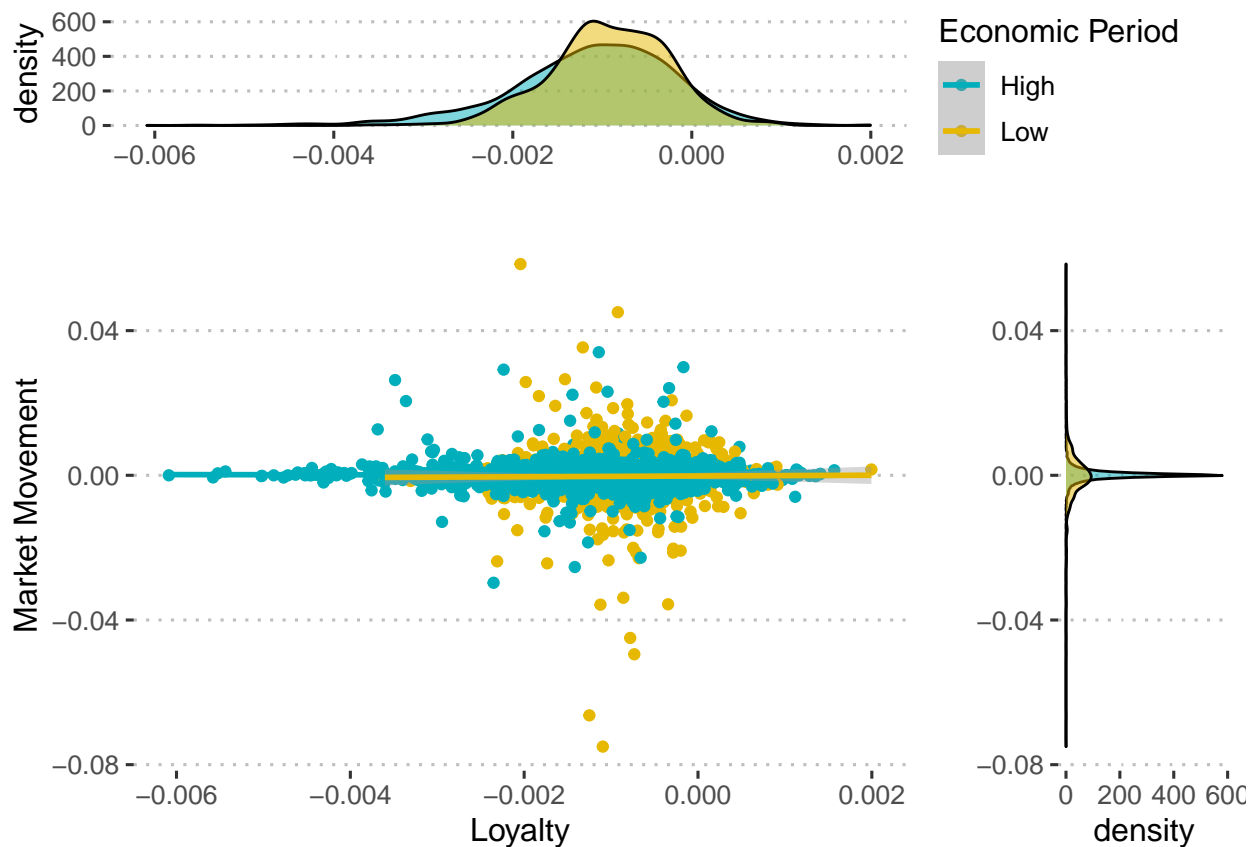
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```



Authority

```
scatterPlot <- plot_scatter(stocks15_ordered$authority_lag, "Authority")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$authority_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

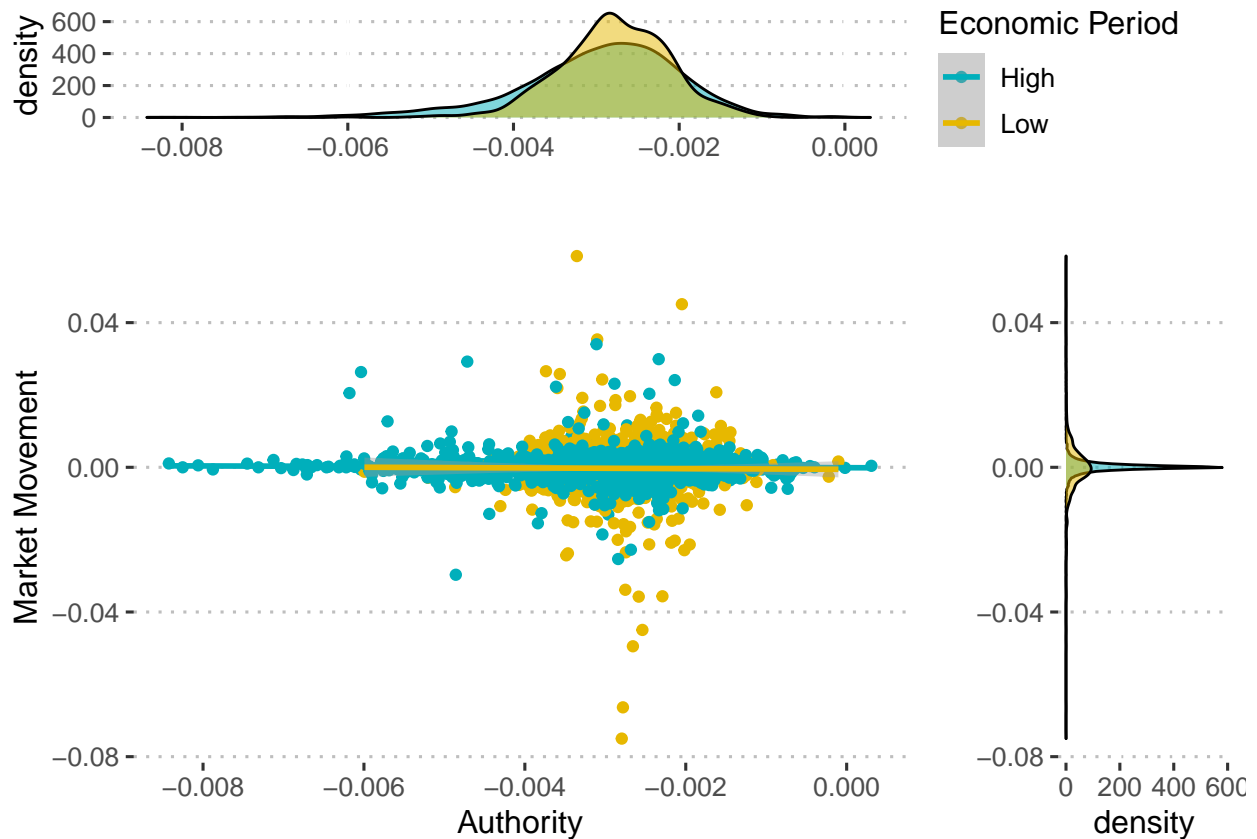
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```




```
#### Sanctity
```

```
scatterPlot <- plot_scatter(stocks15_ordered$sanctity_lag, "Sanctity")
scatterPlot2 <- scatterPlot + theme(legend.position = "none")
xdensity <- plot_xdensity(stocks15_ordered$sanctity_lag)
ydensity <- plot_ydensity() + coord_flip()
legend <- get_legend(scatterPlot)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
grid.arrange(xdensity, legend, scatterPlot2, ydensity, ncol=2, nrow=2, widths=c(4, 1.4), heights=c(1.4,
```

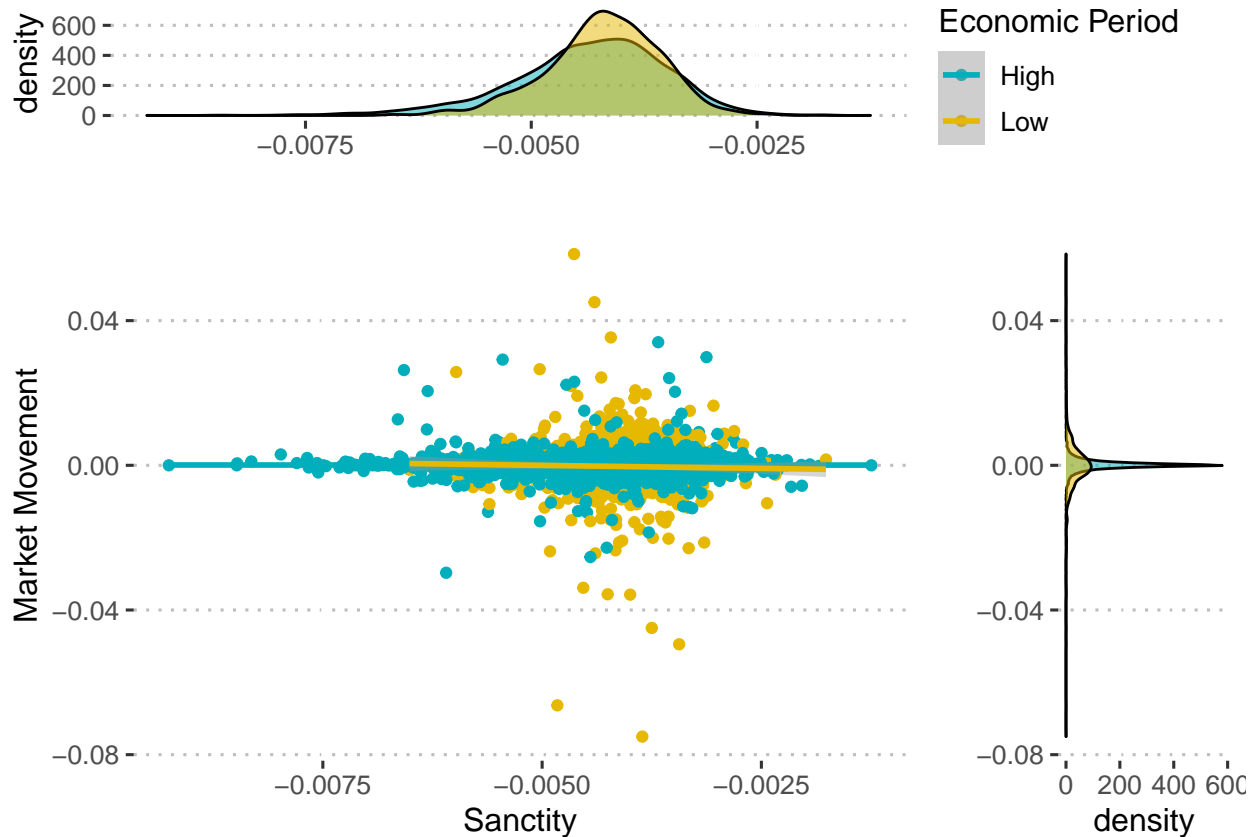
```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```

```
## 'geom_smooth()' using formula 'y ~ x'
```

```
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density).
```



```
#####
# STEP 3: FITTING AND EVALUATING MODEL
library(jtools) # summ()
library(lme4) # lme models
```

```
## Loading required package: Matrix
```

```
library(interactions) # interact()
```

```
## 3.1: Removing outliers
```

```
### Define function
```

```
outlier_rm_IQR <- function(data, df_str, col_str, threshold){
  data <- na.omit(data)
  Q <- quantile(data, probs=c(.25, .75), na.rm = FALSE) # 25/75 QUANTILES AFTER REMOVING ROW 1 (NA row)
  iqr <- IQR(data) # IQR AFTER REMOVING ROW 1 (NA row)
  upper <- Q[2]+threshold*iqr # Upper Range for outliers
  lower <- Q[1]-threshold*iqr # Lower Range for outliers
  df <- get(df_str)
  column <- get(df_str)[col_str]
  a <- subset.data.frame(df, column > lower)
  b <- subset.data.frame(df, column < upper)
  return(intersect(a,b))}
### Run function
```

```
stocks15_outrm_moralityonly <- outlier_rm_IQR(stocks15_ordered$morality_lag[-1], "stocks15_ordered", "m
```

```
## 3.2: Fitting and summarizing models
```

```
### Linear model for morality
```

```
stocks15.model.lm.outrm <- lm(
  stocks_15min_diff_ln ~ tf2_15min * morality_lag,
  data = stocks15_outrm_moralityonly)
Anova(stocks15.model.lm.outrm, type="III", test="F")
```

```
## Anova Table (Type III tests)
```

```
##
```

```
## Response: stocks_15min_diff_ln
```

	Sum Sq	Df	F value	Pr(>F)
(Intercept)	0.000001	1	0.0530	0.8179
tf2_15min	0.000001	1	0.0489	0.8249
morality_lag	0.000003	1	0.2048	0.6509
tf2_15min:morality_lag	0.000000	1	0.0273	0.8689
Residuals	0.075473	4467		

```
### Linear model for foundations
```

```
stocks15.model.lm.foundations.outrm <- lm(
  stocks_15min_diff_ln ~ tf2_15min * (care_lag +fairness_lag +loyalty_lag +authority_lag +sanctity_lag)
  data = stocks15_outrm_moralityonly)
Anova(stocks15.model.lm.foundations.outrm, type="III", test="F")
```

```
## Anova Table (Type III tests)
```

```
##
## Response: stocks_15min_diff_ln
##              Sum Sq   Df F value    Pr(>F)
## (Intercept)    0.000000   1  0.0217   0.88284
## tf2_15min      0.000003   1  0.1583   0.69070
## care_lag       0.000001   1  0.0388   0.84382
## fairness_lag   0.000012   1  0.7186   0.39665
## loyalty_lag    0.000014   1  0.8378   0.36009
## authority_lag  0.000036   1  2.1556   0.14212
## sanctity_lag   0.000031   1  1.8454   0.17439
## tf2_15min:care_lag 0.000037   1  2.1692   0.14087
## tf2_15min:fairness_lag 0.000052   1  3.0856   0.07906 .
## tf2_15min:loyalty_lag 0.000010   1  0.5740   0.44872
## tf2_15min:authority_lag 0.000006   1  0.3494   0.55448
## tf2_15min:sanctity_lag 0.000286   1 16.9971 3.812e-05 ***
## Residuals      0.075070 4459
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Mixed-effect null model

```
stocks15.model.lme.null.outrm <- lmer(
  stocks_15min_diff_ln ~ 1 + (1|day_count),
  data = stocks15_outrm_moralityonly, REML=TRUE)
Anova(stocks15.model.lme.null.outrm, type="III", test="F")
```

```
## Analysis of Deviance Table (Type III Wald F tests with Kenward-Roger df)
##
## Response: stocks_15min_diff_ln
##              F Df Df.res Pr(>F)
## (Intercept) 3e-04  1 171.84 0.9868
```

Mixed-effect model for morality

```
stocks15.model.lme.outrm <- lmer(
  stocks_15min_diff_ln ~ (1 + season_intraday_15min + tf2_15min*morality_lag + (1 + season_intraday_15min|morality_lag))
  data = stocks15_outrm_moralityonly, REML = TRUE)
```

```
## Warning: Some predictor variables are on very different scales: consider rescaling
```

```
## boundary (singular) fit: see ?isSingular
```

```
Anova(stocks15.model.lme.outrm, type="III", test="F")
```

```
## Analysis of Deviance Table (Type III Wald F tests with Kenward-Roger df)
##
## Response: stocks_15min_diff_ln
##              F Df Df.res Pr(>F)
## (Intercept)    0.2288  1 1308.1 0.6325
## season_intraday_15min 0.0006  1  174.3 0.9801
## tf2_15min      0.0843  1 3028.1 0.7715
## morality_lag    0.3969  1 2316.8 0.5288
## tf2_15min:morality_lag 0.0173  1 3387.7 0.8955
```

```

### Mixed-effect model for foundations
stocks15.model.lme.foundations.outrm <- lmer(
  stocks_15min_diff_ln ~ (1 + season_intraday_15min +
                          tf2_15min*(care_lag + fairness_lag + loyalty_lag + authority_lag + sancti
                          ( 1 + season_intraday_15min | day_count))),
  data = stocks15_outrm_moralityonly, REML = TRUE)

## Warning: Some predictor variables are on very different scales: consider rescaling

## boundary (singular) fit: see ?isSingular

Anova(stocks15.model.lme.foundations.outrm, type="III", test="F")

## Analysis of Deviance Table (Type III Wald F tests with Kenward-Roger df)
##
## Response: stocks_15min_diff_ln
##
##           F Df Df.res    Pr(>F)
## (Intercept)    0.0573  1 3644.7  0.81086
## season_intraday_15min  0.0096  1  176.9  0.92195
## tf2_15min      0.2324  1 3811.9  0.62977
## care_lag       0.1293  1 2995.1  0.71921
## fairness_lag    1.0960  1 4170.9  0.29521
## loyalty_lag     0.6462  1 3741.5  0.42151
## authority_lag    2.2186  1 3692.7  0.13644
## sanctity_lag     2.1961  1 2162.5  0.13850
## tf2_15min:care_lag  1.7948  1 3989.0  0.18041
## tf2_15min:fairness_lag  4.5958  1 4056.8  0.03211 *
## tf2_15min:loyalty_lag  0.5404  1 3705.7  0.46230
## tf2_15min:authority_lag  0.8233  1 3698.7  0.36427
## tf2_15min:sanctity_lag 18.0085  1 3847.7 2.251e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## 3.3: Model comparisons

### Comparisons between linear and mixed-effects models
##### Morality models
anova(stocks15.model.lme.outrm, stocks15.model.lm.outrm, type="Chisq")

## refitting model(s) with ML (instead of REML)

## Data: stocks15_outrm_moralityonly
## Models:
## stocks15.model.lm.outrm: stocks_15min_diff_ln ~ tf2_15min * morality_lag
## stocks15.model.lme.outrm: stocks_15min_diff_ln ~ (1 + season_intraday_15min + tf2_15min * morality_l
##
##           npar      AIC      BIC logLik deviance Chisq Df Pr(>Chisq)
## stocks15.model.lm.outrm      5 -36435 -36403  18223   -36445
## stocks15.model.lme.outrm      9 -36516 -36459  18267   -36534 89.06  4 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Foundations models

```
anova(stocks15.model.lme.foundations.outrm, stocks15.model.lm.foundations.outrm, type="Chisq")
```

```
## refitting model(s) with ML (instead of REML)
```

```
## Data: stocks15_outrm_moralityonly
```

```
## Models:
```

```
## stocks15.model.lm.foundations.outrm: stocks_15min_diff_ln ~ tf2_15min * (care_lag + fairness_lag + l
```

```
## stocks15.model.lme.foundations.outrm: stocks_15min_diff_ln ~ (1 + season_intraday_15min + tf2_15min *
```

```
##          npar      AIC      BIC logLik deviance Chisq Df Pr(>Chisq)
```

```
## stocks15.model.lm.foundations.outrm      13 -36443 -36360 18235 -36469
```

```
## stocks15.model.lme.foundations.outrm      17 -36527 -36419 18281 -36561 92.33  4 < 2.2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Comparisons between morality and foundations models

Linear models

```
anova(stocks15.model.lm.outrm, stocks15.model.lm.foundations.outrm)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Model 1: stocks_15min_diff_ln ~ tf2_15min * morality_lag
```

```
## Model 2: stocks_15min_diff_ln ~ tf2_15min * (care_lag + fairness_lag +
```

```
##   loyalty_lag + authority_lag + sanctity_lag)
```

```
##   Res.Df      RSS Df Sum of Sq      F Pr(>F)
```

```
## 1    4467 0.075473
```

```
## 2    4459 0.075070  8 0.00040303 2.9924 0.00239 **
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Mixed-effects models

```
anova(stocks15.model.lme.null.outrm, stocks15.model.lme.outrm, stocks15.model.lme.foundations.outrm)
```

```
## refitting model(s) with ML (instead of REML)
```

```
## Data: stocks15_outrm_moralityonly
```

```
## Models:
```

```
## stocks15.model.lme.null.outrm: stocks_15min_diff_ln ~ 1 + (1 | day_count)
```

```
## stocks15.model.lme.outrm: stocks_15min_diff_ln ~ (1 + season_intraday_15min + tf2_15min * morality_l
```

```
## stocks15.model.lme.foundations.outrm: stocks_15min_diff_ln ~ (1 + season_intraday_15min + tf2_15min *
```

```
##          npar      AIC      BIC logLik deviance Chisq Df Pr(>Chisq)
```

```
## stocks15.model.lme.null.outrm           3 -36434 -36415 18220 -36440
```

```
## stocks15.model.lme.outrm                9 -36516 -36459 18267 -36534 94.102  6 < 2.2e-16 ***
```

```
## stocks15.model.lme.foundations.outrm     17 -36527 -36419 18281 -36561 27.210  8 0.0006504 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

3.4: Standardized betas

Morality mixed-effect model

```
summ(stocks15.model.lme.outrm, scale=TRUE, transform.response=TRUE, confint=TRUE, digits=3)
```

```
## boundary (singular) fit: see ?isSingular
```

```
## MODEL INFO:
```

```
## Observations: 4471
```

```
## Dependent Variable: stocks_15min_diff_ln
```

```
## Type: Mixed effects linear regression
```

```
##
```

```
## MODEL FIT:
```

```
## AIC = 12638.266, BIC = 12695.914
```

```
## Pseudo-R2 (fixed effects) = 0.000
```

```
## Pseudo-R2 (total) = 0.054
```

```
##
```

```
## FIXED EFFECTS:
```

```
## -----
```

	Est.	2.5%	97.5%	t val.	d.f.	p
(Intercept)	0.007	-0.032	0.047	0.356	276.009	0.722
season_intraday_15min	0.001	-0.039	0.040	0.025	208.109	0.980
tf2_15min	-0.035	-0.120	0.049	-0.827	629.269	0.409
morality_lag	-0.010	-0.042	0.022	-0.632	3601.001	0.527
tf2_15min:morality_lag	-0.006	-0.097	0.085	-0.132	4369.268	0.895

```
## -----
```

```
##
```

```
## p values calculated using Satterthwaite d.f.
```

```
##
```

```
## RANDOM EFFECTS:
```

```
## -----
```

Group	Parameter	Std. Dev.
day_count	(Intercept)	0.148
day_count	season_intraday_15min	0.180
Residual		0.973

```
## -----
```

```
##
```

```
## Grouping variables:
```

```
## -----
```

Group	# groups	ICC
day_count	173	0.022

```
## -----
```

```
##
```

```
## Continuous variables are mean-centered and scaled by 1 s.d.
```

```
### Foundations mixed-effect model
```

```
summ(stocks15.model.lme.foundations.outrm, scale=TRUE, transform.response=TRUE, confint=TRUE, digits=3)
```

```
## boundary (singular) fit: see ?isSingular
```

```
## MODEL INFO:
```

```
## Observations: 4471
```

```
## Dependent Variable: stocks_15min_diff_ln
```

```
## Type: Mixed effects linear regression
```

```
##
```

```

## MODEL FIT:
## AIC = 12657.014, BIC = 12765.905
## Pseudo-R2 (fixed effects) = 0.006
## Pseudo-R2 (total) = 0.061
##
## FIXED EFFECTS:
## -----
##               Est.      2.5%      97.5%      t val.      d.f.      p
## -----
## (Intercept)      0.006     -0.033      0.046      0.306      271.633    0.760
## season_intraday_15min 0.002     -0.038      0.042      0.098      210.110    0.922
## tf2_15min       -0.059     -0.145      0.026     -1.362      663.167    0.174
## care_lag         0.015     -0.067      0.098      0.361     4046.750    0.718
## fairness_lag     -0.050     -0.142      0.043     -1.049     4389.643    0.294
## loyalty_lag       0.037     -0.053      0.126      0.806     4303.918    0.421
## authority_lag    -0.065     -0.150      0.020     -1.493     4314.688    0.136
## sanctity_lag      0.054     -0.017      0.126      1.488     3761.342    0.137
## tf2_15min:care_lag  0.153     -0.070      0.377      1.342     4287.126    0.180
## tf2_15min:fairness_lag 0.265      0.023      0.506      2.148     4357.006    0.032
## tf2_15min:loyalty_lag 0.087     -0.144      0.318      0.737     4340.746    0.461
## tf2_15min:authority_lag -0.102     -0.321      0.117     -0.910     4228.991    0.363
## tf2_15min:sanctity_lag -0.425     -0.621     -0.229     -4.253     4347.225    0.000
## -----
##
## p values calculated using Satterthwaite d.f.
##
## RANDOM EFFECTS:
## -----
##      Group      Parameter      Std. Dev.
## -----
## day_count      (Intercept)      0.149
## day_count      season_intraday_15min 0.182
## Residual                          0.970
## -----
##
## Grouping variables:
## -----
##      Group      # groups      ICC
## -----
## day_count      173      0.023
## -----
##
## Continuous variables are mean-centered and scaled by 1 s.d.

```

```

## 3.6: Interaction plots
### Define interaction plots
#### Morality plot
plot1 <- interact_plot(stocks15.model.lme.outrm,
  pred = morality_lag,
  modx = tf2_15min,
  plot.points = TRUE,
  linearity.check = FALSE,
  x.label = "Morality",
  y.label = "Difference in Market Movement",

```

```

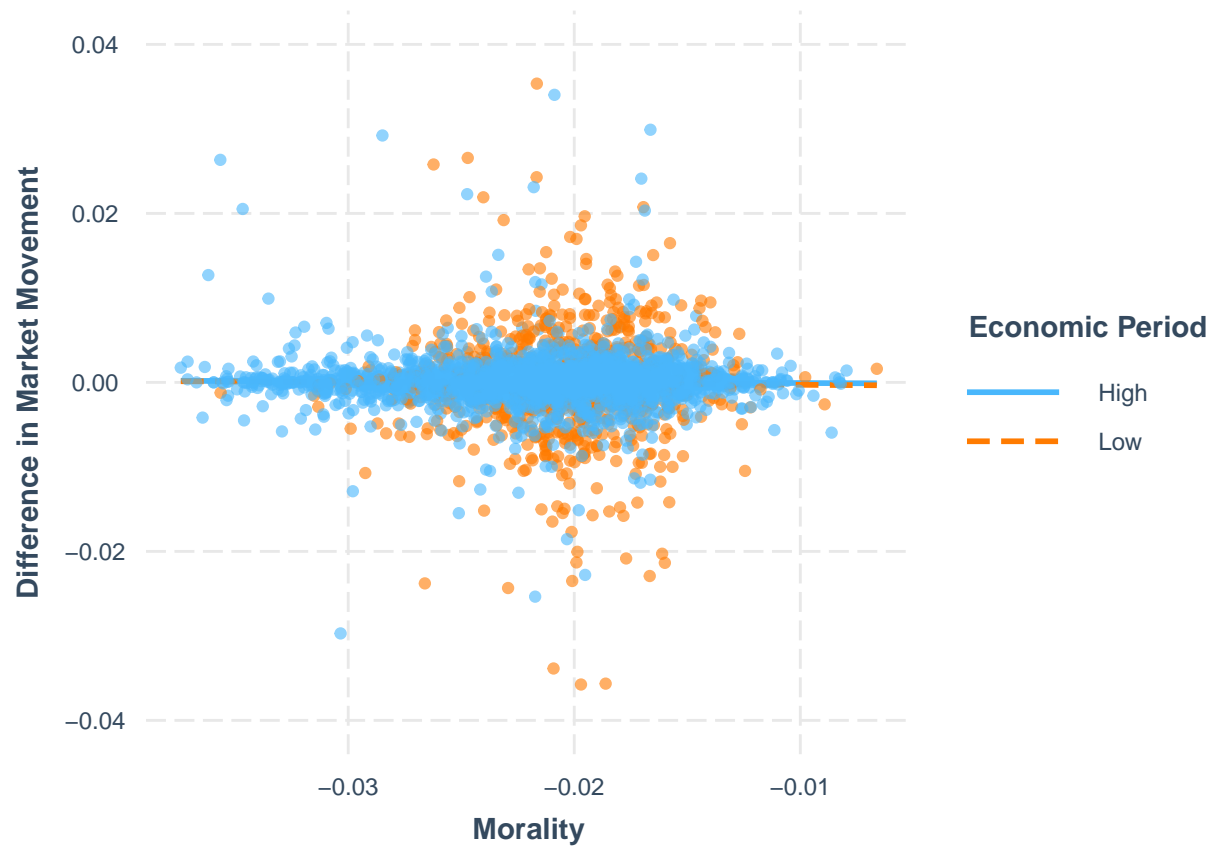
        modx.labels = c("High","Low"),
        legend.main = "Economic Period") +
  ylim(-0.04,0.04)
plot1a <- plot1 + theme(legend.position = "none")
#### Fairness plot
plot2 <- interact_plot(stocks15.model.lme.foundations.outrm,
  pred = fairness_lag,
  modx = tf2_15min,
  plot.points = TRUE,
  linearity.check = FALSE,
  x.label = "Fairness",
  y.label = "Difference in Market Movement") +
  ylim(-0.04,0.04) +
  theme(legend.position = "none")
#### Sanctity plot
plot3 <- interact_plot(stocks15.model.lme.foundations.outrm,
  pred = sanctity_lag,
  modx = tf2_15min,
  plot.points = TRUE,
  linearity.check = FALSE,
  x.label = "Sanctity",
  y.label = "Difference in Market Movement") +
  ylim(-0.04,0.04) +
  theme(legend.position = "none")
### Plot interactions
legend <- get_legend(plot1)

```

```
## Warning: Removed 6 rows containing missing values (geom_point).
```

```
grid.arrange(plot1)
```

```
## Warning: Removed 6 rows containing missing values (geom_point).
```

```
grid.arrange(plot2, plot3, legend, ncol=3, widths=c(3,3,1.5))
```

```
## Warning: Removed 6 rows containing missing values (geom_point).
```

```
## Warning: Removed 6 rows containing missing values (geom_point).
```

