

# R Notebook

```
require(dplyr)
```

```
## Loading required package: dplyr
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##     filter, lag
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
```

```
require(ggplot2)
```

```
## Loading required package: ggplot2
```

```
require(xts)
```

```
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##     as.Date, as.Date.numeric
##
## Attaching package: 'xts'
## The following objects are masked from 'package:dplyr':
##
##     first, last
```

```
require(rugarch)
```

```
## Loading required package: rugarch
## Loading required package: parallel
##
## Attaching package: 'rugarch'
## The following object is masked from 'package:stats':
##
##     sigma
```

```
require(PerformanceAnalytics)
```

```
## Loading required package: PerformanceAnalytics
```

```

##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##      legend
require(quantmod)

## Loading required package: quantmod
## Loading required package: TTR
## Registered S3 method overwritten by 'quantmod':
##   method      from
##   as.zoo.data.frame zoo

## Version 0.4-0 included new data defaults. See ?getSymbols.
library(skewt)

read_funds <- function(lf) {
  dfs <- list()
  browser()
  for (f in 1:length(lf)) {
    cond <- substr(lf[f], nchar(lf[f])-3, nchar(lf[f])) == '.csv'
    if (cond) {
      temp <- data.frame(read.csv(lf[f], stringsAsFactors = FALSE))
      temp$Date <- as.Date(temp$Date)
      temp$Close <- as.numeric(temp$Close)
      dfs[[f]] <- temp[,c(1,5)]
    }
  }

  return(dfs)
}

funds_names <- c("Vanguard", "Blackrock", "Statestreet",
                 "JPMorgan", "Bankmellon", "Allianz")

# Data loading
df <- read.csv('Funds.csv', stringsAsFactors = F)
df <- df[,-1]
df$Date <- as.Date(df$Date)

# Transforming data from df to xts
funds <- xts(df[,2:ncol(df)], order.by = df$Date)

rm(df)

# Subsetting original data for April of 2020 year
funds_red <- funds['/202004']

get_volatiles <- function(funds, width = 22, time_scale = 1, funds_names) {
  vol_df <- data.frame()

  for (i in 1:ncol(funds)) {
    fund <- funds[,i]

```

```

    temp <- rollapply(data = fund, width = 22, FUN = 'sd.annualized', scale = time_scale)

    if (nrow(vol_df) == 0) {
      vol_df <- temp
    } else {
      vol_df <- cbind(vol_df, temp)
    }

  }
  names(vol_df) <- funds_names
  return(vol_df)
}

visualize_funds_lines <- function(funds, y_axis_label = 'Volatility') {

  for (i in 1:ncol(funds)) {
    temp <- funds[,i]
    temp_mean <- mean(temp, na.rm = TRUE)

    p <- ggplot(temp, aes(x = index(temp), y = temp)) +
      geom_line(aes(color = 'Volatility')) +
      geom_hline(aes(yintercept = temp_mean, color = 'Mean'),
                  size=.5, linetype='dashed') +
      geom_text( aes( min(index(temp)) , temp_mean, label = round(temp_mean, 4), vjust = 2)) +
      labs(x = 'Date', y = y_axis_label, title = names(funds)[i]) +
      theme(plot.title = element_text(hjust = 0.5))

    suppressMessages(suppressWarnings(print(p)))
  }
}

visualize_funds_hist <- function(funds, x_axis_label = 'Return') {
  for (i in 1:ncol(funds)) {
    temp <- funds[,i]
    title <- paste(funds_names[i], 'returns')

    chart.Histogram(temp,
                     methods = c('add.normal', 'add.density'),
                     main = title)

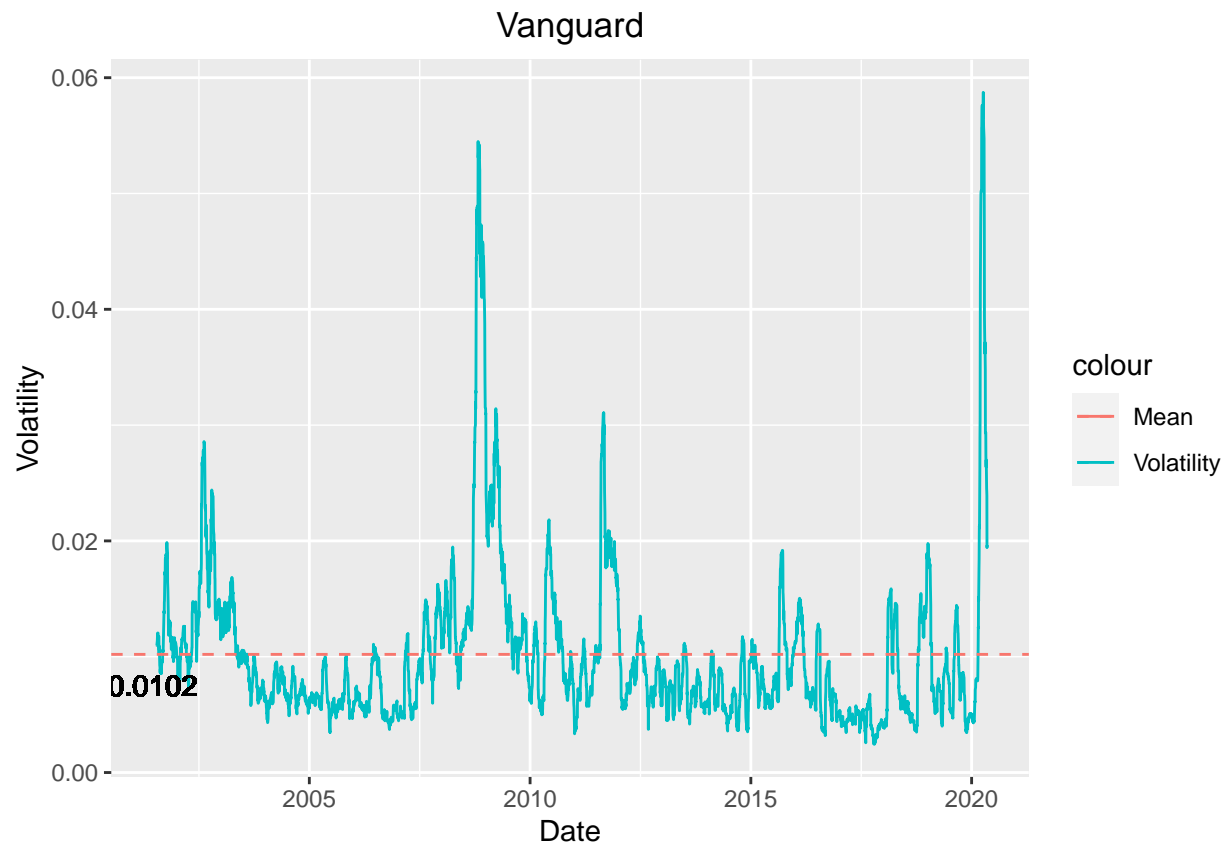
    temp <- (temp - mean(temp, na.rm = T))/sd(temp, na.rm = T)
    title <- paste('Standardized', title)

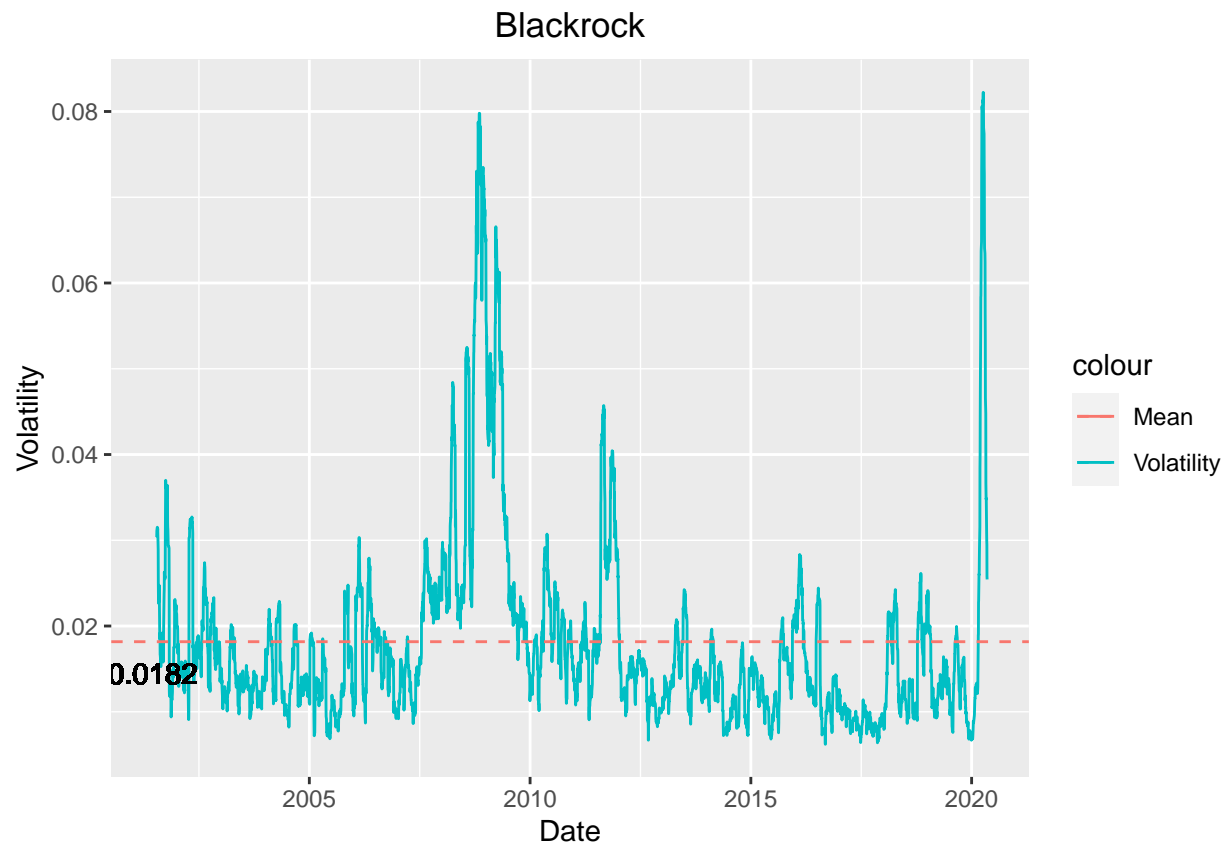
    chart.Histogram(temp,
                     methods = c('add.normal', 'add.density'),
                     main = title)
  }
}

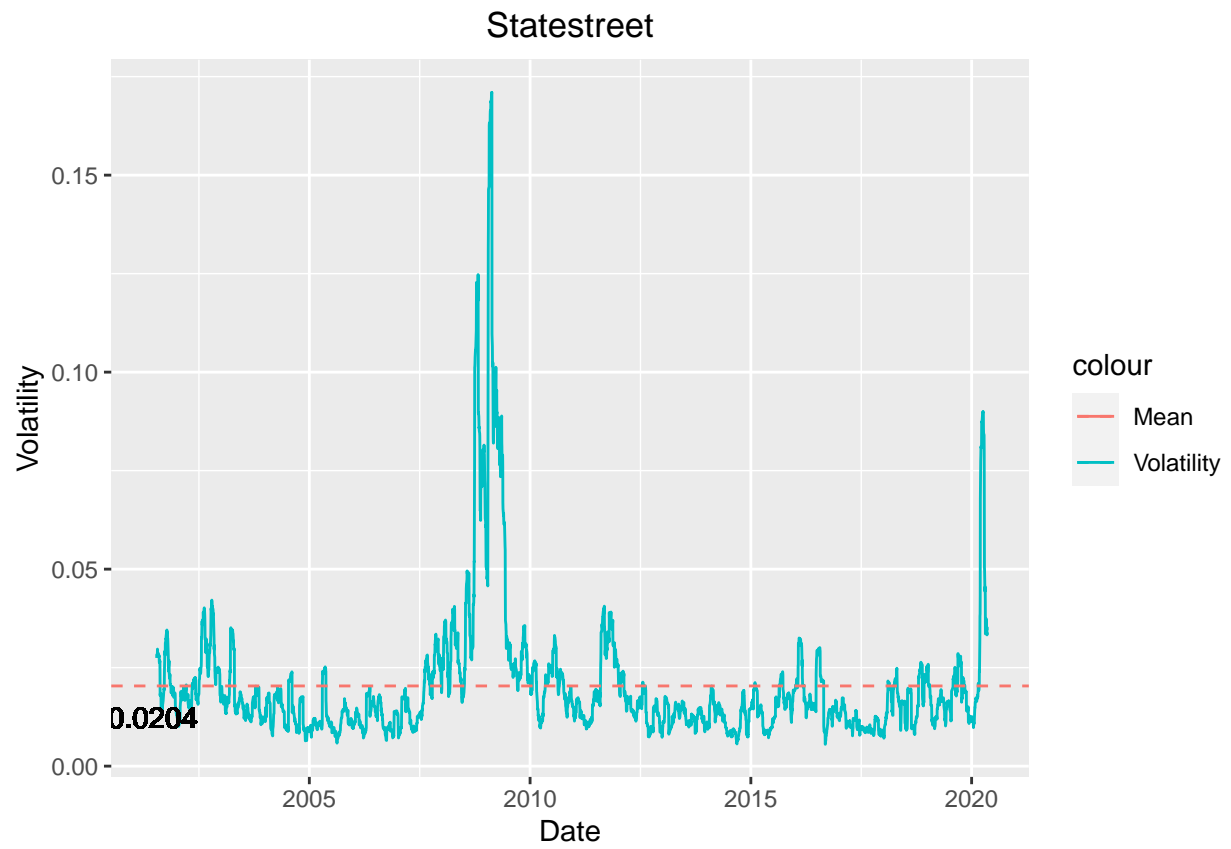
vol_df <- get_volatiles(funds, funds_names = funds_names)

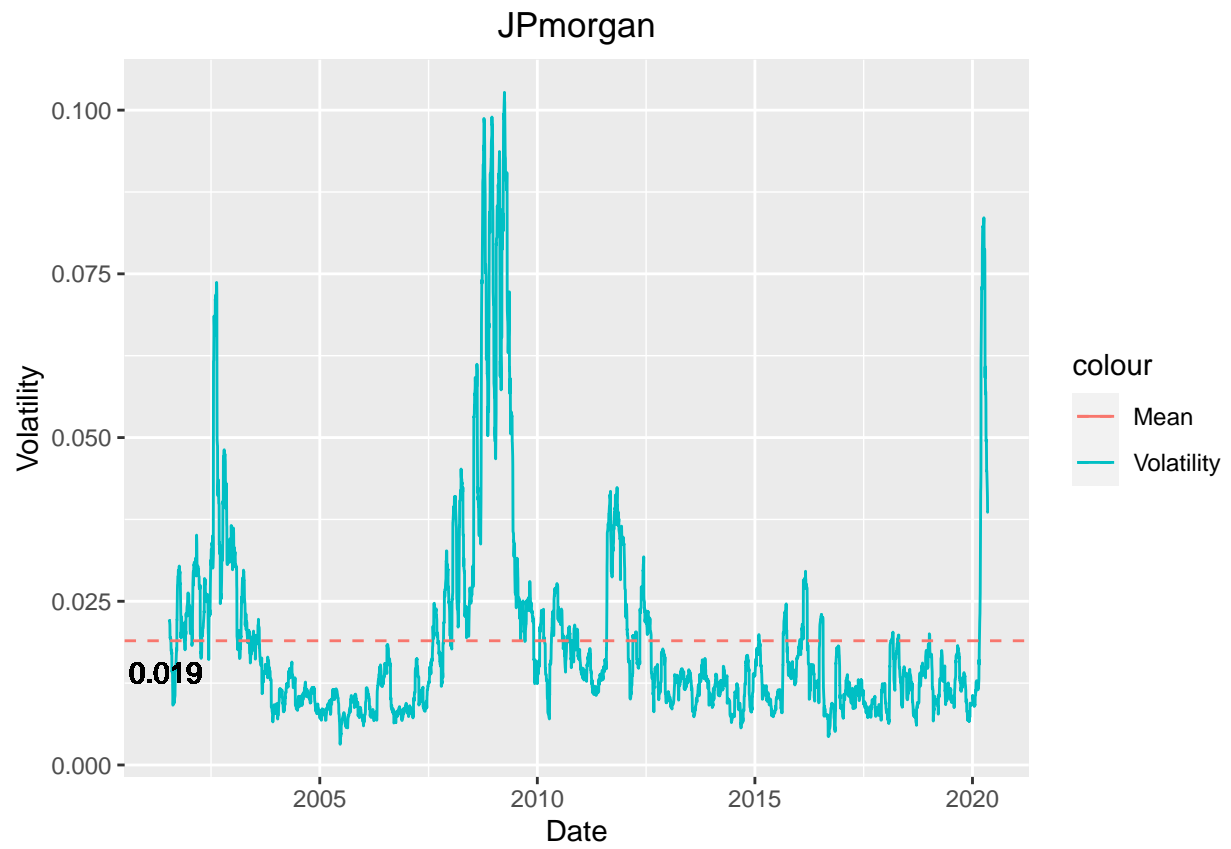
visualize_funds_lines(vol_df)

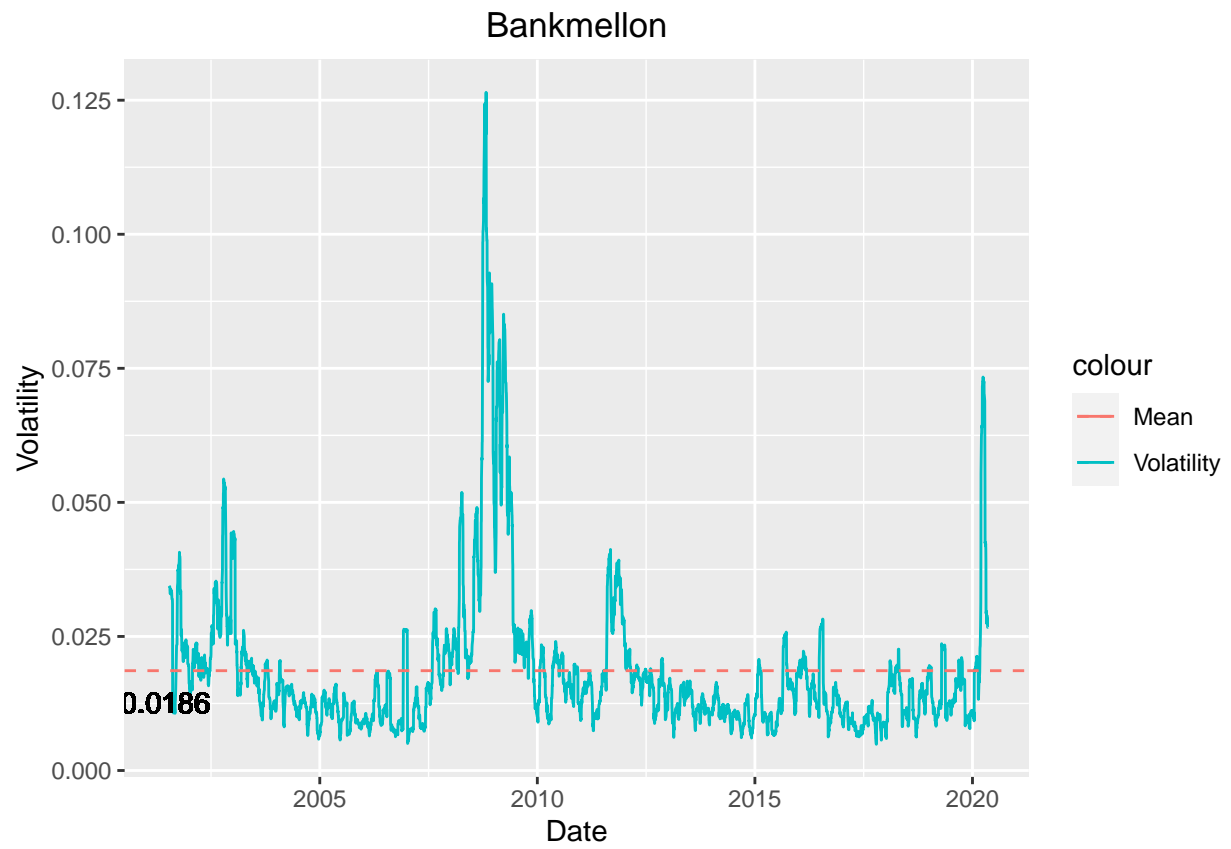
```



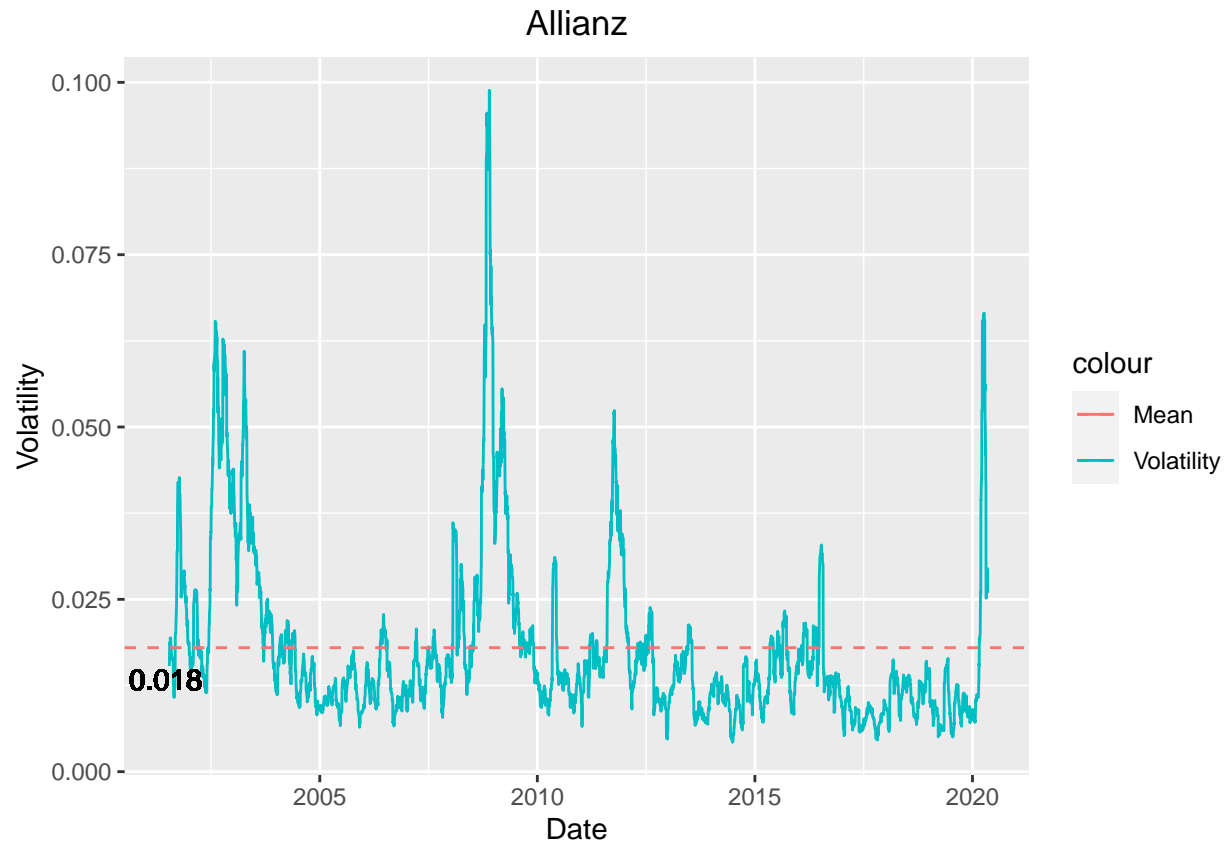






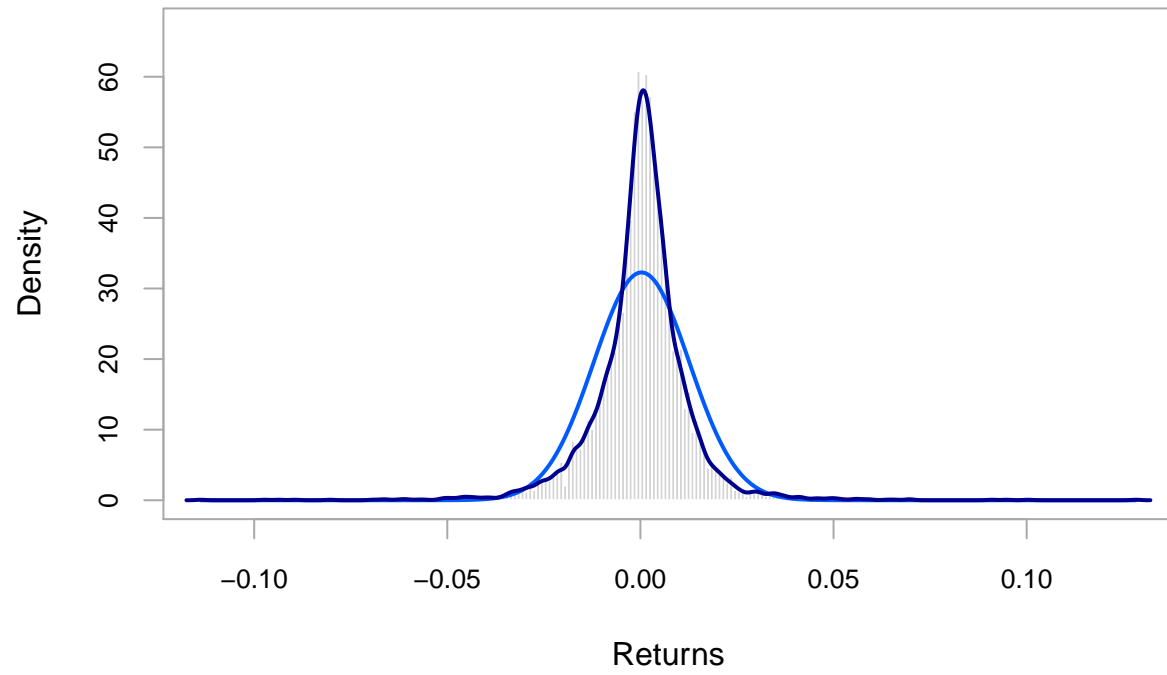




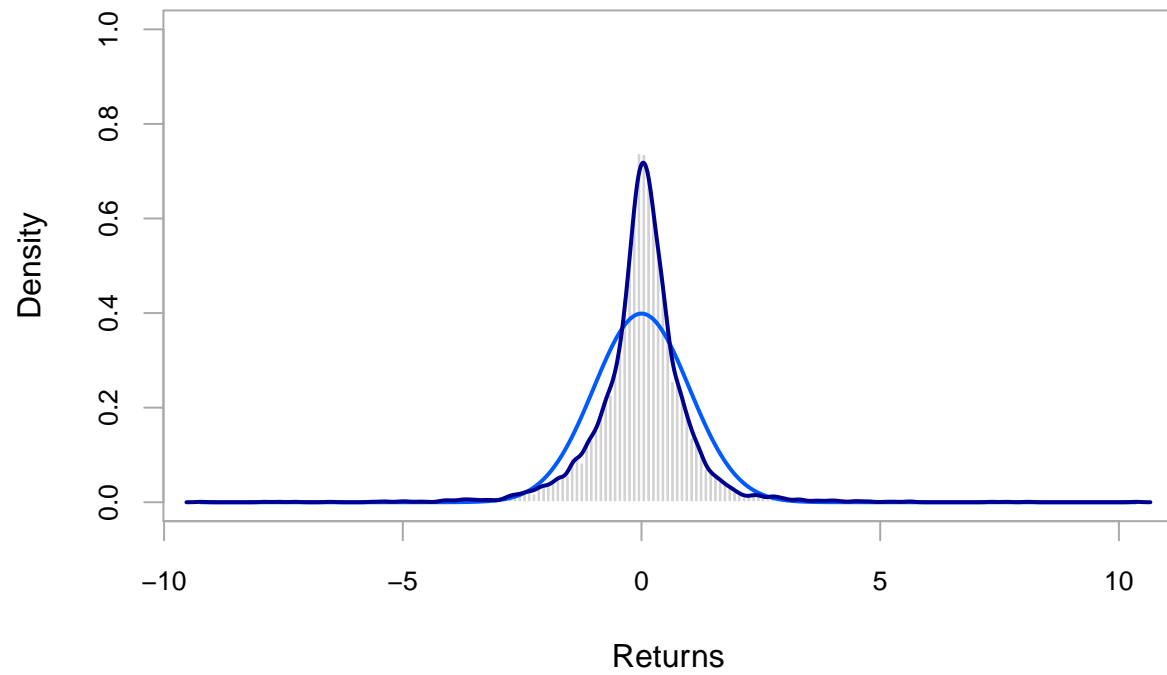


```
visualize_funds_hist(funds)
```

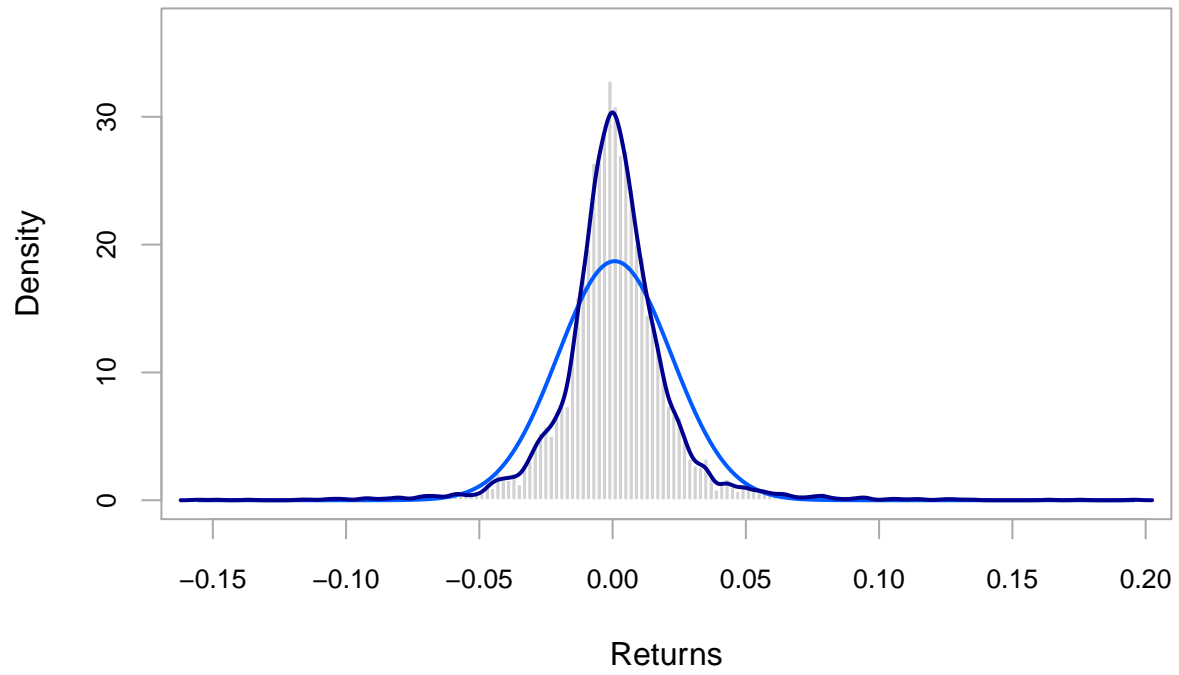
### Vanguard returns



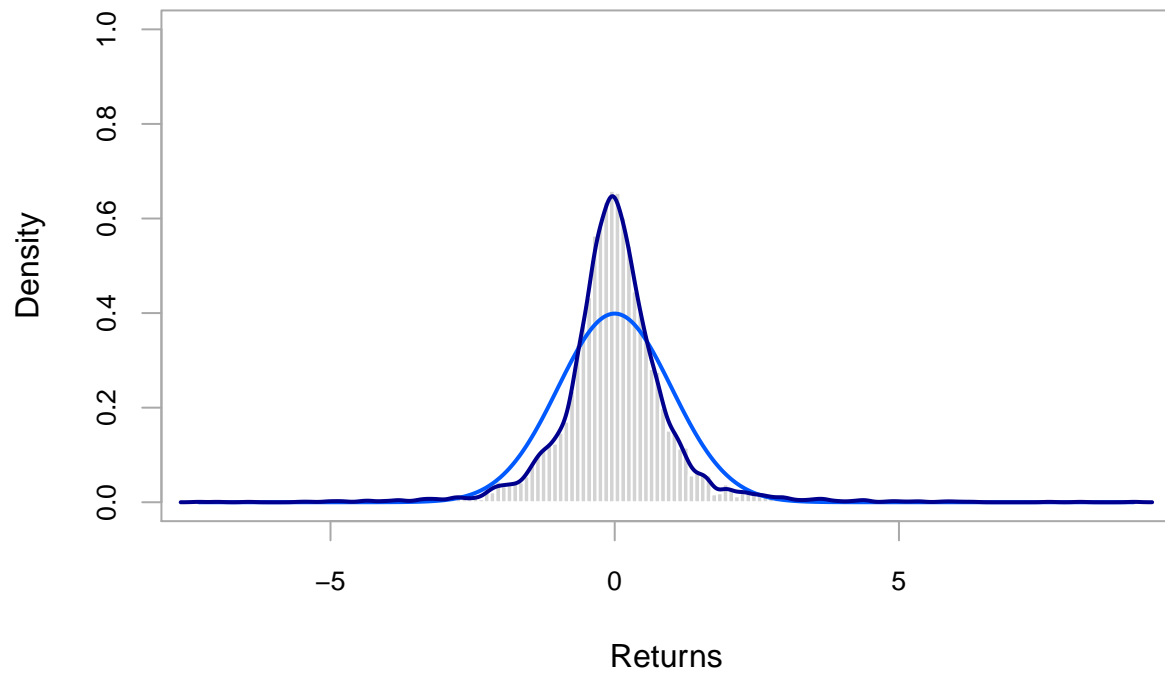
### Standardized Vanguard returns



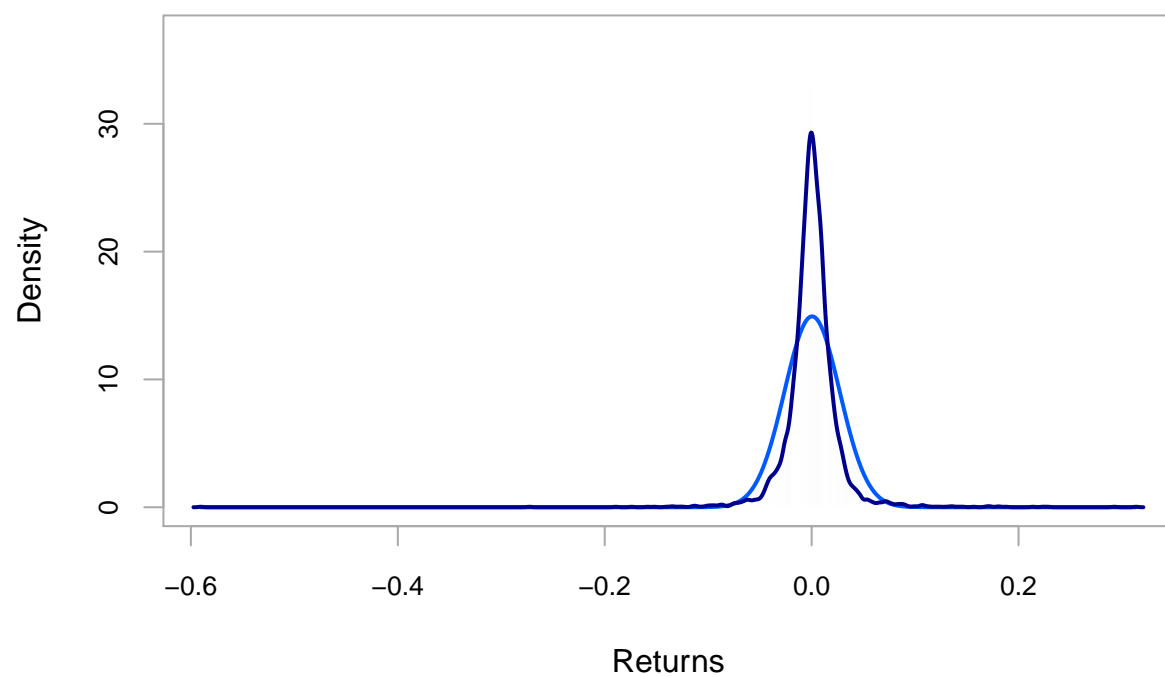
### Blackrock returns



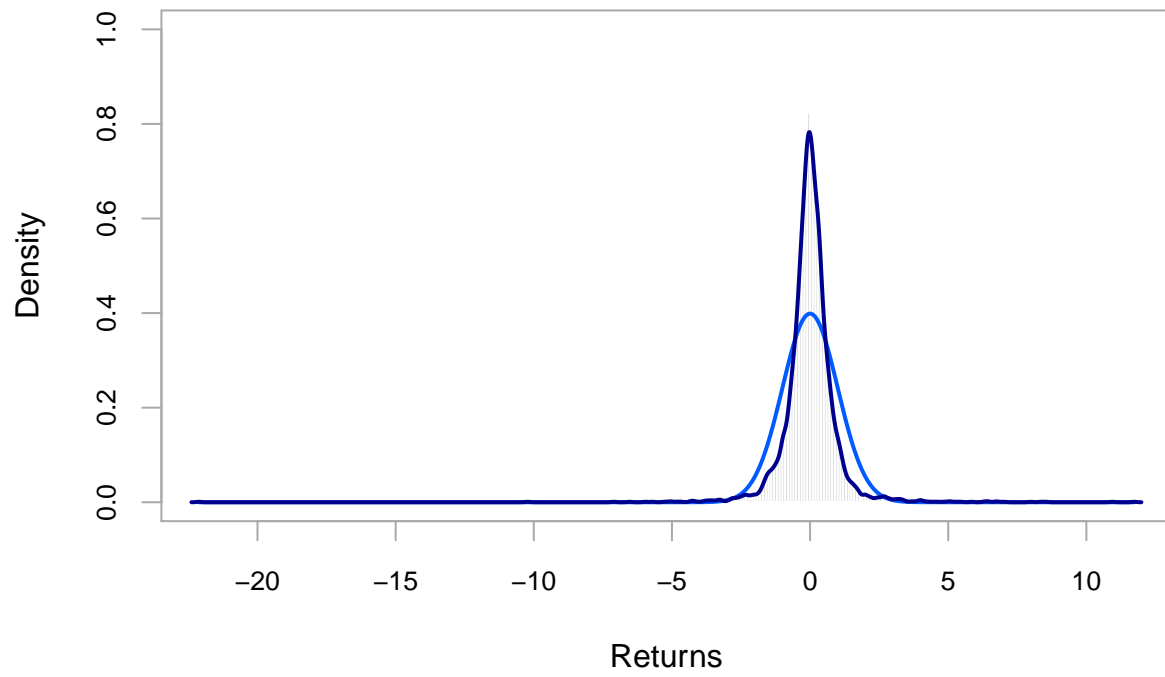
### Standardized Blackrock returns



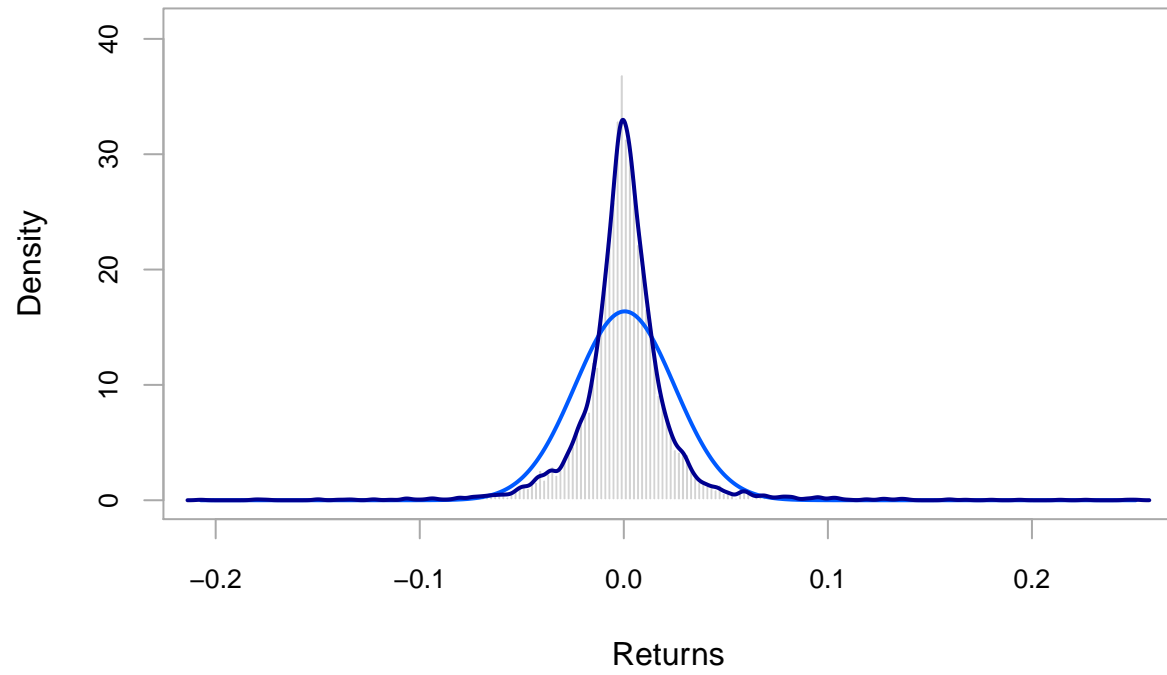
### Statestreet returns



### Standardized Statestreet returns

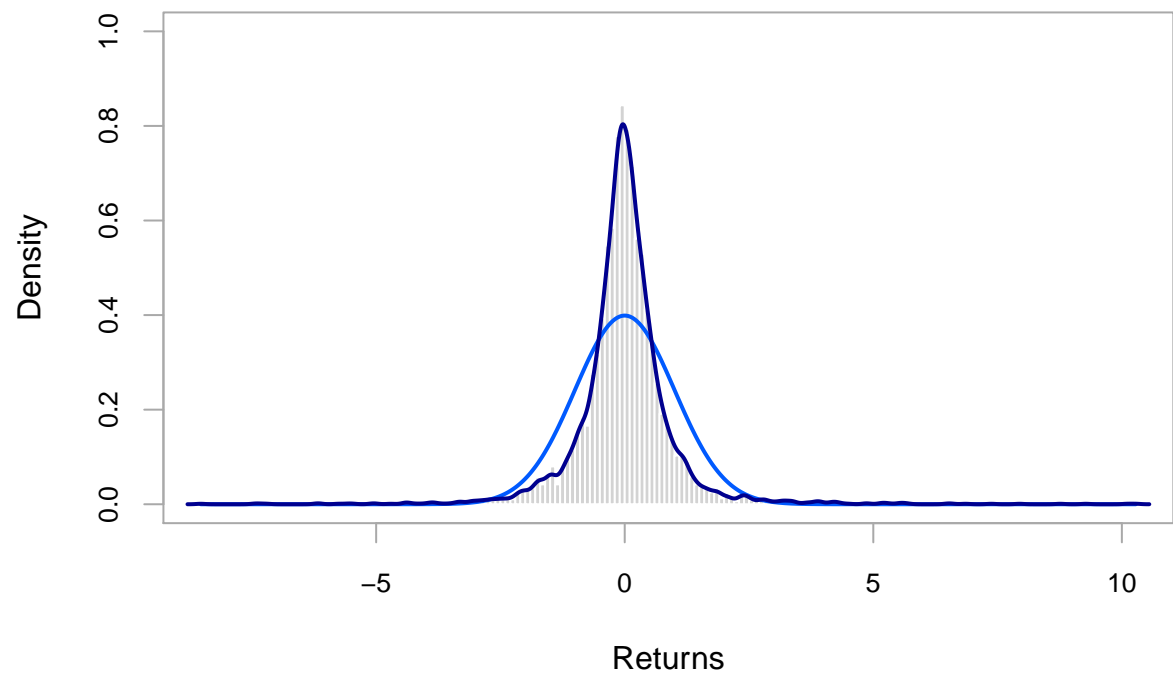


### JPmorgan returns

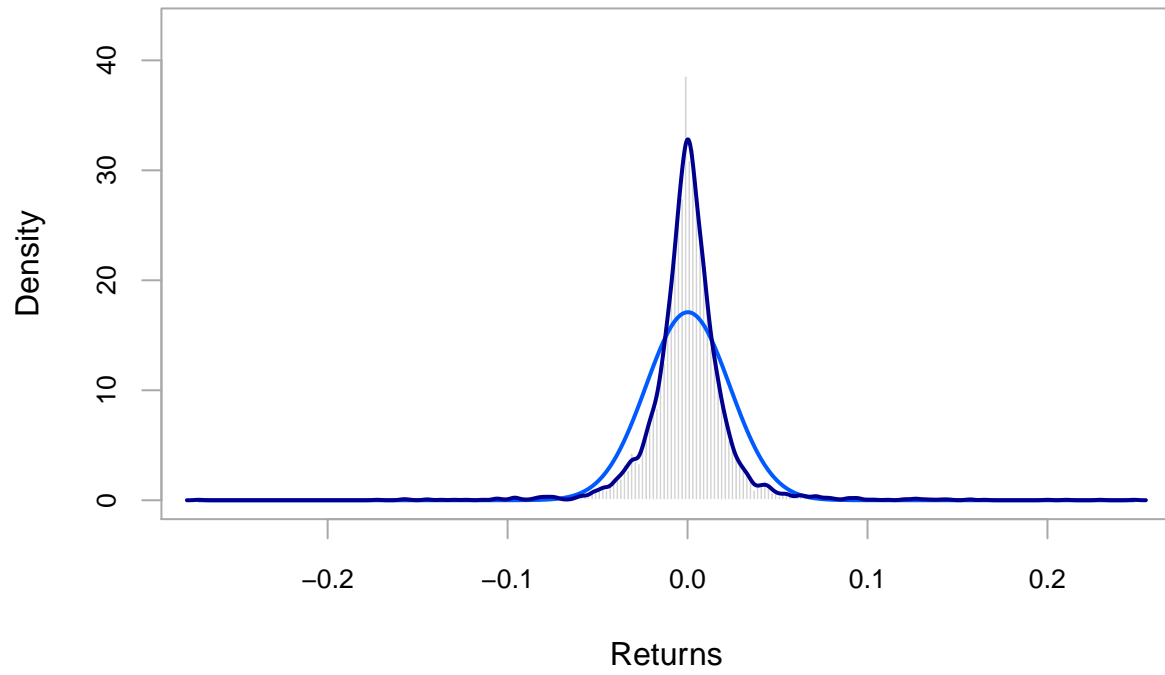




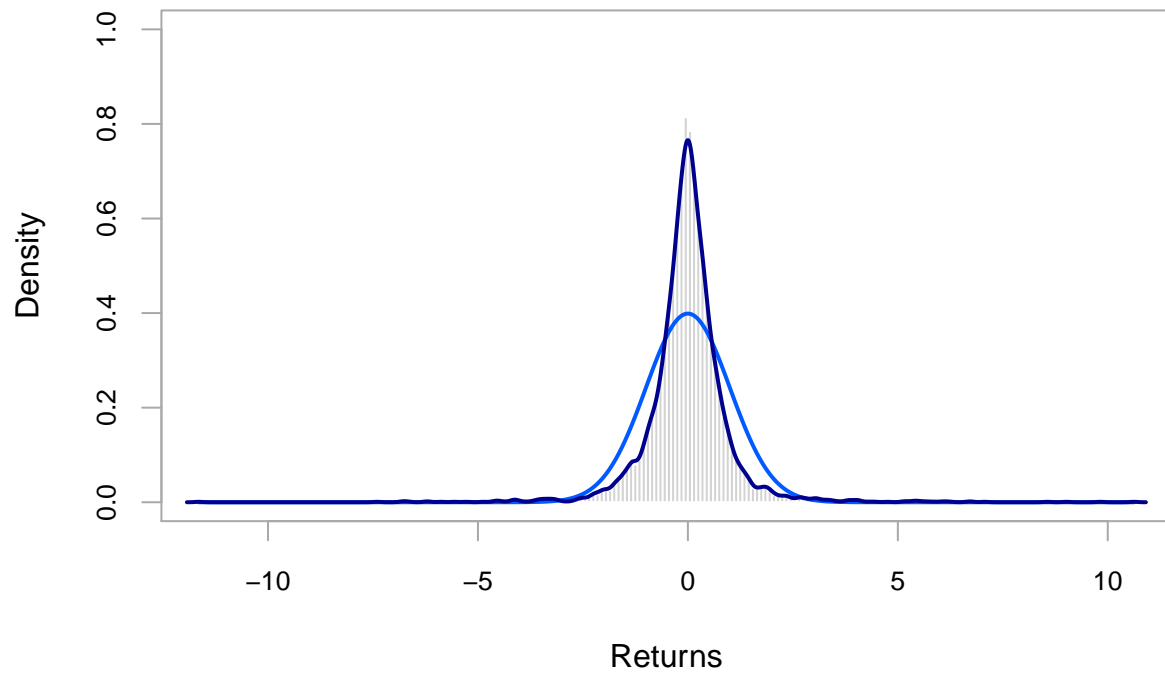
Standardized JPMorgan returns



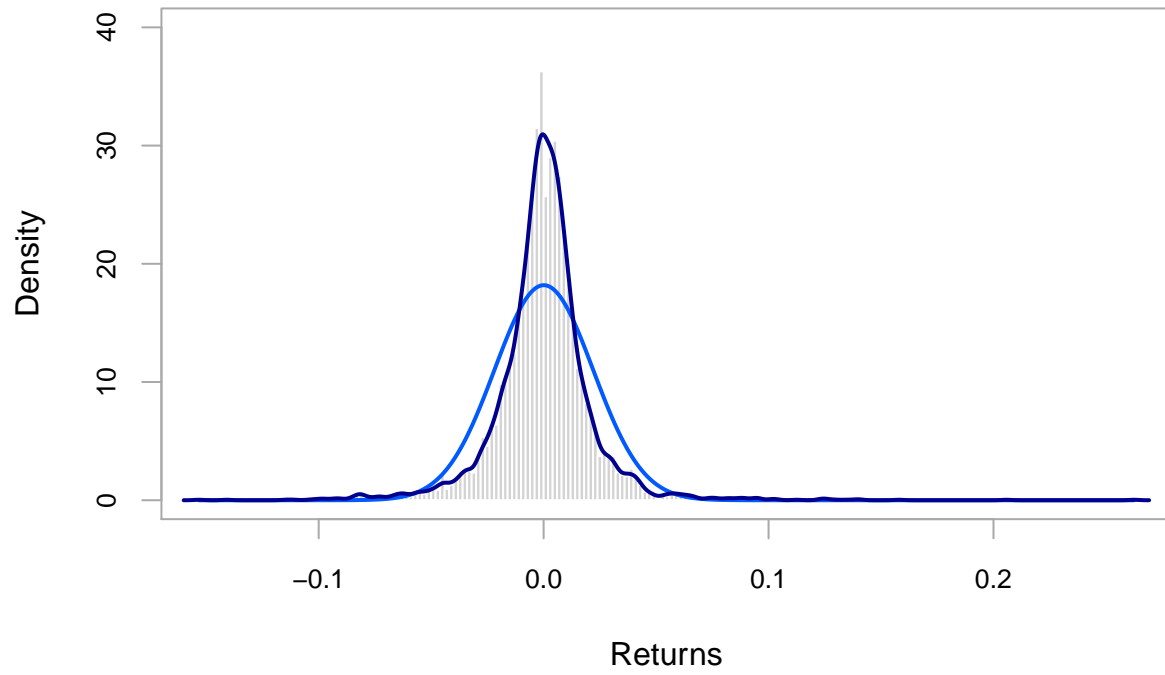
### Bankmellon returns



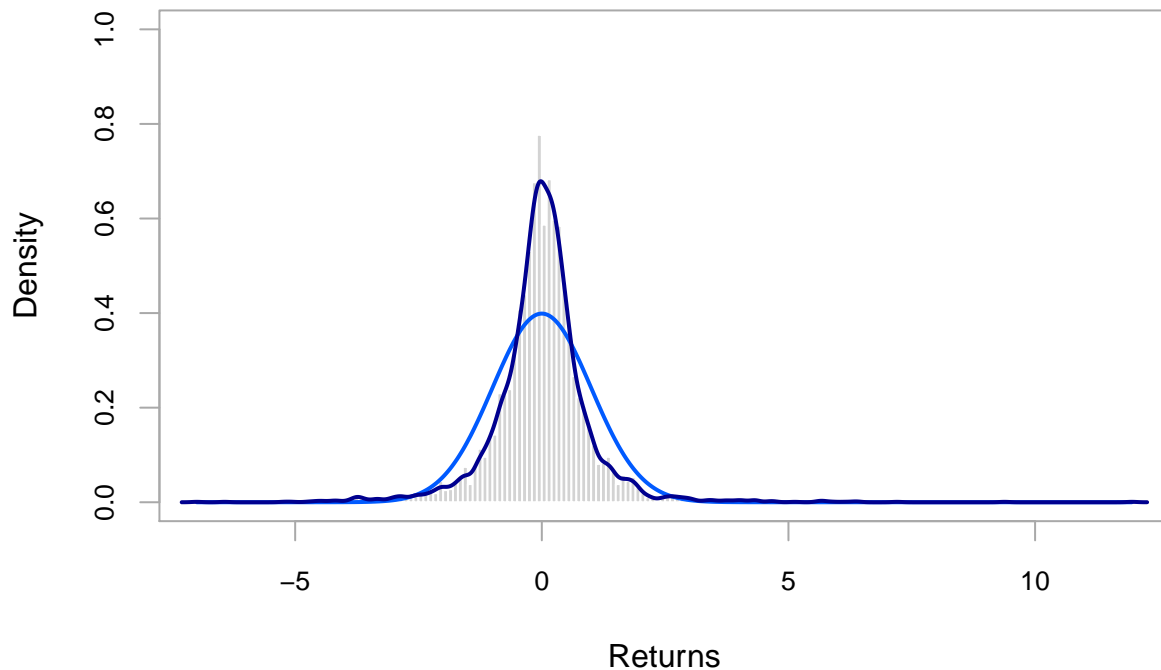
### Standardized Bankmellon returns



### Allianz returns



## Standardized Allianz returns



```
vanguard_tail_volatility <- tail(vol_df$Vanguard, 10)
```

```
plot_acf <- function(fund) {  
  f_mean <- mean(fund)  
  acf(abs(f_mean))  
}
```

*# In this chunk, prefix van\_ - is for Vanguard company*

```
fund <- funds$vanguard_return
```

*# Forming set of parameters for different GARCH model*

*# The most important is distribution.model - we are specifying type of distribution*

*# Standard GARCH with normal distribution of errors*

```
norm_garch_spec <- ugarchspec(mean.model = list(armaOrder = c(0,0)),  
                              variance.model = list(model = 'sGARCH'),  
                              distribution.model = 'norm')
```

*# GJR GARCH with normal distribution of errors*

```
norm_gjr_spec <- ugarchspec(mean.model = list(armaOrder = c(0,0)),  
                             variance.model = list(model = 'gjrGARCH'),  
                             distribution.model = 'norm')
```

*# Standard GARCH with skewed Student t distribution of errors*

```
sstd_garch_spec <- ugarchspec(mean.model = list(armaOrder = c(0,0)),  
                               variance.model = list(model = 'sGARCH'),  
                               distribution.model = 'sstd')
```

```

# GJR GARCH with skewed Student t distribution of errors
sstd_gjr_spec <- ugarchspec(mean.model = list(armaOrder = c(0,0)),
                             variance.model = list(model = 'gjrGARCH'),
                             distribution.model = 'sstd')

# Apply GARCH model to our data
norm_garch_fit <- ugarchfit(data = funds_red$vanguard_return,
                             spec = norm_garch_spec)

norm_gjr_fit <- ugarchfit(data = funds_red$vanguard_return,
                             spec = norm_gjr_spec)

sstd_garch_fit <- ugarchfit(data = funds_red$vanguard_return,
                             spec = sstd_garch_spec)

sstd_gjr_fit <- ugarchfit(data = funds_red$vanguard_return,
                             spec = sstd_gjr_spec)

#Coefficients
cat('\nCoefficients of', 'Standard GARCH with normal distribution of errors','\n')

##
## Coefficients of Standard GARCH with normal distribution of errors
print(round(norm_garch_fit@fit$matcoef,10))

##          Estimate   Std. Error   t value   Pr(>|t|)
## mu      0.0006660551 0.0001134153  5.872709 0.0000000043
## omega   0.0000024796 0.0000007827  3.167999 0.0015349183
## alpha1  0.1236562746 0.0098739025 12.523546 0.0000000000
## beta1   0.8564012304 0.0106465927 80.438996 0.0000000000

cat('\nRobust coefficients of', 'Standard GARCH with normal distribution of errors','\n')

##
## Robust coefficients of Standard GARCH with normal distribution of errors
print(round(norm_garch_fit@fit$robust.matcoef,10))

##          Estimate   Std. Error   t value   Pr(>|t|)
## mu      0.0006660551 0.0000945864  7.0417614 0.0000000000
## omega   0.0000024796 0.0000043291  0.5727754 0.5667967733
## alpha1  0.1236562746 0.0242583805  5.0974662 0.0000003442
## beta1   0.8564012304 0.0397101847 21.5662868 0.0000000000

cat('\nCoefficients of', 'GJR GARCH with normal distribution of errors','\n')

##
## Coefficients of GJR GARCH with normal distribution of errors
print(round(norm_gjr_fit@fit$matcoef,10))

##          Estimate   Std. Error   t value   Pr(>|t|)
## mu      0.0002599302 0.0000981960  2.647054e+00 0.008119637
## omega   0.0000024736 0.0000002385  1.037265e+01 0.0000000000
## alpha1  0.0000001769 0.0033173146  5.333410e-05 0.999957445
## beta1   0.8729915992 0.0066767202  1.307516e+02 0.0000000000
## gamma1  0.2067021938 0.0129895853  1.591292e+01 0.0000000000

```

```

cat('\nRobust coefficients of', 'Standard GARCH with normal distribution of errors','\n')

##
## Robust coefficients of Standard GARCH with normal distribution of errors
print(round(norm_gjr_fit@fit$robust.matcoef,10))

##          Estimate   Std. Error    t value    Pr(>|t|)
## mu      0.0002599302 0.0001309420 1.985079e+00 0.0471356725
## omega   0.0000024736 0.0000005885 4.203159e+00 0.0000263216
## alpha1  0.0000001769 0.0140922669 1.255480e-05 0.9999899827
## beta1   0.8729915992 0.0072802743 1.199119e+02 0.0000000000
## gamma1  0.2067021938 0.0315575960 6.549998e+00 0.0000000001
cat('\nCoefficients of', 'Standard GARCH with skewed Student t distribution of errors','\n')

##
## Coefficients of Standard GARCH with skewed Student t distribution of errors
print(round(ssstd_garch_fit@fit$matcoef,10))

##          Estimate   Std. Error    t value    Pr(>|t|)
## mu      0.0006062493 0.0001117549 5.4248140 0.0000000580
## omega   0.0000014778 0.0000018608 0.7941576 0.4271037050
## alpha1  0.1209330654 0.0313502725 3.8574805 0.0001145618
## beta1   0.8717911859 0.0295432800 29.5089504 0.0000000000
## skew    0.8888211922 0.0185481764 47.9195998 0.0000000000
## shape   7.1084616197 1.1187769377 6.3537792 0.0000000002
cat('\nRobust coefficients of', 'Standard GARCH with normal distribution of errors','\n')

##
## Robust coefficients of Standard GARCH with normal distribution of errors
print(round(ssstd_garch_fit@fit$robust.matcoef,10))

##          Estimate   Std. Error    t value    Pr(>|t|)
## mu      0.0006062493 0.0001642235 3.6916119 0.0002228373
## omega   0.0000014778 0.0000155483 0.0950453 0.9242788484
## alpha1  0.1209330654 0.2496181390 0.4844723 0.6280507427
## beta1   0.8717911859 0.2373673347 3.6727513 0.0002399530
## skew    0.8888211922 0.0575212007 15.4520626 0.0000000000
## shape   7.1084616197 6.9887168178 1.0171340 0.3090896792
cat('\nCoefficients of', 'GJR GARCH with skewed Student t distribution of errors','\n')

##
## Coefficients of GJR GARCH with skewed Student t distribution of errors
print(round(ssstd_gjr_fit@fit$matcoef,10))

##          Estimate   Std. Error    t value    Pr(>|t|)
## mu      0.0002495716 0.0001175544 2.123031e+00 0.03375129
## omega   0.0000020107 0.0000010712 1.876978e+00 0.06052114
## alpha1  0.0000000450 0.0127386126 3.535900e-06 0.99999718
## beta1   0.8732336384 0.0109951208 7.942010e+01 0.00000000
## gamma1  0.2246589215 0.0309933212 7.248624e+00 0.00000000
## skew    0.8546648715 0.0168886550 5.060586e+01 0.00000000
## shape   8.2593642262 0.9887244708 8.353555e+00 0.00000000

```

```
cat('\nRobust coefficients of', 'GJR GARCH with normal distribution of errors', '\n')
```

```
##
```

```
## Robust coefficients of GJR GARCH with normal distribution of errors
```

```
print(round(sstd_gjr_fit@fit$robust.matcoef,10))
```

```
##           Estimate   Std. Error    t value   Pr(>|t|)
## mu      0.0002495716 0.0002846399 8.767975e-01 0.38059664
## omega   0.0000020107 0.0000056600 3.552445e-01 0.72240643
## alpha1  0.0000000450 0.0445173275 1.011800e-06 0.99999919
## beta1   0.8732336384 0.0338930298 2.576440e+01 0.00000000
## gamma1  0.2246589215 0.1280897966 1.753917e+00 0.07944466
## skew    0.8546648715 0.0164777668 5.186776e+01 0.00000000
## shape   8.2593642262 1.0175550758 8.116872e+00 0.00000000
```

```
# Visualizing impact of negative previous return on variance
```

```
norm_garch_news <- as.data.frame(newsimpact(norm_garch_fit)[1:2])
```

```
norm_gjr_news <- as.data.frame(newsimpact(norm_gjr_fit) [1:2])
```

```
sstd_garch_news <- as.data.frame(newsimpact(sstd_garch_fit)[1:2])
```

```
sstd_gjr_news <- as.data.frame(newsimpact(sstd_gjr_fit)[1:2])
```

```
ggplot() +
```

```
  geom_point(data = norm_garch_news,
             aes(x = zx, y = zy, color = 'Normal GARCH')) +
```

```
  geom_line(data = norm_gjr_news,
            aes(x = zx, y = zy, color = 'Normal GJR')) +
```

```
  geom_line(data = sstd_garch_news,
            aes(x = zx, y = zy, color = 'Skewed t GARCH')) +
```

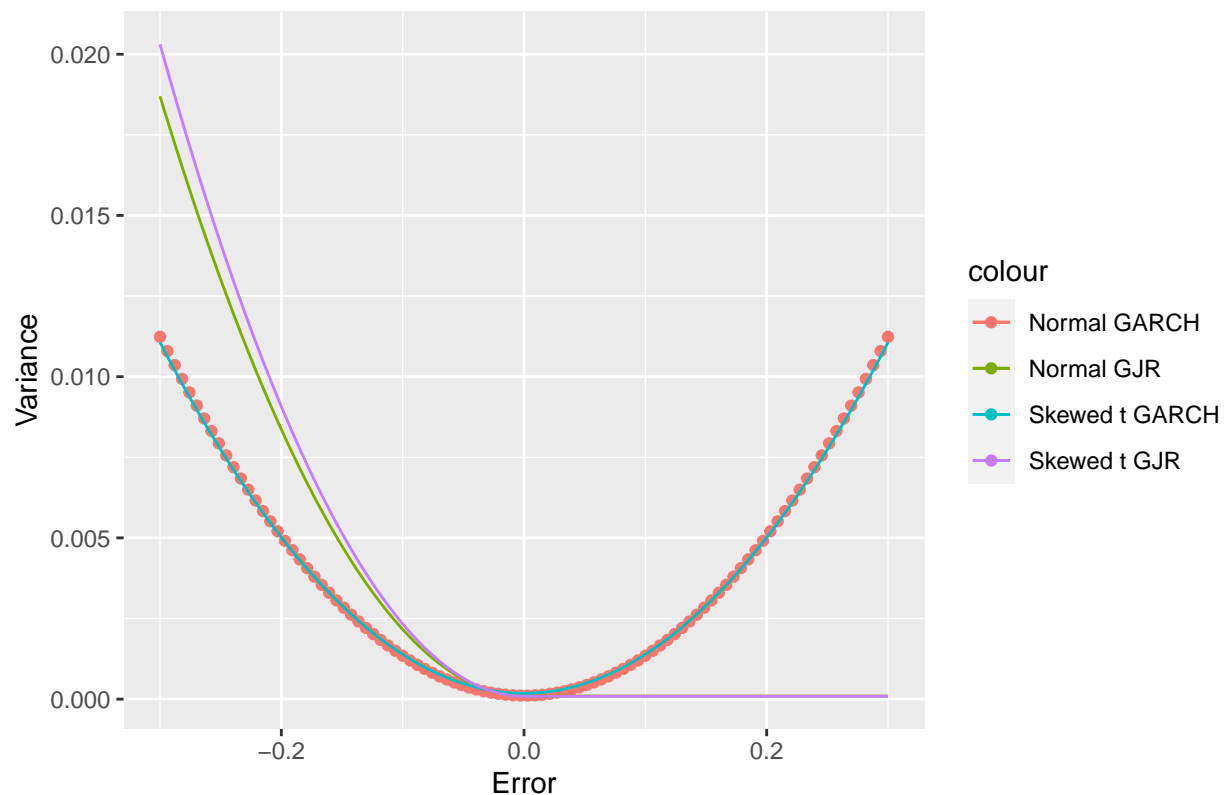
```
  geom_line(data = sstd_gjr_news,
            aes(x = zx, y = zy, color = 'Skewed t GJR')) +
```

```
  labs(x = 'Error', y = 'Variance',
        title = 'Dependence of variance on errors in different models') +
```

```
  theme(plot.title = element_text(hjust = 0.5))
```



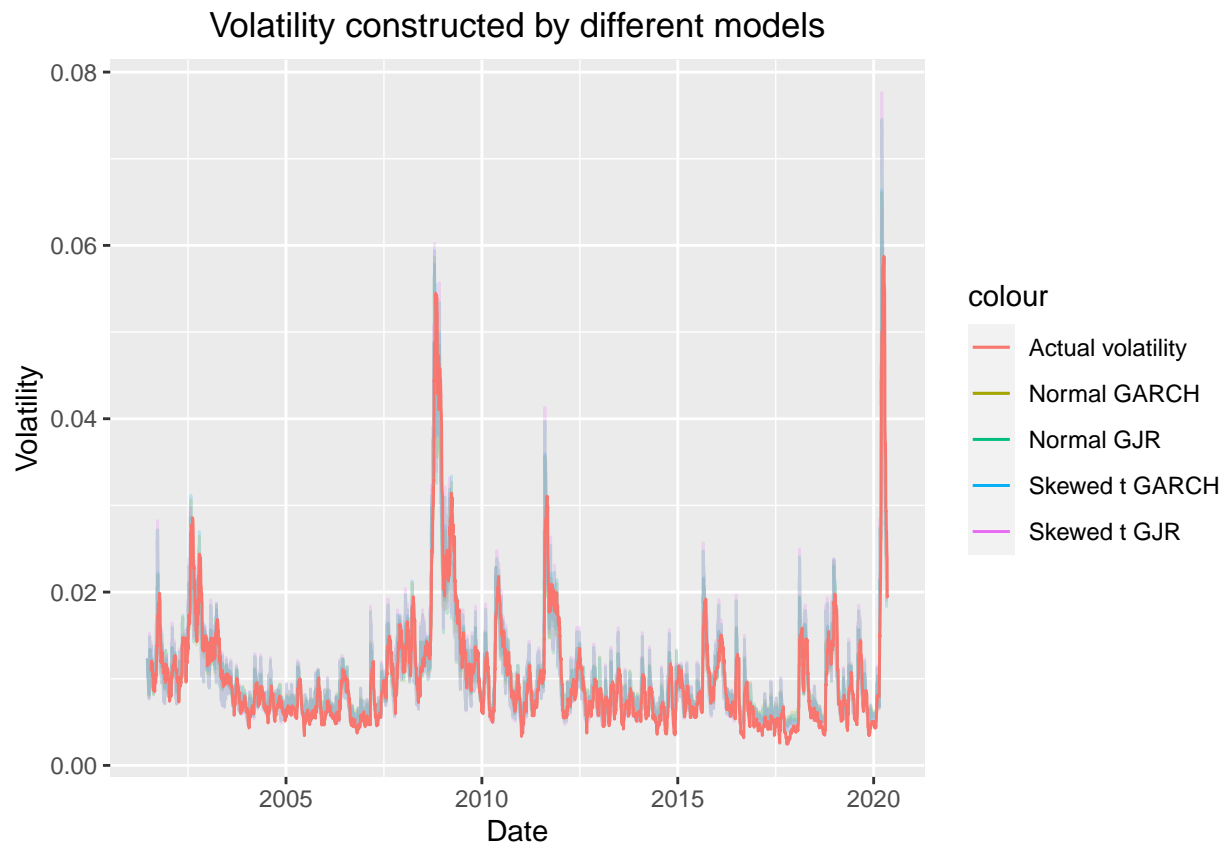
Dependence of variance on errors in different models



```
# Visualizing volatility
norm_garch_vol <- sigma(norm_garch_fit)
norm_gjr_vol <- sigma(norm_gjr_fit)
sstd_garch_vol <- sigma(sstd_garch_fit)
sstd_gjr_vol <- sigma(sstd_gjr_fit)

p <- ggplot() +
  geom_line(data = norm_garch_vol, aes(x = index(norm_garch_vol[,1]),
    y = norm_garch_vol[,1],
    color = 'Normal GARCH'), alpha = 0.2) +
  geom_line(data = norm_gjr_vol, aes(x = index(norm_gjr_vol[,1]),
    y = norm_gjr_vol[,1],
    color = 'Normal GJR'), alpha = 0.2) +
  geom_line(data = sstd_garch_vol, aes(x = index(sstd_garch_vol[,1]),
    y = sstd_garch_vol[,1],
    color = 'Skewed t GARCH'), alpha = 0.2) +
  geom_line(data = sstd_gjr_vol, aes(x = index(sstd_gjr_vol[,1]),
    y = sstd_gjr_vol[,1],
    color = 'Skewed t GJR'), alpha = 0.2) +
  geom_line(data = vol_df[,1], aes(y = vol_df[,1], x = index(vol_df[,1]),
    color = 'Actual volatility')) +
  labs(x = 'Date', y = 'Volatility',
    title = 'Volatility constructed by different models') +
  theme(plot.title = element_text(hjust = 0.5))

suppressMessages(suppressWarnings(print(p)))
```



```
rm(p)
```

```
# Predicting volatility for n.ahead periods
norm_garch_forecast <- ugarchforecast(fitORspec = norm_garch_fit,
                                     data = norm_garch_vol, n.ahead = 10)
norm_gjr_forecast <- ugarchforecast(fitORspec = norm_gjr_fit,
                                    data = norm_gjr_vol, n.ahead = 10)
sstd_garch_forecast <- ugarchforecast(fitORspec = sstd_garch_fit,
                                     data = sstd_garch_vol, n.ahead = 10)
sstd_gjr_forecast <- ugarchforecast(fitORspec = sstd_gjr_fit,
                                    data = sstd_gjr_vol, n.ahead = 10)
```

```
# 252 - number of working days in year
#van_norm_res <- sigma(van_forecast) ** sqrt(252)
```

```
# garch_for_funds <- function(funds) {
#   for (i in 1:ncol(funds)) {
#     return()
#   }
# }
```

```
# chart.Histogram(residuals(van_sstd_fit, standardize = T),
#               methods = c('add.normal', 'add.density', 'add.student'))
```

```
# cat('SST for norm\n')
# sum(vanguard_tail_volatility - van_norm_res)
#
```

```

# cat('\nSST for sstd\n')
# sum(vanguard_tail_volatility - van_sstd_res)
#
# cbind(vanguard_tail_volatility, van_norm_res, van_sstd_res)

# vanguard_std <- (funds$vanguard_return - mean(funds$vanguard_return))/sd(funds$vanguard_return)
#
# ggplot(data = data.frame(x = c(-10, 10)), aes(x)) +
#   stat_function(fun = dnorm, n = 4800, args = list(mean = 0, sd = 3)
#   , aes(x = x, colour = 'Normal')) + ylab("") +
#   scale_y_continuous(breaks = NULL) +
#   stat_function(fun = dskt, n = 4800, args = list(df = 100, gamma = 1.2),
#   aes(colour = 'Skewed Student t')) +
#   geom_histogram(data = vanguard_std, aes(x = vanguard_std, y = ..density..), bins = 100, alpha = 0.3)

```