

# Ravenpack news and sovereign debt trading

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```
### Launching all pair spread computations
### Trying to forecast the spread between ?? bonds futures
# library("SIT")
library("RPQuantUtils")
```

```
## Loading required package: waveslim
##
## waveslim: Wavelet Method for 1/2/3D Signals (version = 1.7.5)
##
## Loading required package: sendmailR
```

```
library("RPToolsDB")
```

```
## Loading required package: RPostgreSQL
## Loading required package: DBI
## Loading required package: lubridate
##
## Attaching package: 'lubridate'
##
## The following object is masked from 'package:waveslim':
##
##     pm
```

```
library("RPBackTesting")
```

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

## Warning: package 'xts' was built under R version 3.2.3

## Loading required package: zoo

## Warning: package 'zoo' was built under R version 3.2.3

##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##     as.Date, as.Date.numeric
##
##
## Attaching package: 'PerformanceAnalytics'
##
## The following object is masked from 'package:graphics':
##
##     legend
```

```
require(ggplot2)
```

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

```
require(scales)
```

```
## Loading required package: scales
```

```
require(grid)
```

```
## Loading required package: grid
```

```
require("ppcor")
```

```
## Loading required package: ppcor
```

```
require(graphics)  
require("TTR")
```

```
## Loading required package: TTR
```

```
require(plyr)
```

```
## Loading required package: plyr  
##  
## Attaching package: 'plyr'  
##  
## The following object is masked from 'package:lubridate':  
##  
##     here
```

```
# require(reshape)  
require(reshape2)
```

```
## Loading required package: reshape2
```

```
require(RColorBrewer)
```

```
## Loading required package: RColorBrewer
```

```
require(stats)  
require(Rsolnp)
```

```
## Loading required package: Rsolnp  
## Loading required package: truncnorm  
## Loading required package: parallel
```

```
require(zoo)
require(xts)
require(vars)
```

```
## Loading required package: vars
## Loading required package: MASS
## Loading required package: strucchange
## Loading required package: sandwich
## Loading required package: urca
## Loading required package: lmtest
```

```
# require(Quandl)
require(rpart)
```

```
## Loading required package: rpart
```

```
require(randomForest)
```

```
## Loading required package: randomForest
## randomForest 4.6-12
## Type rfNews() to see new features/changes/bug fixes.
```

```
# require(rpart.plot)
# require(rattle)
# install.packages(pkgs = "caret", dependencies = c("Depends", "Imports"))
# require(caret)
require(xgboost)
```

```
## Loading required package: xgboost
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
library(ROCR)
```

```
## Warning: package 'ROCR' was built under R version 3.2.3
```

```
## Loading required package: gplots
```

```
## Warning: package 'gplots' was built under R version 3.2.3
```

```
##
## Attaching package: 'gplots'
##
## The following object is masked from 'package:PerformanceAnalytics':
##
##     textplot
##
## The following object is masked from 'package:stats':
##
##     lowess
```

```

# source("E:/research/Projects/sduprey/NAR-271/RCode/RP_Plotting_Utils.R")
# source("E:/research/Projects/sduprey/NAR-271/RCode/RP_Macro_Monthly_Utils.R")
# source("E:/research/Projects/sduprey/NAR-271/RCode/RP_Spread_Utils.R")

user = 'sduprey'
# JIRA Code (e.g. NAR-#)
JIRACode = 'NAR-271'
repoPath = RP_GetSharedPath(user)
# Input Data Path
inputDataPath = paste(repoPath, 'InputData/', user, '/', JIRACode, '/', sep="")
# Output Data Path
outputDataPath = paste(repoPath, 'OutputData/', user, '/', JIRACode, '/', sep="")

# outputDataPathMonth <- paste(outputDataPath, "Month_2007/01_11_2015/", sep="")
# outputDataPathStrategyMonth <- paste(outputDataPath, "Month_2007/01_11_2015/", sep="")
outputDataPathMonth <- paste(outputDataPath, "Month_2007/", sep="")
outputDataPathStrategyMonth <- paste(outputDataPath, "Month_2007/", sep="")

results <- readRDS(paste(outputDataPathStrategyMonth, "all_pairs_results_month_2007.rds", sep = ""))

```

## Individual pair US/Germany trading strategy

```

### Deutschland visualization
my_result_spread_name <- "US_DE"
my_pair <- c("US", "DE")
my_pair[1] <- "US"
my_pair[2] <- "DE"
# Long/short position returns distribution
long_US <- results$DE_FIRST_WEIGHT>=0
long_DE <- results$DE_SECOND_WEIGHT>=0
LONG_STRATEGY_RETURN_US <- results$DE_FIRST_WEIGHT[long_US]*results$DE_FIRST_BOND_NEXT_RETURN[long_US]
LONG_STRATEGY_RETURN_DE <- results$DE_SECOND_WEIGHT[long_DE]*results$DE_SECOND_BOND_NEXT_OPEN_RETURN[long_DE]

print("Mean return for long position both US and DE")

## [1] "Mean return for long position both US and DE"

mean((LONG_STRATEGY_RETURN_US+LONG_STRATEGY_RETURN_DE))

## Warning in LONG_STRATEGY_RETURN_US + LONG_STRATEGY_RETURN_DE: longer object
## length is not a multiple of shorter object length

## [1] 0.00683276

print("Mean return for long position for US only")

## [1] "Mean return for long position for US only"

```

```
mean(LONG_STRATEGY_RETURN_US)
```

```
## [1] 0.0008238065
```

```
print("Mean return for long position for DE only")
```

```
## [1] "Mean return for long position for DE only"
```

```
mean(LONG_STRATEGY_RETURN_DE)
```

```
## [1] 0.005327057
```

```
# Short position returns distribution
```

```
short_US <- results$DE_FIRST_WEIGHT<=0
```

```
short_DE <- results$DE_SECOND_WEIGHT<=0
```

```
SHORT_STRATEGY_RETURN_US <- results$DE_FIRST_WEIGHT[short_US]*results$DE_FIRST_BOND_NEXT_RETURN[short_US]
```

```
SHORT_STRATEGY_RETURN_DE <- results$DE_SECOND_WEIGHT[short_DE]*results$DE_SECOND_BOND_NEXT_OPEN_RETURN[short_DE]
```

```
print("Mean return for short position both US and DE")
```

```
## [1] "Mean return for short position both US and DE"
```

```
mean((SHORT_STRATEGY_RETURN_US+SHORT_STRATEGY_RETURN_DE))
```

```
## Warning in SHORT_STRATEGY_RETURN_US + SHORT_STRATEGY_RETURN_DE: longer
```

```
## object length is not a multiple of shorter object length
```

```
## [1] -0.001629671
```

```
print("Mean return for long position for US only")
```

```
## [1] "Mean return for long position for US only"
```

```
mean(SHORT_STRATEGY_RETURN_US)
```

```
## [1] -0.001982707
```

```
print("Mean return for long position for DE only")
```

```
## [1] "Mean return for long position for DE only"
```

```
mean(SHORT_STRATEGY_RETURN_DE)
```

```
## [1] 0.0005104123
```

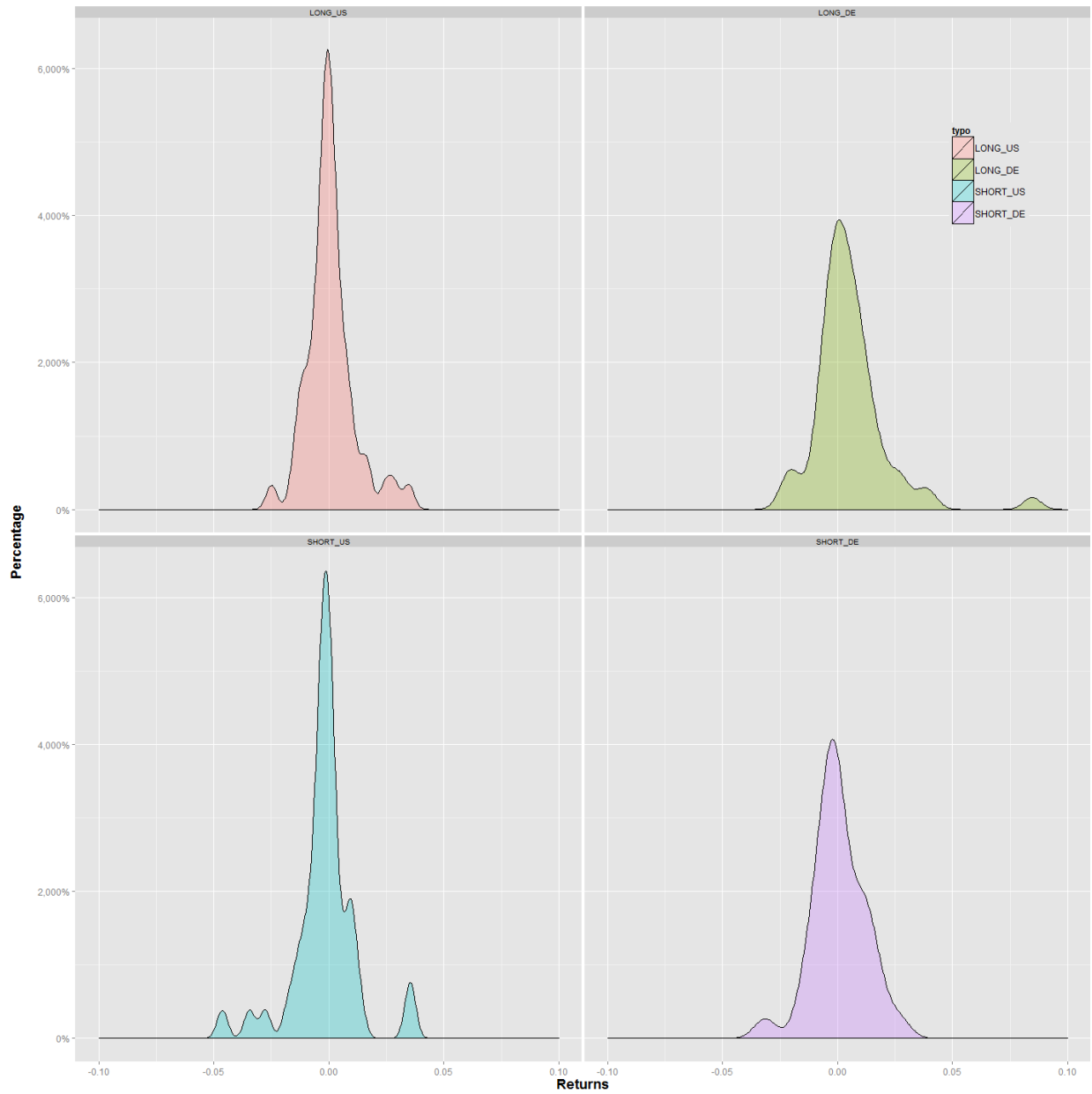
```

### Nice visualization of superposed densities
return_decomposition_df <- data.frame(returns=LONG_STRATEGY_RETURN_US,typo="LONG_US")
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=LONG_STRATEGY_RETURN_DE,typo="LONG_DE"))
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_US,typo="SHORT_US"))
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_DE,typo="SHORT_DE"))

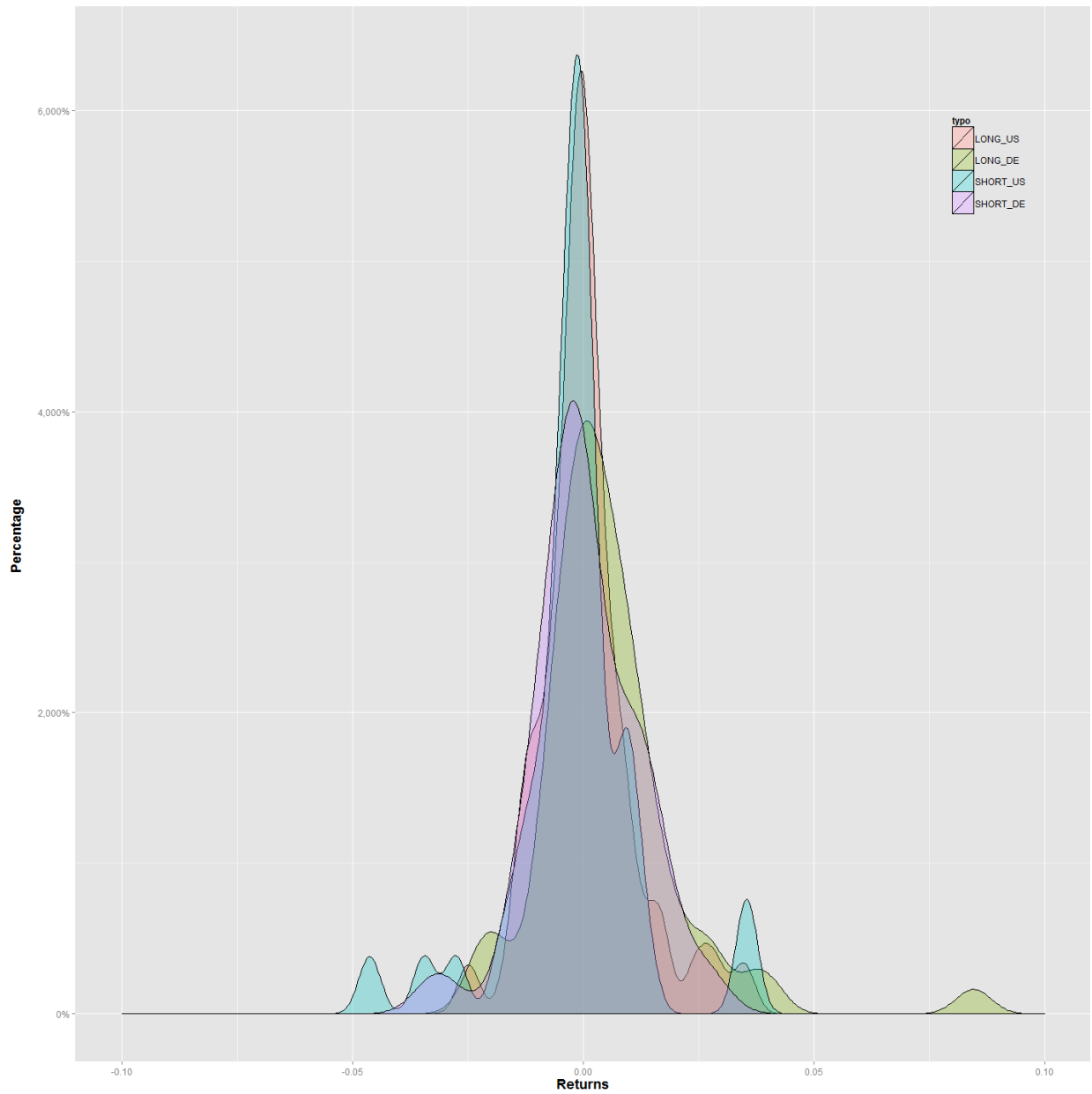
my_title <- "Distribution of returns per trade type"
my_xaxis_title <- paste("Returns")
my_yaxis_title <- paste("Percentage", "\n")

g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
  scale_y_continuous(labels = percent_format())+scale_x_continuous(limits=c(-0.1,0.1))+facet_wrap(~typo)
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))+
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
  theme(legend.background = element_rect(fill="gray90"))+
  theme(legend.key.size = unit(1., "cm"))+
  theme(legend.text = element_text(size=12,colour="black"))+
  theme(legend.title = element_text(size=12,colour="black"))
print(g)

```



```
g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
  scale_y_continuous(labels = percent_format())+scale_x_continuous(limits=c(-0.1,0.1))
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))+
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
  theme(legend.background = element_rect(fill="gray90"))+
  theme(legend.key.size = unit(1., "cm"))+
  theme(legend.text = element_text(size=12,colour="black"))+
  theme(legend.title = element_text(size=12,colour="black"))
print(g)
```



```
print("Outputing statistical results for our individual trading strategy US_DE")
```

```
## [1] "Outputing statistical results for our individual trading strategy US_DE"
```

```
WeightMatrix <- results[,c("DATES", "DE_FIRST_WEIGHT", "DE_SECOND_WEIGHT")]
colnames(WeightMatrix) <- c("DATE", "DE_FIRST_WEIGHT", "DE_SECOND_WEIGHT")
ReturnSerie <- results[,c("DE_STRATEGY_RETURN")]
turnover <- RP_GetTurnOver(WeightMatrix, -1)
RP_ReturnStats(ReturnSerie, 21, TRUE, WeightMatrix, 0.05, F)
```

```
## Warning in if (is.na(WeightMatrix)) {: the condition has length > 1 and
## only the first element will be used
```



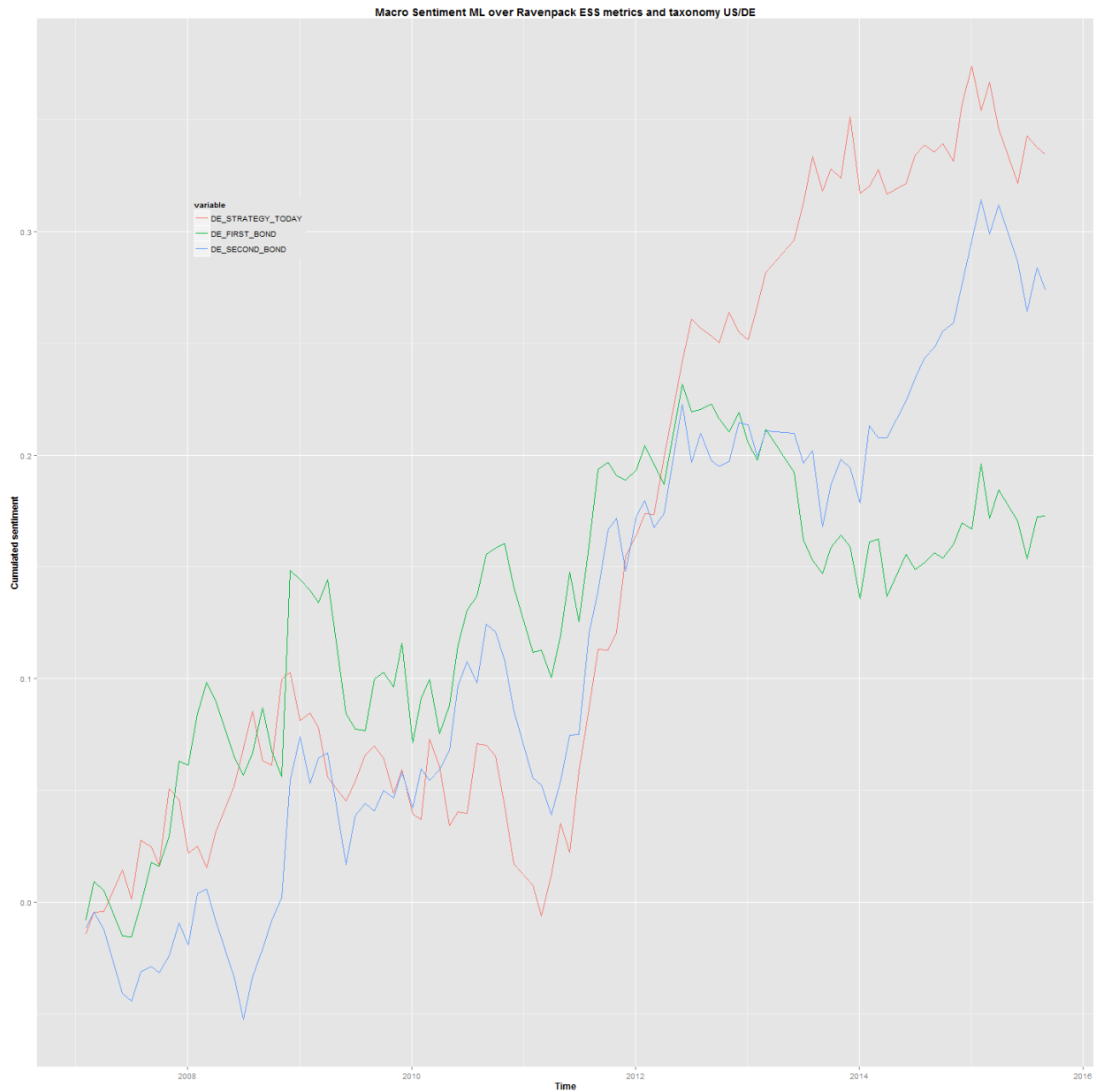
```

## *****
## ***** STRATEGY STATISTICS *****
## *****
## ***** NO FEE *****
## *****
## Annualized Return: 0.0376075073506816
## Annualized Volatility: 0.0582558020743051
## Information Ratio: 0.645558142049324
## Hit Ratio: 0.526315789473684
## W/L Ratio: 1.42449866331473
## Turnover: 0.563689662853424
## P-Value: 0.0724995334285696
## Max Drawdown: -0.0714420110363596
## Drawdown Recovery: 5
## Break-Even Fee (bps): 27.8422253966276
## *****
## *****
## ***** WITH FEE *****
## *****
## Annualized Return: 0.0308606737829419
## Annualized Volatility: 0.0583045697815325
## Information Ratio: 0.529301114793865
## Hit Ratio: 0.515789473684211
## W/L Ratio: 1.36626893887106
## Turnover: 0.563689662853424
## P-Value: 0.139763373803876
## Max Drawdown: -0.0855281748386419
## Drawdown Recovery: 6
## *****
## *****
## *****
## $STATS
## Annualized_Return Annualized_Return_Fee Annualized_Volatility
## 1 0.03760751 0.03086067 0.0582558
## Annualized_Volatility_Fee IR IR_Fee TurnOver HitRatio
## 1 0.05830457 0.6455581 0.5293011 0.5636897 0.5263158
## HitRatio_Fee WLRatio WLRatio_Fee PVal PVal_Fee BreakEvenFee
## 1 0.5157895 1.424499 1.366269 0.07249953 0.1397634 27.84223
## maxDrawdown drawdownRecovery maxDrawdown_Fee drawdownRecovery_Fee
## 1 -0.07144201 5 -0.08552817 6
##
## $RETURN
## [1] -0.0147415415 0.0098944084 0.0003817756 -0.0003242333 -0.0130223102
## [6] 0.0262203124 -0.0030818799 -0.0084807014 0.0337137763 -0.0046859600
## [11] -0.0243604897 0.0031168435 -0.0095099736 0.0158170757 0.0222803716
## [16] 0.0163894344 0.0165189266 -0.0222926028 -0.0019988156 0.0376778689
## [21] 0.0030905772 -0.0219164646 0.0033281989 -0.0064004883 -0.0225024727
## [26] 0.0211102290 0.0088306265 0.0114855314 0.0042702137 -0.0055053949
## [31] -0.0157740481 0.0103657675 -0.0199413544 -0.0025951627 0.0354449295
## [36] -0.0122336555 -0.0270514720 0.0062234699 -0.0005128009 0.0305983735
## [41] -0.0007745535 -0.0047841905 -0.0226219139 -0.0262728080 0.0002038848
## [46] -0.0135818102 0.0181240082 0.0227395695 -0.0129369214 0.0357414085
## [51] 0.0270815742 0.0266751692 -0.0005795302 0.0079775949 0.0337399251
## [56] 0.0091660653 0.0096653244 -0.0005134620 0.0252152657 0.0173045407

```

```
## [61] 0.0186240077 -0.0040935092 -0.0033848976 -0.0031378253 0.0135181131
## [66] -0.0089855775 -0.0032922076 0.0148665435 0.0150333073 -0.0216342855
## [71] 0.0162096172 0.0207475015 -0.0157453820 0.0100720922 -0.0041580464
## [76] 0.0270882779 -0.0349400644 0.0030648041 0.0074844023 -0.0109002445
## [81] 0.0148095810 0.0125378981 0.0043995127 -0.0031560042 0.0039022537
## [86] -0.0080657650 0.0250968331 0.0169794972 -0.0199903280 0.0126879284
## [91] -0.0211508766 -0.0212604740 0.0210863362 -0.0050156459 -0.0029673031
##
## $RETURN_FEE
## [1] -1.524910e-02 9.222475e-03 -5.556449e-04 -1.325058e-03 -1.403593e-02
## [6] 2.548946e-02 -3.461743e-03 -9.242577e-03 3.301796e-02 -5.000813e-03
## [11] -2.512928e-02 2.410691e-03 -1.033044e-02 1.483228e-02 2.203585e-02
## [16] 1.564949e-02 1.635035e-02 -2.305980e-02 -2.444886e-03 3.695534e-02
## [21] 2.563004e-03 -2.245740e-02 2.890592e-03 -7.407416e-03 -2.352575e-02
## [26] 2.037563e-02 8.334899e-03 1.095492e-02 3.950461e-03 -6.368165e-03
## [31] -1.672804e-02 9.740477e-03 -2.085908e-02 -2.929966e-03 3.502530e-02
## [36] -1.295434e-02 -2.778292e-02 5.505403e-03 -1.033228e-03 3.009389e-02
## [41] -1.525416e-03 -5.789491e-03 -2.287535e-02 -2.679227e-02 4.279504e-06
## [46] -1.366587e-02 1.782373e-02 2.224453e-02 -1.319449e-02 3.560568e-02
## [51] 2.684028e-02 2.627008e-02 -6.014799e-04 7.075735e-03 3.277263e-02
## [56] 8.528069e-03 8.922262e-03 -7.636217e-04 2.436420e-02 1.644672e-02
## [61] 1.787779e-02 -4.634254e-03 -3.514752e-03 -3.890466e-03 1.277791e-02
## [66] -9.155670e-03 -3.695341e-03 1.449518e-02 1.429422e-02 -2.240098e-02
## [71] 1.547140e-02 2.019470e-02 -1.582714e-02 9.081623e-03 -4.205080e-03
## [76] 2.635805e-02 -3.585794e-02 2.656363e-03 7.295407e-03 -1.097940e-02
## [81] 1.422962e-02 1.204401e-02 3.646346e-03 -4.159669e-03 3.070156e-03
## [86] -8.655680e-03 2.445685e-02 1.694798e-02 -2.047484e-02 1.267423e-02
## [91] -2.155020e-02 -2.129790e-02 2.095794e-02 -5.318370e-03 -3.805535e-03
```

```
toplot_df <- melt(results[,c("DATES", "DE_STRATEGY_TODAY", "DE_FIRST_BOND", "DE_SECOND_BOND")], "DATES")
my_title <- paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ", my_pair[1], "/", my_pair[2])
g <- ggplot(
  toplot_df, aes(
    x = DATES, y = value, group = variable, color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2, 0.8), legend.box = "vertical") +
  theme(legend.background = element_rect(fill = "gray90")) +
  theme(legend.key.size = unit(0.7, "cm"))
print(g)
```



```
# Individual pair US/UK trading strategy
```

```
### Great Britain visualization
my_result_spread_name <- "US_GB"
my_pair[1]<-"US"
my_pair[2]<-"GB"
# Long/short position returns distribution
long_US <- results$GB_FIRST_WEIGHT>=0
long_GB <- results$GB_SECOND_WEIGHT>=0
LONG_STRATEGY_RETURN_US <- results$GB_FIRST_WEIGHT[long_US]*results$GB_FIRST_BOND_NEXT_RETURN[long_US]
LONG_STRATEGY_RETURN_GB <- results$GB_SECOND_WEIGHT[long_GB]*results$GB_SECOND_BOND_NEXT_OPEN_RETURN[long_GB]

print("Mean return for long position both US and GB")
```

```
## [1] "Mean return for long position both US and GB"
```

```
mean((LONG_STRATEGY_RETURN_US+LONG_STRATEGY_RETURN_GB))
```

```
## Warning in LONG_STRATEGY_RETURN_US + LONG_STRATEGY_RETURN_GB: longer object  
## length is not a multiple of shorter object length
```

```
## [1] 0.01207859
```

```
print("Mean return for long position for US only")
```

```
## [1] "Mean return for long position for US only"
```

```
mean(LONG_STRATEGY_RETURN_US)
```

```
## [1] 0.003974042
```

```
print("Mean return for long position for GB only")
```

```
## [1] "Mean return for long position for GB only"
```

```
mean(LONG_STRATEGY_RETURN_GB)
```

```
## [1] 0.00557202
```

```
# Short position returns distribution
```

```
short_US <- results$GB_FIRST_WEIGHT<=0
```

```
short_GB <- results$GB_SECOND_WEIGHT<=0
```

```
SHORT_STRATEGY_RETURN_US <- results$GB_FIRST_WEIGHT[short_US]*results$GB_FIRST_BOND_NEXT_RETURN[short_US]
```

```
SHORT_STRATEGY_RETURN_GB <- results$GB_SECOND_WEIGHT[short_GB]*results$GB_SECOND_BOND_NEXT_OPEN_RETURN[short_GB]
```

```
print("Mean return for short position both US and GB")
```

```
## [1] "Mean return for short position both US and GB"
```

```
mean((SHORT_STRATEGY_RETURN_US+SHORT_STRATEGY_RETURN_GB))
```

```
## Warning in SHORT_STRATEGY_RETURN_US + SHORT_STRATEGY_RETURN_GB: longer  
## object length is not a multiple of shorter object length
```

```
## [1] -0.004035019
```

```
print("Mean return for long position for US only")
```

```
## [1] "Mean return for long position for US only"
```

```
mean(SHORT_STRATEGY_RETURN_US)
```

```
## [1] -0.001678582
```

```
print("Mean return for long position for GB only")
```

```
## [1] "Mean return for long position for GB only"
```

```
mean(SHORT_STRATEGY_RETURN_GB)
```

```
## [1] -0.0005464763
```

```
### Nice visualization of superposed densities
```

```
return_decomposition_df <- data.frame(returns=LONG_STRATEGY_RETURN_US,typo="LONG_US")
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=LONG_STRATEGY_RETURN_GB,typo="LONG_GB"))
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_US,typo="SHORT_US"))
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_GB,typo="SHORT_GB"))
```

```
my_title <- "Distribution of returns per trade type"
```

```
my_xaxis_title <- paste("Returns")
```

```
my_yaxis_title <- paste("Percentage", "\n")
```

```
g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
```

```
  scale_y_continuous(labels = percent_format())+scale_x_continuous(limits=c(-0.1,0.1))+facet_wrap(~typo)
```

```
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
```

```
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))
```

```
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
```

```
  theme(legend.background = element_rect(fill="gray90"))+
```

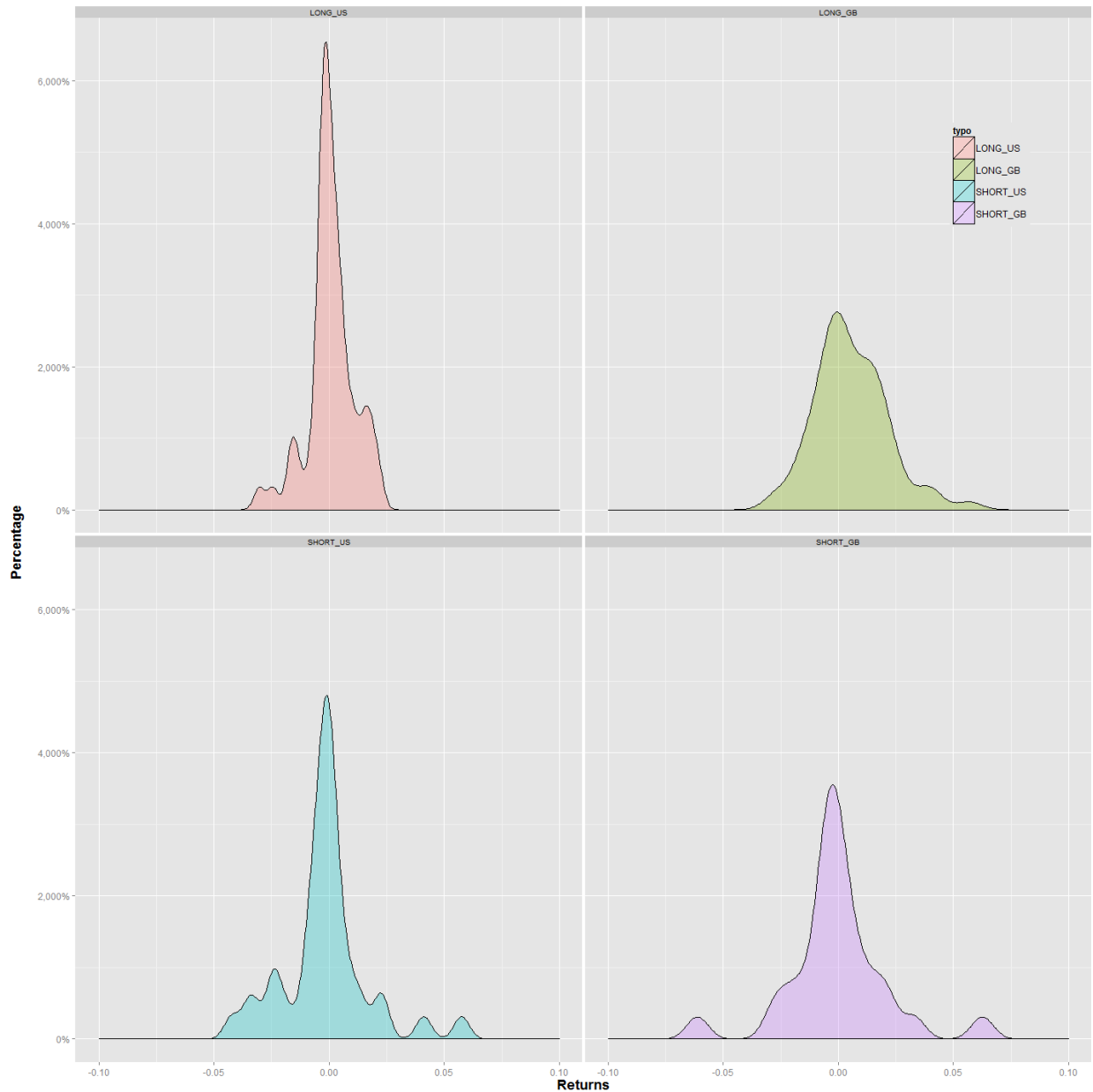
```
  theme(legend.key.size = unit(1., "cm"))+
```

```
  theme(legend.text = element_text(size=12,colour="black"))+
```

```
  theme(legend.title = element_text(size=12,colour="black"))
```

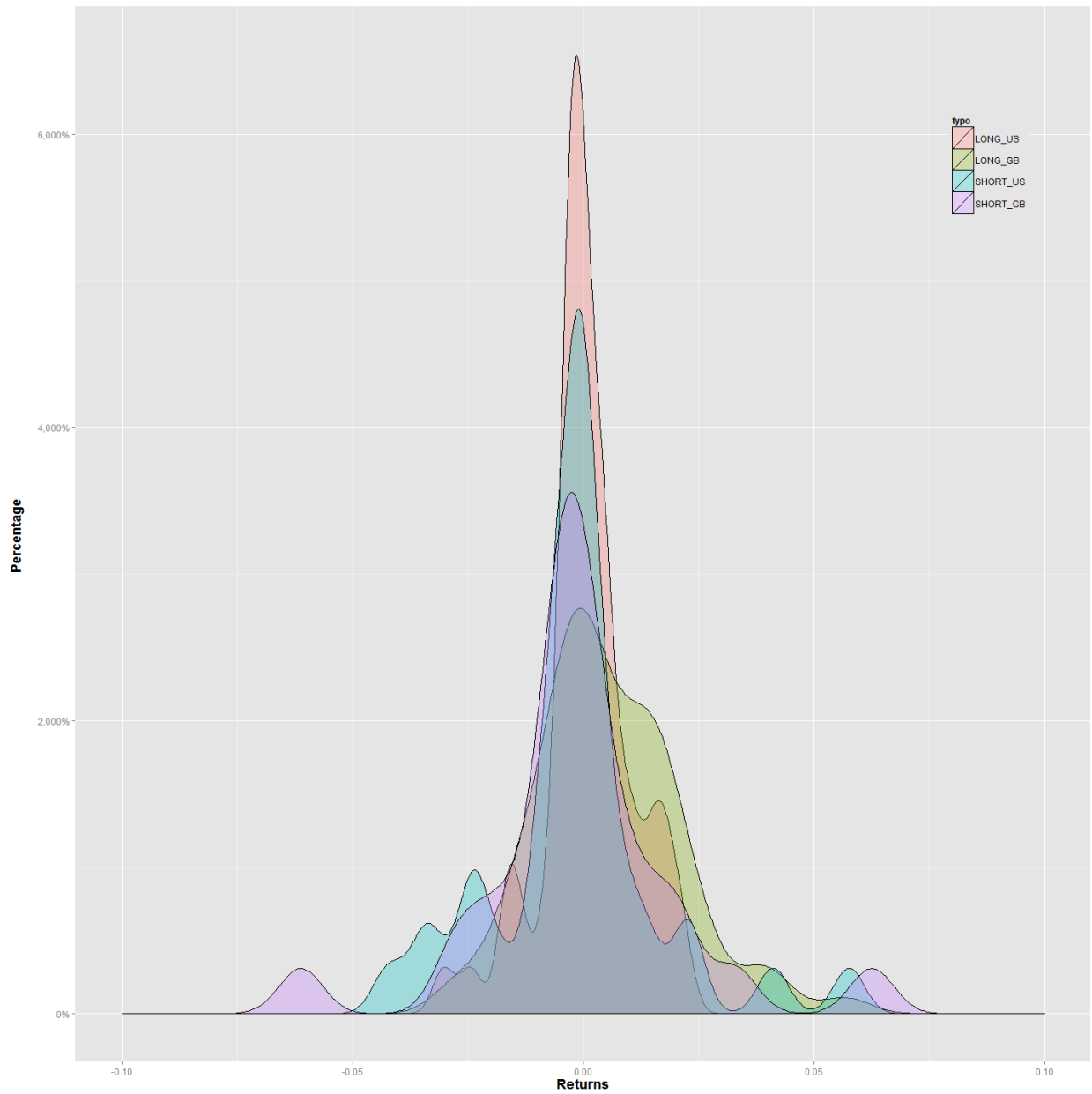
```
print(g)
```

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



```
g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
  scale_y_continuous(labels = percent_format()) +scale_x_continuous(limits=c(-0.1,0.1))
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))+
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
  theme(legend.background = element_rect(fill="gray90"))+
  theme(legend.key.size = unit(1., "cm"))+
  theme(legend.text = element_text(size=12,colour="black"))+
  theme(legend.title = element_text(size=12,colour="black"))
print(g)
```

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



```
print("Outputting statistical results for our individual trading strategy US_GB")
```

```
## [1] "Outputting statistical results for our individual trading strategy US_GB"
```

```
WeightMatrix <- results[,c("DATES", "GB_FIRST_WEIGHT", "GB_SECOND_WEIGHT")]
colnames(WeightMatrix) <- c("DATE", "GB_FIRST_WEIGHT", "GB_SECOND_WEIGHT")
ReturnSerie <- results[,c("GB_STRATEGY_RETURN")]
turnover <- RP_GetTurnOver(WeightMatrix, -1)
RP_ReturnStats(ReturnSerie, 21, TRUE, WeightMatrix, 0.05, F)
```

```
## Warning in if (is.na(WeightMatrix)) {: the condition has length > 1 and
## only the first element will be used
```

```

## *****
## ***** STRATEGY STATISTICS *****
## *****
## ***** NO FEE *****
## *****
## Annualized Return: 0.063850214335324
## Annualized Volatility: 0.0709319349945313
## Information Ratio: 0.900161744357415
## Hit Ratio: 0.536842105263158
## W/L Ratio: 1.77556898598927
## Turnover: 0.540070451839978
## P-Value: 0.0129739348046073
## Max Drawdown: -0.0943845111427219
## Drawdown Recovery: 3
## Break-Even Fee (bps): 49.3920010382621
## *****
## *****
## ***** WITH FEE *****
## *****
## Annualized Return: 0.0573975464822265
## Annualized Volatility: 0.0710178955504144
## Information Ratio: 0.808212437687356
## Hit Ratio: 0.536842105263158
## W/L Ratio: 1.64328334599237
## Turnover: 0.540070451839978
## P-Value: 0.025239075826137
## Max Drawdown: -0.0967948811177515
## Drawdown Recovery: 3
## *****
## *****
## *****

## $STATS
## Annualized_Return Annualized_Return_Fee Annualized_Volatility
## 1 0.06385021 0.05739755 0.07093193
## Annualized_Volatility_Fee IR IR_Fee TurnOver HitRatio
## 1 0.0710179 0.9001617 0.8082124 0.5400705 0.5368421
## HitRatio_Fee WLRatio WLRatio_Fee PVal PVal_Fee BreakEvenFee
## 1 0.5368421 1.775569 1.643283 0.01297393 0.02523908 49.392
## maxDrawdown drawdownRecovery maxDrawdown_Fee drawdownRecovery_Fee
## 1 -0.09438451 3 -0.09679488 3
##
## $RETURN
## [1] -0.0143943355 0.0140486505 -0.0078602108 0.0068235209 -0.0146493472
## [6] 0.0181961308 -0.0020566097 0.0049652503 0.0217288301 -0.0069755063
## [11] 0.0144598046 0.0094331819 0.0063669465 -0.0172247652 0.0035804137
## [16] 0.0206474712 0.0241576848 0.0142580215 0.0175903154 0.1168042874
## [21] 0.0236454296 -0.0168139194 0.0219754262 0.0227199156 0.0312180688
## [26] 0.0175546266 -0.0155132482 0.0322613904 -0.0063635265 -0.0124418308
## [31] -0.0247680184 -0.0308359256 0.0043018405 -0.0033860024 0.0223793605
## [36] -0.0101146376 0.0291615696 0.0130047373 0.0029314152 0.0242141433
## [41] -0.0032937563 -0.0024127039 -0.0208766737 0.0032598237 0.0107792409
## [46] -0.0065352324 0.0195947225 0.0256440544 -0.0063287967 0.0479876195
## [51] 0.0058750045 -0.0068577960 -0.0048651314 -0.0030068192 0.0173202756
## [56] 0.0064418489 -0.0079062236 -0.0041992614 0.0140346453 -0.0127770707

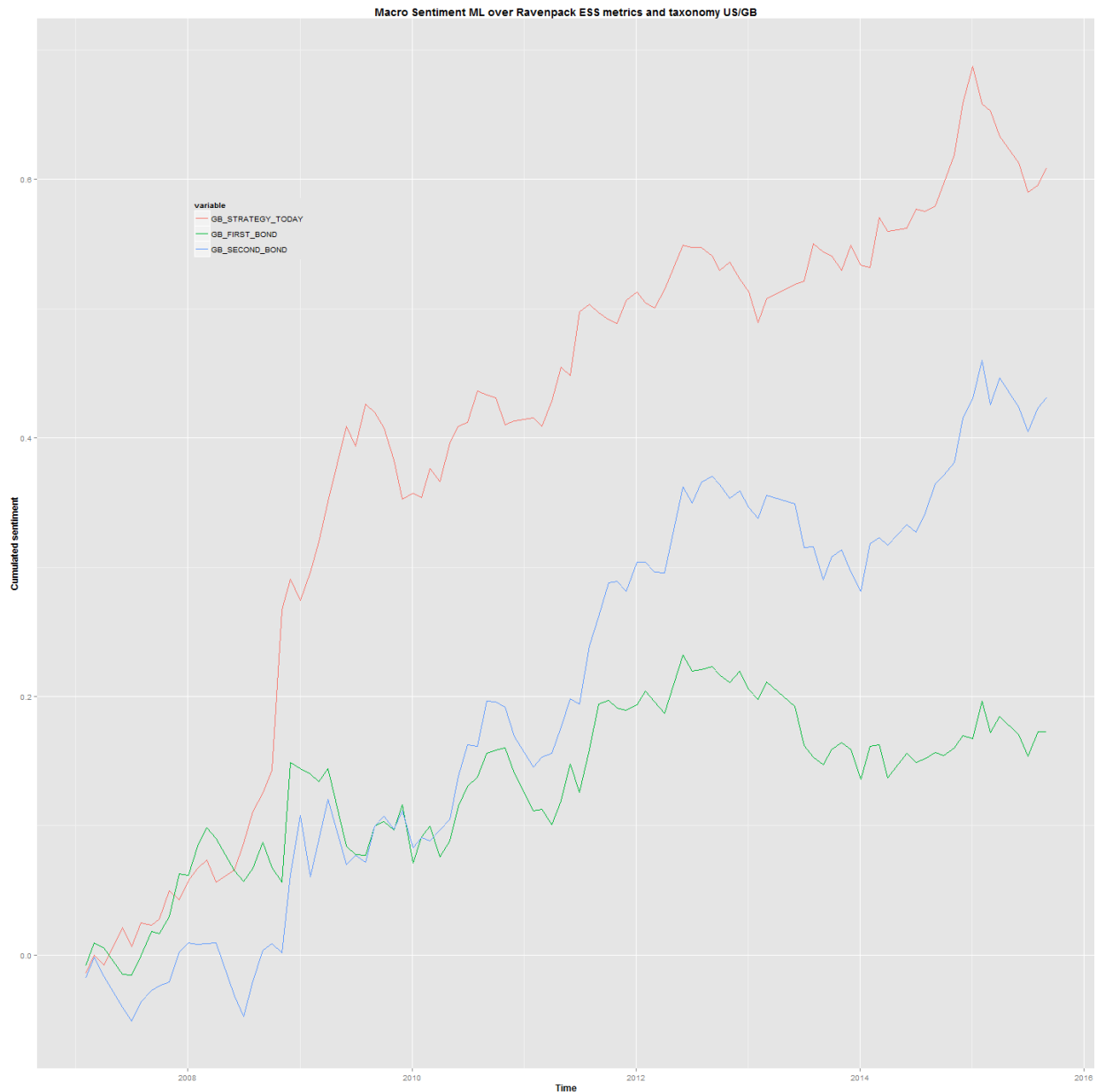
```



```
## [61] -0.0015599247 -0.0003023358 -0.0065132701 -0.0113118386 0.0065298237
## [66] -0.0131918533 -0.0099042831 -0.0241532435 0.0179294983 0.0317916442
## [71] 0.0021374171 0.0290605906 -0.0067346787 -0.0030739111 -0.0113466916
## [76] 0.0192728992 -0.0153452134 -0.0015127132 0.0376815186 -0.0104618692
## [81] 0.0057924926 0.0145710349 -0.0018693113 0.0041737541 0.0173144127
## [86] 0.0219031513 0.0390714082 0.0281924666 -0.0291640348 -0.0055289088
## [91] -0.0195406309 -0.0210449661 -0.0227324524 0.0053592332 0.0130833290
##
## $RETURN_FEE
## [1] -0.0149017134 0.0137885828 -0.0088039961 0.0059553190 -0.0152824698
## [6] 0.0173194766 -0.0023715331 0.0039697078 0.0216991030 -0.0079138733
## [11] 0.0134736743 0.0087725563 0.0053727992 -0.0181649588 0.0032482166
## [16] 0.0200911558 0.0240727768 0.0139175951 0.0172287837 0.1159141321
## [21] 0.0230138712 -0.0171316770 0.0213609131 0.0223282765 0.0305716839
## [26] 0.0171731065 -0.0155689769 0.0321279365 -0.0071734772 -0.0134548633
## [31] -0.0255015100 -0.0315738831 0.0033682305 -0.0043898978 0.0215992955
## [36] -0.0104514163 0.0285138535 0.0124654867 0.0022499328 0.0239813992
## [41] -0.0039628462 -0.0030812044 -0.0211987363 0.0024758516 0.0107218695
## [46] -0.0071624561 0.0194775402 0.0254906713 -0.0068783103 0.0478381948
## [51] 0.0056994484 -0.0075292757 -0.0057273056 -0.0033790771 0.0169939010
## [56] 0.0061777255 -0.0079322883 -0.0045756339 0.0140299262 -0.0135688104
## [61] -0.0017654158 -0.0005075686 -0.0070580633 -0.0117270942 0.0062839517
## [66] -0.0132962936 -0.0100368953 -0.0249748021 0.0169467853 0.0308224663
## [71] 0.0018047401 0.0287973028 -0.0077419432 -0.0030739111 -0.0120211932
## [76] 0.0183448572 -0.0163611929 -0.0017943420 0.0374190445 -0.0107372906
## [81] 0.0049830650 0.0135977268 -0.0028496701 0.0031774230 0.0167974741
## [86] 0.0215262147 0.0387008890 0.0276568989 -0.0291886732 -0.0062163718
## [91] -0.0202377976 -0.0212300100 -0.0237559688 0.0043640831 0.0129460148
```

```
toplot_df <- melt(results[,c("DATES", "GB_STRATEGY_TODAY", "GB_FIRST_BOND", "GB_SECOND_BOND")], "DATES")

my_title <- paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ", my_pair[1], "/", my_pair[2])
g <- ggplot(
  toplot_df, aes(
    x = DATES, y = value, group = variable, color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2, 0.8), legend.box = "vertical") +
  theme(legend.background = element_rect(fill = "gray90")) +
  theme(legend.key.size = unit(0.7, "cm"))
print(g)
```



# Individual pair US/Japan trading strategy

### Japan visualization

```
my_result_spread_name <- "US_JP"
```

```
my_pair[1]<-"US"
```

```
my_pair[2]<-"JP"
```

*# Long/short position returns distribution*

```
long_US <- results$JP_FIRST_WEIGHT>=0
```

```
long_JP <- results$JP_SECOND_WEIGHT>=0
```

```
LONG_STRATEGY_RETURN_US <- results$JP_FIRST_WEIGHT[long_US]*results$JP_FIRST_BOND_NEXT_RETURN[long_US]
```

```
LONG_STRATEGY_RETURN_JP <- results$JP_SECOND_WEIGHT[long_JP]*results$JP_SECOND_BOND_NEXT_OPEN_RETURN[long_JP]
```

```
print("Mean return for long position both US and JP")
```

```
## [1] "Mean return for long position both US and JP"
```

```
mean((LONG_STRATEGY_RETURN_US+LONG_STRATEGY_RETURN_JP))
```

```
## Warning in LONG_STRATEGY_RETURN_US + LONG_STRATEGY_RETURN_JP: longer object  
## length is not a multiple of shorter object length
```

```
## [1] 0.008098824
```

```
print("Mean return for long position for US only")
```

```
## [1] "Mean return for long position for US only"
```

```
mean(LONG_STRATEGY_RETURN_US)
```

```
## [1] 0.003310682
```

```
print("Mean return for long position for JP only")
```

```
## [1] "Mean return for long position for JP only"
```

```
mean(LONG_STRATEGY_RETURN_JP)
```

```
## [1] 0.004674318
```

```
# Short position returns distribution
```

```
short_US <- results$JP_FIRST_WEIGHT<=0
```

```
short_JP <- results$JP_SECOND_WEIGHT<=0
```

```
SHORT_STRATEGY_RETURN_US <- results$JP_FIRST_WEIGHT[short_US]*results$JP_FIRST_BOND_NEXT_RETURN[short_US]
```

```
SHORT_STRATEGY_RETURN_JP <- results$JP_SECOND_WEIGHT[short_JP]*results$JP_SECOND_BOND_NEXT_OPEN_RETURN[short_JP]
```

```
print("Mean return for short position both US and JP")
```

```
## [1] "Mean return for short position both US and JP"
```

```
mean((SHORT_STRATEGY_RETURN_US+SHORT_STRATEGY_RETURN_JP))
```

```
## Warning in SHORT_STRATEGY_RETURN_US + SHORT_STRATEGY_RETURN_JP: longer  
## object length is not a multiple of shorter object length
```

```
## [1] 0.01194968
```

```
print("Mean return for long position for US only")
```

```
## [1] "Mean return for long position for US only"
```

```
mean(SHORT_STRATEGY_RETURN_US)
```

```
## [1] 0.001151783
```

```
print("Mean return for long position for JP only")
```

```
## [1] "Mean return for long position for JP only"
```

```
mean(SHORT_STRATEGY_RETURN_JP)
```

```
## [1] 0.01142507
```

```
### Nice visualization of superposed densities
```

```
return_decomposition_df <- data.frame(returns=LONG_STRATEGY_RETURN_US,typo="LONG_US")
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=LONG_STRATEGY_RETURN_JP,typo="LONG_JP"))
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_US,typo="SHORT_US"))
```

```
return_decomposition_df <- rbind(return_decomposition_df, data.frame(returns=SHORT_STRATEGY_RETURN_JP,typo="SHORT_JP"))
```

```
my_title <- "Distribution of returns per trade type"
```

```
my_xaxis_title <- paste("Returns")
```

```
my_yaxis_title <- paste("Percentage", "\n")
```

```
g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
```

```
  scale_y_continuous(labels = percent_format())+scale_x_continuous(limits=c(-0.1,0.1))+facet_wrap(~typo)
```

```
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
```

```
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))
```

```
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
```

```
  theme(legend.background = element_rect(fill="gray90"))+
```

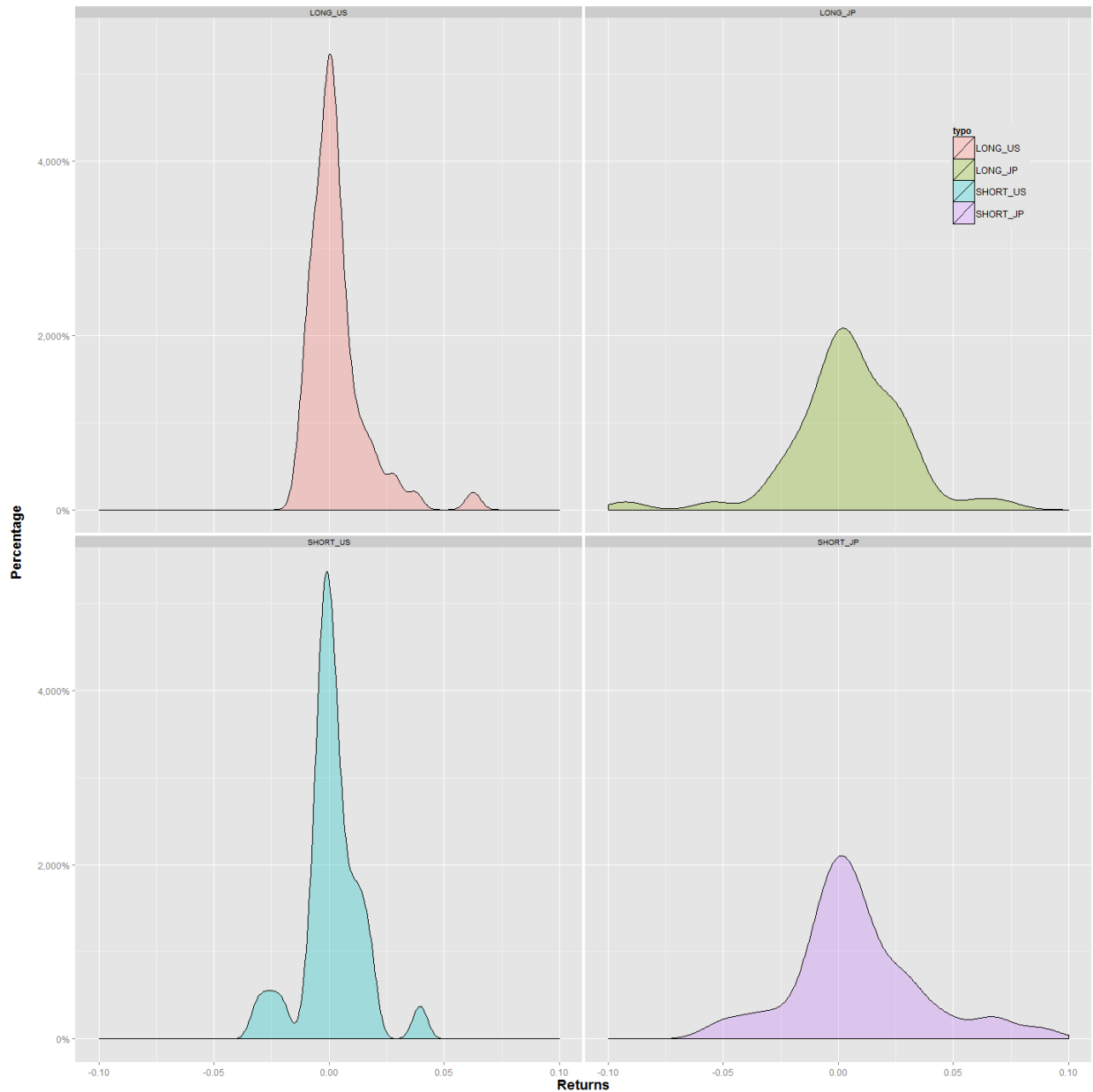
```
  theme(legend.key.size = unit(1., "cm"))+
```

```
  theme(legend.text = element_text(size=12,colour="black"))+
```

```
  theme(legend.title = element_text(size=12,colour="black"))
```

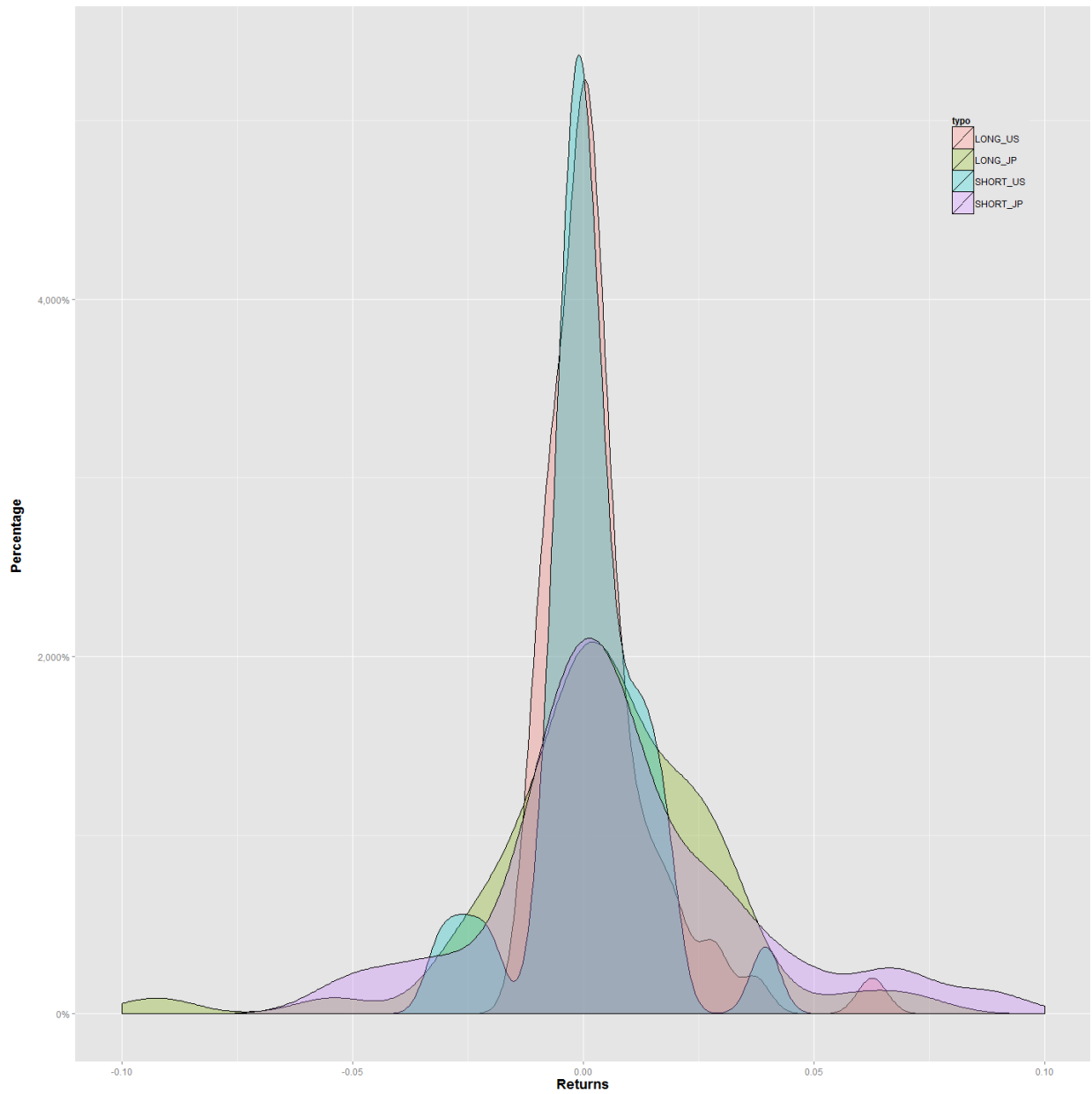
```
print(g)
```

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



```
g <- ggplot(return_decomposition_df, aes(x=returns, fill=typo)) +geom_density(alpha=.3)+
  scale_y_continuous(labels = percent_format()) +scale_x_continuous(limits=c(-0.1,0.1))
g <- g +ylab(my_yaxis_title)+xlab(my_xaxis_title)+
  theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))+
  theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
  theme(legend.background = element_rect(fill="gray90"))+
  theme(legend.key.size = unit(1., "cm"))+
  theme(legend.text = element_text(size=12,colour="black"))+
  theme(legend.title = element_text(size=12,colour="black"))
print(g)
```

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



```
print("Outputting statistical results for our individual trading strategy US_JP")
```

```
## [1] "Outputting statistical results for our individual trading strategy US_JP"
```

```
WeightMatrix <- results[,c("DATES", "JP_FIRST_WEIGHT", "JP_SECOND_WEIGHT")]
colnames(WeightMatrix) <- c("DATE", "JP_FIRST_WEIGHT", "JP_SECOND_WEIGHT")
ReturnSerie <- results[,c("JP_STRATEGY_RETURN")]
turnover <- RP_GetTurnOver(WeightMatrix, -1)
RP_ReturnStats(ReturnSerie, 21, TRUE, WeightMatrix, 0.05, F)
```

```
## Warning in if (is.na(WeightMatrix)) {: the condition has length > 1 and
## only the first element will be used
```

```

## *****
## ***** STRATEGY STATISTICS *****
## *****
## ***** NO FEE *****
## *****
## Annualized Return:      0.114882853523995
## Annualized Volatility:  0.104721540461109
## Information Ratio:      1.09703173786543
## Hit Ratio:              0.631578947368421
## W/L Ratio:              1.40742956601709
## Turnover:               0.490458149369551
## P-Value:                0.00265980532210497
## Max Drawdown:           -0.139656251691509
## Drawdown Recovery:      11
## Break-Even Fee (bps):   98.0669233819284
## *****
## *****
## ***** WITH FEE *****
## *****
## Annualized Return:      0.109053618673122
## Annualized Volatility:  0.104735608036695
## Information Ratio:      1.04122772300051
## Hit Ratio:              0.631578947368421
## W/L Ratio:              1.34587491838384
## Turnover:               0.490458149369551
## P-Value:                0.00425858465997352
## Max Drawdown:           -0.142228735529995
## Drawdown Recovery:      11
## *****
## *****
## *****
## $STATS
## Annualized_Return Annualized_Return_Fee Annualized_Volatility
## 1 0.1148829 0.1090536 0.1047215
## Annualized_Volatility_Fee IR IR_Fee TurnOver HitRatio
## 1 0.1047356 1.097032 1.041228 0.4904581 0.6315789
## HitRatio_Fee WLRatio WLRatio_Fee PVal PVal_Fee BreakEvenFee
## 1 0.6315789 1.40743 1.345875 0.002659805 0.004258585 98.06692
## maxDrawdown drawdownRecovery maxDrawdown_Fee drawdownRecovery_Fee
## 1 -0.1396563 11 -0.1422287 11
##
## $RETURN
## [1] -3.686762e-02 2.888366e-02 -2.971248e-02 7.022668e-03 -2.557431e-02
## [6] 3.488884e-02 3.674268e-03 -7.509945e-03 4.067472e-02 -7.435743e-05
## [11] -1.834110e-02 2.673699e-02 -4.208073e-03 -2.278333e-02 5.964611e-03
## [16] -4.465535e-02 1.211611e-02 6.926467e-02 6.093844e-02 8.212850e-02
## [21] -5.388129e-03 -5.468665e-02 -8.495474e-02 2.939115e-02 -8.117029e-03
## [26] 1.920339e-03 1.467937e-03 3.151046e-02 -5.363361e-03 6.224561e-03
## [31] 3.102396e-02 4.589596e-02 1.268789e-02 2.061457e-02 7.662860e-02
## [36] -2.634674e-03 2.624911e-02 3.450383e-02 1.596120e-02 1.933608e-02
## [41] 6.302224e-03 2.447606e-02 -3.159381e-02 1.208670e-02 -1.344319e-03
## [46] -4.280931e-02 2.936288e-02 2.122258e-02 -6.818524e-03 4.480623e-02
## [51] -1.828946e-02 2.681611e-03 -1.551329e-02 1.759670e-03 1.295970e-02
## [56] 7.707929e-03 -3.209309e-02 -3.770608e-03 1.743648e-02 -1.548602e-02

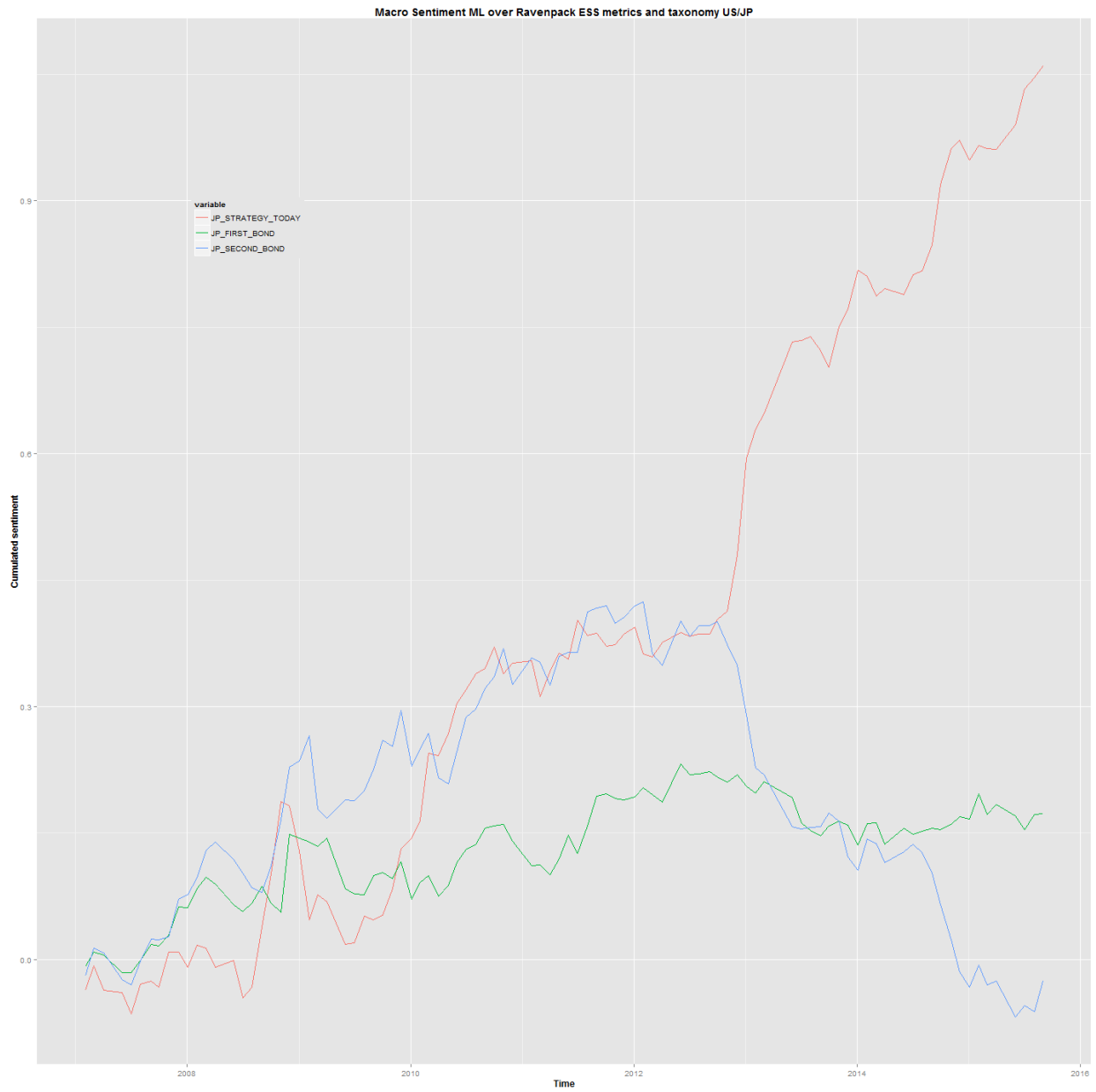
```

```
## [61] -3.782116e-03  1.942672e-03  1.349788e-04  1.720638e-02  9.931921e-03
## [66]  6.383025e-02  1.089873e-01  3.444364e-02  1.819703e-02 -5.769111e-03
## [71]  1.515723e-03  4.486890e-03 -1.662597e-02 -1.993546e-02  4.618041e-02
## [76]  2.222911e-02  4.453120e-02 -7.512173e-03 -2.346509e-02  9.071041e-03
## [81] -8.154362e-03  2.332726e-02  4.616968e-03  3.011477e-02  6.918661e-02
## [86]  4.211462e-02  9.179887e-03 -2.337121e-02  1.694828e-02 -3.736111e-03
## [91] -1.149941e-03 -1.168736e-02  4.221740e-02  1.334742e-02  1.301479e-02
##
## $RETURN_FEE
## [1] -0.0373865310  0.0286267205 -0.0307431645  0.0060291731 -0.0260590465
## [6]  0.0341104470  0.0026774387 -0.0080995121  0.0397141131 -0.0001411854
## [11] -0.0193601266  0.0257628982 -0.0045192531 -0.0229538134  0.0053016893
## [16] -0.0446553500  0.0114572516  0.0686424220  0.0605153029  0.0815045903
## [21] -0.0060691231 -0.0557434128 -0.0860440013  0.0284196465 -0.0087135466
## [26]  0.0015125586  0.0008425505  0.0310071179 -0.0063692441  0.0060805101
## [31]  0.0301946766  0.0449403601  0.0119278056  0.0204849740  0.0757019321
## [36] -0.0027814836  0.0253288338  0.0341996484  0.0158324724  0.0190845511
## [41]  0.0062641584  0.0242333620 -0.0317370386  0.0118997491 -0.0014678519
## [46] -0.0431572814  0.0289495222  0.0209113443 -0.0073489702  0.0448011859
## [51] -0.0186158463  0.0018056100 -0.0160912586  0.0017182992  0.0128554790
## [56]  0.0075266721 -0.0322952559 -0.0039818190  0.0174310303 -0.0157012761
## [61] -0.0041531689  0.0012940645  0.0001193068  0.0165830195  0.0092716253
## [66]  0.0635079121  0.1087115774  0.0341215404  0.0172145791 -0.0067754025
## [71]  0.0012210894  0.0042203313 -0.0170846247 -0.0208419849  0.0452250878
## [76]  0.0215884024  0.0438933668 -0.0075695045 -0.0237481268  0.0083412920
## [81] -0.0089709324  0.0230015604  0.0039721691  0.0297059125  0.0689367703
## [86]  0.0418817774  0.0086787773 -0.0238889055  0.0167872938 -0.0039354056
## [91] -0.0014837133 -0.0119563861  0.0416428086  0.0129505450  0.0126177892
```

```
toplot_df <- melt(results[,c("DATES", "JP_STRATEGY_TODAY", "JP_FIRST_BOND", "JP_SECOND_BOND")], "DATES")

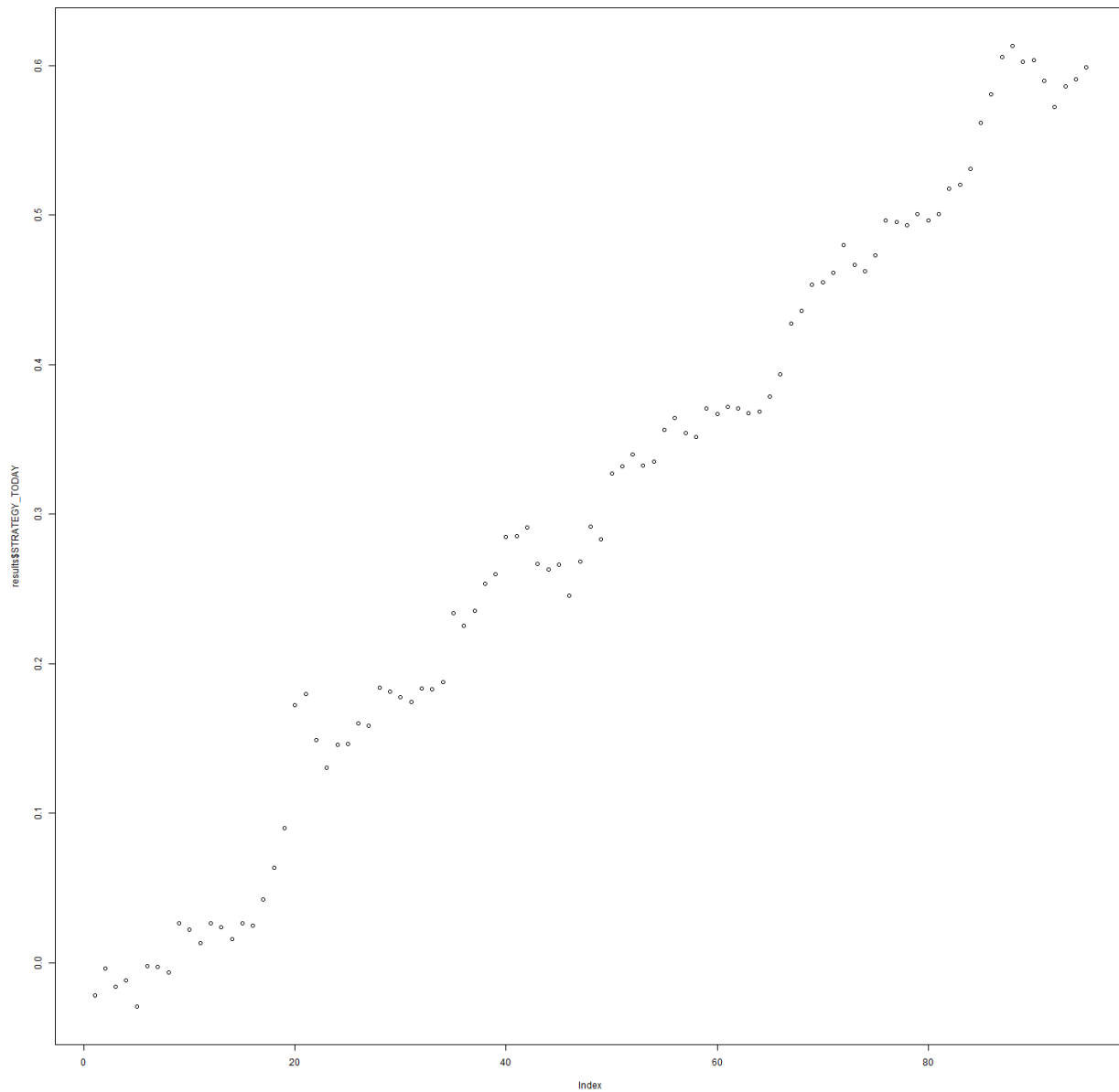
my_title <- paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ", my_pair[1], "/", my_pair[2])
g <- ggplot(
  toplot_df, aes(
    x = DATES, y = value, group = variable, color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2, 0.8), legend.box = "vertical") +
  theme(legend.background = element_rect(fill = "gray90")) +
  theme(legend.key.size = unit(0.7, "cm"))
print(g)
```





# Equally weighted trading strategy plot

```
plot(results$STRATEGY_TODAY)
```



```
# Equally weighted trading strategy metrics
```

```
print("Outputing statistical results for our equally weighted strategy")
```

```
## [1] "Outputing statistical results for our equally weighted strategy"
```

```
WeightMatrix <- results[,c("DATES", "US_WEIGHT", "DE_WEIGHT", "JP_WEIGHT", "GB_WEIGHT")]
colnames(WeightMatrix) <- c("DATE", "US_WEIGHT", "DE_WEIGHT", "JP_WEIGHT", "GB_WEIGHT")
ReturnSerie <- results[,c("STRATEGY_RETURN")]
turnover <- RP_GetTurnOver(WeightMatrix, -1)
RP_ReturnStats(ReturnSerie, 21, TRUE, WeightMatrix, 0.05, F)
```

```
## Warning in if (is.na(WeightMatrix)) {: the condition has length > 1 and
## only the first element will be used
```

```

## *****
## ***** STRATEGY STATISTICS *****
## *****
## ***** NO FEE *****
## *****
## Annualized Return: 0.073791949568223
## Annualized Volatility: 0.0563971157998027
## Information Ratio: 1.3084348112795
## Hit Ratio: 0.621052631578947
## W/L Ratio: 1.78201338802681
## Turnover: 0.543292558707383
## P-Value: 0.000386851381761672
## Max Drawdown: -0.0492227406580829
## Drawdown Recovery: 5
## Break-Even Fee (bps): 56.7675276537005
## *****
## *****
## ***** WITH FEE *****
## *****
## Annualized Return: 0.0673098508982655
## Annualized Volatility: 0.056428204113902
## Information Ratio: 1.19284056537398
## Hit Ratio: 0.610526315789474
## W/L Ratio: 1.68619995447791
## Turnover: 0.543292558707383
## P-Value: 0.00114129791587543
## Max Drawdown: -0.0508020220834215
## Drawdown Recovery: 12
## *****
## *****
## *****
## $STATS
## Annualized_Return Annualized_Return_Fee Annualized_Volatility
## 1 0.07379195 0.06730985 0.05639712
## Annualized_Volatility_Fee IR IR_Fee TurnOver HitRatio
## 1 0.0564282 1.308435 1.192841 0.5432926 0.6210526
## HitRatio_Fee WLRatio WLRatio_Fee PVal PVal_Fee BreakEvenFee
## 1 0.6105263 1.782013 1.6862 0.0003868514 0.001141298 56.76753
## maxDrawdown drawdownRecovery maxDrawdown_Fee drawdownRecovery_Fee
## 1 -0.04922274 5 -0.05080202 12
##
## $RETURN
## [1] -0.0219460399 0.0176421752 -0.0123165215 0.0045131533 -0.0177331437
## [6] 0.0264583292 -0.0004836522 -0.0036563629 0.0320696956 -0.0039078211
## [11] -0.0092678937 0.0131456002 -0.0024285697 -0.0079179938 0.0106430607
## [16] -0.0020976632 0.0176099533 0.0211200647 0.0258570079 0.0793937057
## [21] 0.0071903931 -0.0309987406 -0.0188069931 0.0153572277 0.0004582643
## [26] 0.0135630792 -0.0016863516 0.0251319771 -0.0024740518 -0.0038778577
## [31] -0.0028720798 0.0089662849 -0.0008883985 0.0049399257 0.0450859355
## [36] -0.0083191725 0.0097848788 0.0179835029 0.0061518079 0.0247268355
## [41] 0.0007528944 0.0058480313 -0.0250197866 -0.0035079795 0.0032274647
## [46] -0.0208525253 0.0223729894 0.0232037503 -0.0086902316 0.0428585345
## [51] 0.0050607356 0.0075993884 -0.0069662954 0.0022535977 0.0213800616
## [56] 0.0077725672 -0.0099649645 -0.0028264223 0.0189064209 -0.0035420610

```

```
## [61] 0.0044782358 -0.0008146225 -0.0032507085 0.0009909278 0.0099973568
## [66] 0.0145144525 0.0334366981 0.0086813424 0.0170543039 0.0017145136
## [71] 0.0066439734 0.0181503377 -0.0130253435 -0.0042370520 0.0105545576
## [76] 0.0228686220 -0.0013432971 -0.0019773175 0.0075451382 -0.0040535643
## [81] 0.0041938212 0.0168230315 0.0023869096 0.0104796466 0.0305328345
## [86] 0.0188628025 0.0245239139 0.0075110160 -0.0105359754 0.0011746327
## [91] -0.0139059057 -0.0179876310 0.0138882557 0.0045919149 0.0077387249
##
## $RETURN_FEE
## [1] -0.0224572648 0.0172241740 -0.0133174884 0.0035171604 -0.0185116895
## [6] 0.0256320426 -0.0009449736 -0.0045498364 0.0314609901 -0.0043430474
## [11] -0.0101648326 0.0123472131 -0.0030323216 -0.0085596817 0.0102553319
## [16] -0.0028495207 0.0171411961 0.0205054428 0.0256324860 0.0785293063
## [21] 0.0064059071 -0.0318114732 -0.0196145155 0.0145269386 -0.0003844931
## [26] 0.0133865564 -0.0021310839 0.0247949648 -0.0033658561 -0.0044366790
## [31] -0.0033110552 0.0083025206 -0.0017842640 0.0043395153 0.0443801291
## [36] -0.0086644243 0.0089923545 0.0173303009 0.0058138278 0.0243950764
## [41] 0.0000270802 0.0051538647 -0.0251750093 -0.0039261083 0.0031473014
## [46] -0.0213304395 0.0218084933 0.0230388414 -0.0090509501 0.0427472235
## [51] 0.0047576188 0.0067887512 -0.0074667113 0.0018894960 0.0206643527
## [56] 0.0072458361 -0.0103437918 -0.0031771747 0.0185203503 -0.0042905715
## [61] 0.0040308401 -0.0011719584 -0.0036184189 0.0005611597 0.0094927374
## [66] 0.0143614388 0.0331635245 0.0083315188 0.0161362754 0.0007841981
## [71] 0.0061435349 0.0176700947 -0.0136229826 -0.0046620045 0.0098052343
## [76] 0.0220974316 -0.0019056206 -0.0021083291 0.0073055016 -0.0048755640
## [81] 0.0036038586 0.0162061036 0.0017394746 0.0096340982 0.0299583499
## [86] 0.0186215155 0.0242795548 0.0069129742 -0.0106440760 0.0006622027
## [91] -0.0144317826 -0.0182453829 0.0133864453 0.0041984423 0.0073618767
```

```
toplot_df <- melt(results[,c("DATES", "STRATEGY_TODAY", "EQ_WEIGHTS_BOND")], "DATES")
my_title <- paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ", sep="")
g<-ggplot(
  toplot_df, aes(
    x = DATES, y = value, group = variable, color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2, 0.8), legend.box = "vertical") +
  theme(legend.background = element_rect(fill = "gray90")) +
  theme(legend.key.size = unit(0.7, "cm"))
print(g)
```



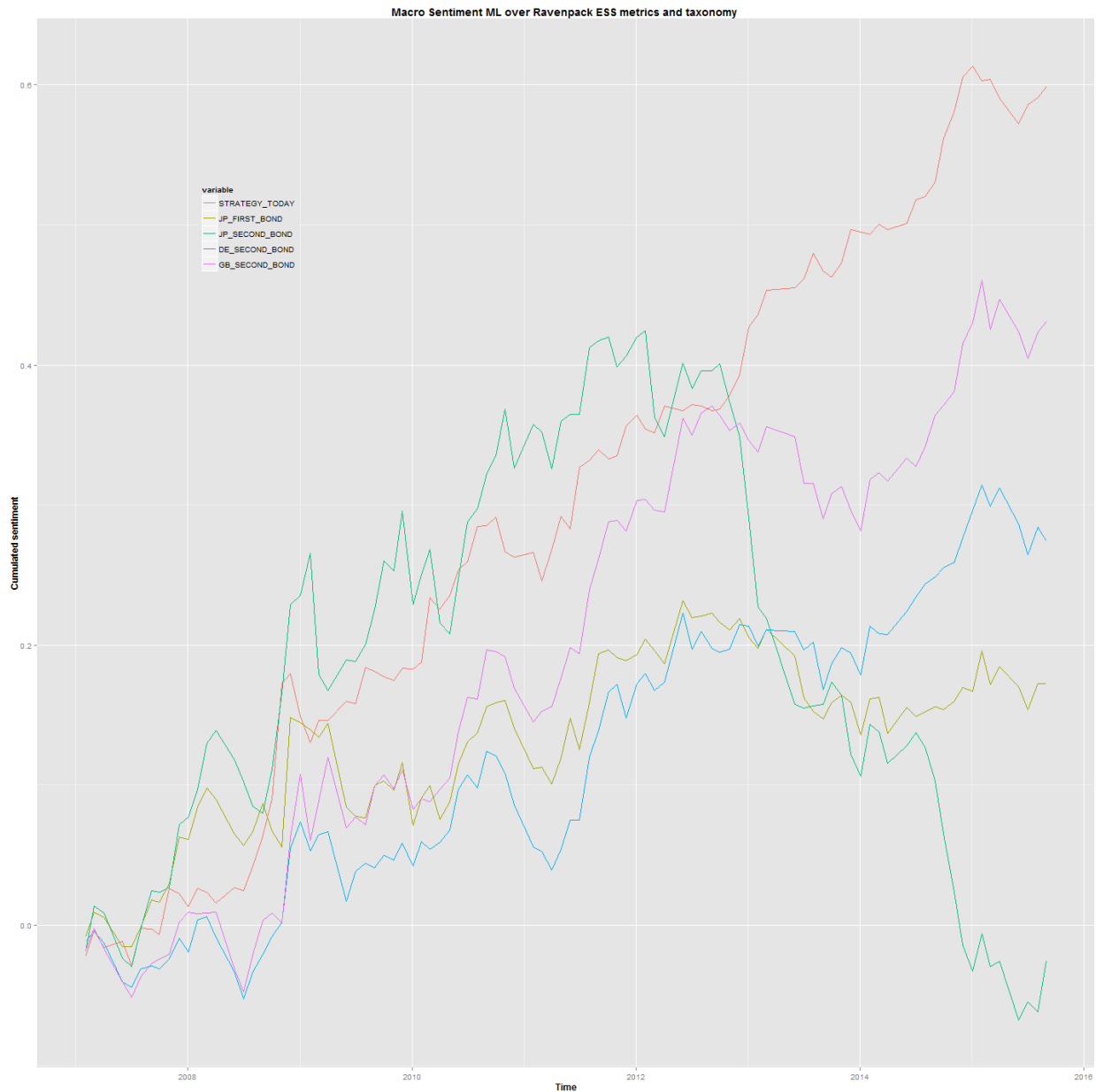
```
# ExportPlot(g,outputDataPathStrategyMonth,paste(my_result_spread_name,"_bench",sep=""))

toplot_df <- melt(results[,c("DATES", "STRATEGY_TODAY", "JP_FIRST_BOND", "JP_SECOND_BOND", "DE_SECOND_BOND")])
my_title <- paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ",sep="")
g<-ggplot(
  toplot_df,aes(
    x = DATES,y = value,group = variable,color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2,0.8), legend.box = "vertical") +
```

```

theme(legend.background = element_rect(fill = "gray90")) +
theme(legend.key.size = unit(0.7, "cm"))
print(g)

```



```

weights_decomposition_df <- data.frame(weights=results$US_WEIGHT,typo="US")
weights_decomposition_df <- rbind(weights_decomposition_df, data.frame(weights=results$JP_WEIGHT,typo="JP"))
weights_decomposition_df <- rbind(weights_decomposition_df, data.frame(weights=results$GB_WEIGHT,typo="GB"))
weights_decomposition_df <- rbind(weights_decomposition_df, data.frame(weights=results$DE_WEIGHT,typo="DE"))
my_title <- "Distribution of weights per country"
my_xaxis_title <- paste("Weights")
my_yaxis_title <- paste("Percentage", "\n")

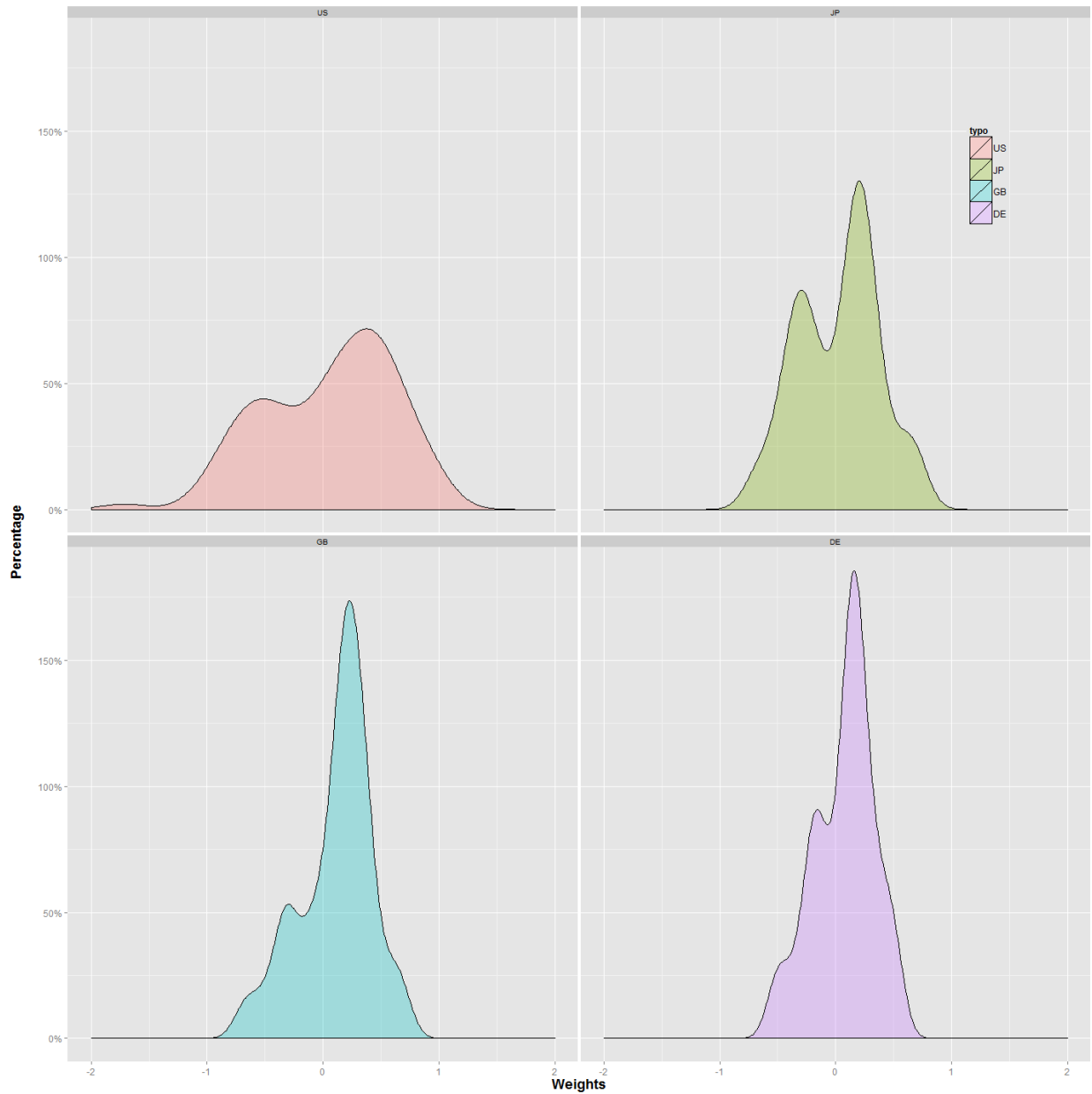
g <- ggplot(weights_decomposition_df, aes(x=weights, fill=typo)) +geom_density(alpha=.3)+

```

```

scale_y_continuous(labels = percent_format()) + scale_x_continuous(limits=c(-2,2)) + facet_wrap(~typo, ncol=2)
g <- g + ylab(my_yaxis_title) + xlab(my_xaxis_title) +
  theme(axis.text.x = element_text(size=12), axis.text.y = element_text(size=12), title = element_text(size=12)) +
  theme(legend.position=c(0.9,0.85), legend.box = "vertical") +
  theme(legend.background = element_rect(fill="gray90")) +
  theme(legend.key.size = unit(1., "cm")) +
  theme(legend.text = element_text(size=12, colour="black")) +
  theme(legend.title = element_text(size=12, colour="black"))
print(g)

```



```

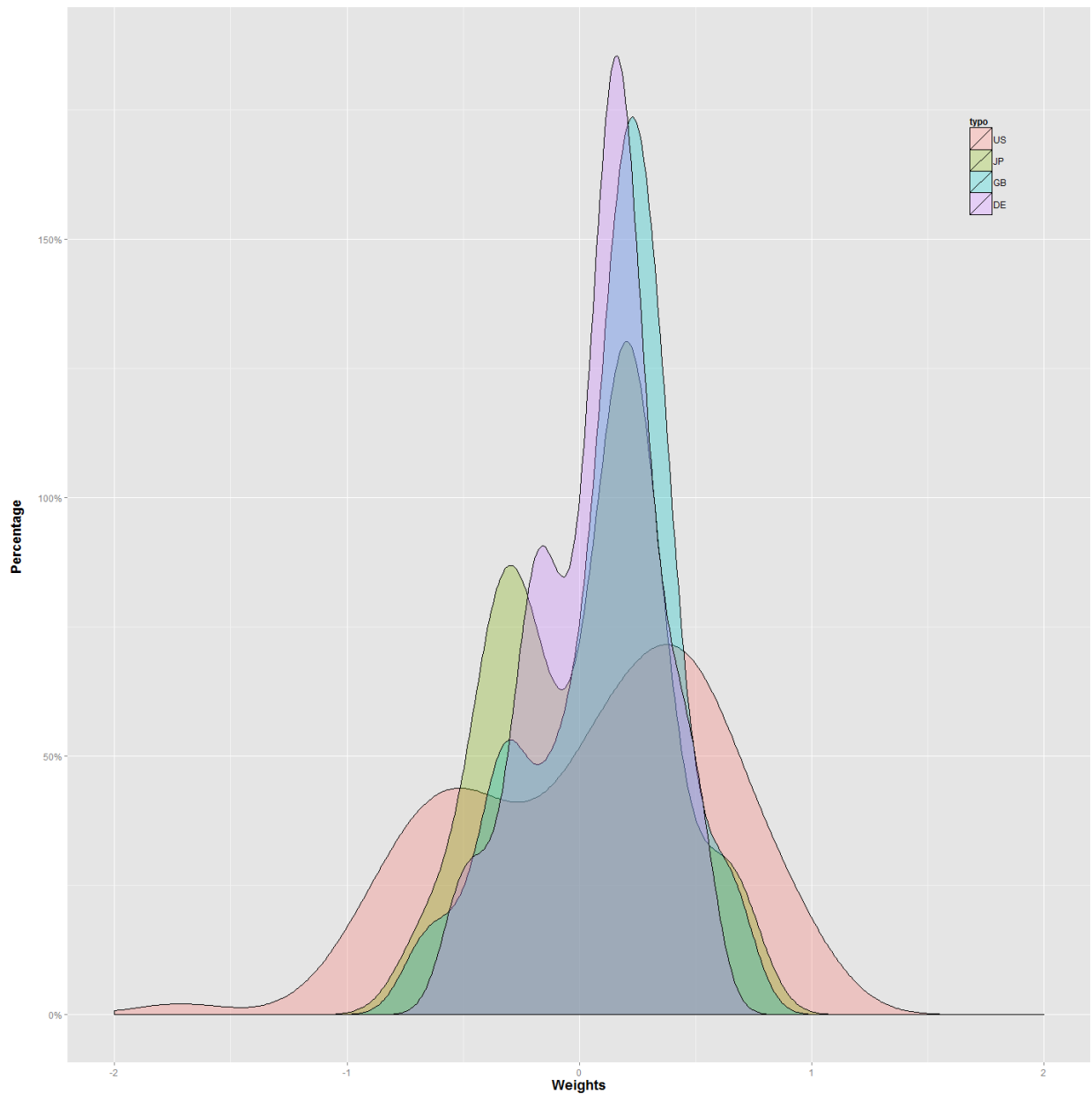
g <- ggplot(weights_decomposition_df, aes(x=weights, fill=typo)) + geom_density(alpha=.3) +
  scale_y_continuous(labels = percent_format()) + scale_x_continuous(limits=c(-2,2))
g <- g + ylab(my_yaxis_title) + xlab(my_xaxis_title) +

```

```

theme(axis.text.x = element_text(size=12),axis.text.y = element_text(size=12),title =element_text(size=12))+
theme(legend.position=c(0.9,0.85), legend.box = "vertical")+
theme(legend.background = element_rect(fill="gray90"))+
theme(legend.key.size = unit(1., "cm"))+
theme(legend.text = element_text(size=12,colour="black"))+
theme(legend.title = element_text(size=12,colour="black"))
print(g)

```



```

toplot_df <- melt(results[,c("DATES","US_WEIGHT","DE_WEIGHT","JP_WEIGHT","GB_WEIGHT")], "DATES")

my_title <-paste("Macro Sentiment ML over Ravenpack ESS metrics and taxonomy ",my_pair[1],"/",my_pair[2])
g<-ggplot(
  toplot_df,aes(

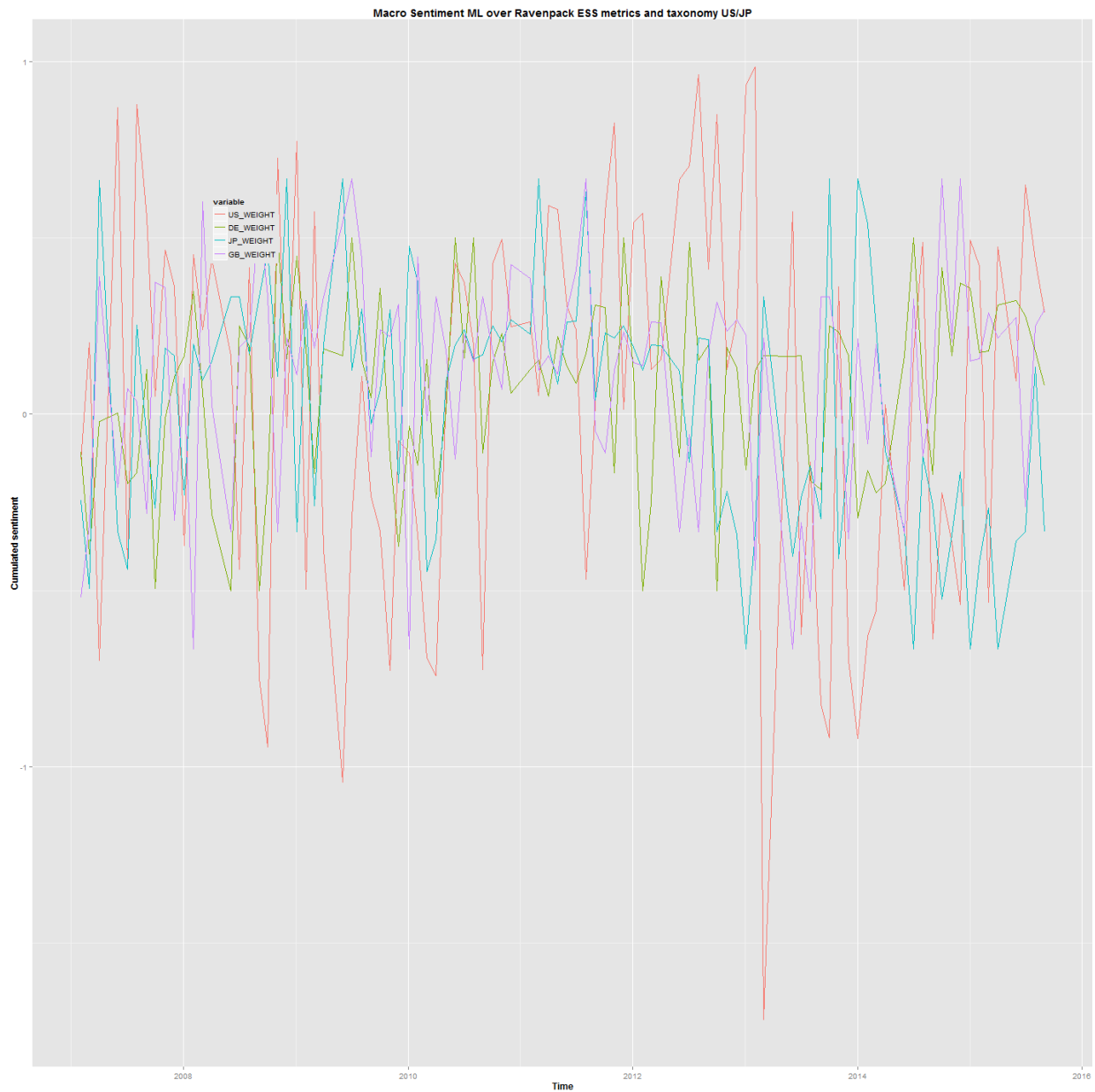
```



```

    x = DATES,y = value,group = variable,color = variable
  )
) +
  geom_line() +
  scale_x_date() +
  ggtitle(my_title) + xlab("Time") + ylab("Cumulated sentiment") +
  theme(title = element_text(size = 12, face = 'bold')) +
  theme(legend.position = c(0.2,0.8), legend.box = "vertical") +
  theme(legend.background = element_rect(fill = "gray90")) #+
  # theme(legend.key.size = unit(0.7, "cm"))
print(g)

```



```

# ExportPlot(g,outputDataPathStrategyMonth,paste(my_result_spread_name,"_weights",sep=""))

prediction_results <- tryCatch( readRDS(paste(outputDataPathStrategyMonth,"all_pairs_prediction_results.rds",sep=""))
                               error = function(e) {NULL})
if (!is.null(prediction_results)){

  # correlation between prediction and returns for US
  print("Correlation US unidirection prediction")
  print(cor(prediction_results$USPrediction,prediction_results$USEffective))

  for (my_quantile in c(0.1,1/5,1/3)){
    US_df <- prediction_results[,c("USPrediction","USEffective")]
    US_df$my_tiles <- with(US_df, cut(USPrediction,
                                     breaks=quantile(USPrediction, probs=seq(0,1, by=my_quantile)),
                                     include.lowest=TRUE))

    my_IR_computation <- function(x){
      IR <- mean(x,na.rm=TRUE)/sd(x,na.rm=TRUE)*sqrt(12)
      return(IR)
    }

    agg_US_df <- aggregate(US_df$USEffective, by=list(US_df$my_tiles),FUN=my_IR_computation)
    colnames(agg_US_df) <- c("Pred_quantile","Mean_Ret")
    ## Histogram plotting of the return average per prediction quantile
    g <- ggplot(agg_US_df, aes(x=Pred_quantile,y=Mean_Ret)) +
      geom_histogram(stat = "identity")
    print(g)
  }

  Predictionbullish <- prediction_results$USPrediction >= 0
  NextReturnbullish <- prediction_results$USEffective >= 0

  Predictionbearish <- prediction_results$USPrediction <= 0
  NextReturnbearish <- prediction_results$USEffective <= 0

  print("Long confusion matrix")
  print(confusionMatrix(Predictionbullish, NextReturnbullish))

  print("Short confusion matrix")
  print(confusionMatrix(Predictionbearish, NextReturnbearish))

  print("Correlation GB unidirection prediction")
  print(cor(prediction_results$GBPrediction,prediction_results$GBEffective))
  for (my_quantile in c(0.1,1/5,1/3)){
    df <- prediction_results[,c("GBPrediction","GBEffective")]
    df$my_tiles <- with(df, cut(GBPrediction,
                               breaks=quantile(GBPrediction, probs=seq(0,1, by=my_quantile)),
                               include.lowest=TRUE))
    agg_df <- aggregate(df$GBEffective, by=list(df$my_tiles),FUN=my_IR_computation)
    colnames(agg_df) <- c("Pred_quantile","Mean_Ret")
    ## Histogram plotting of the return average per prediction quantile
    g <- ggplot(agg_df, aes(x=Pred_quantile,y=Mean_Ret)) +

```

```

    geom_histogram(stat = "identity")
  print(g)
}
Predictionbullish <- prediction_results$GBPrediction >= 0
NextReturnbullish <- prediction_results$GBEffective >= 0

Predictionbearish <- prediction_results$GBPrediction <= 0
NextReturnbearish <- prediction_results$GBEffective <= 0

print("Long confusion matrix")
print(confusionMatrix(Predictionbullish, NextReturnbullish))

print("Short confusion matrix")
print(confusionMatrix(Predictionbearish, NextReturnbearish))

print("Correlation DE unidirection prediction")
print(cor(prediction_results$DEPrediction, prediction_results$DEEffective))
for (my_quantile in c(0.1, 1/5, 1/3)){
  df <- prediction_results[, c("DEPrediction", "DEEffective")]
  df$my_tiles <- with(df, cut(DEPrediction,
                             breaks=quantile(DEPrediction, probs=seq(0,1, by=my_quantile)),
                             include.lowest=TRUE))
  agg_df <- aggregate(df$DEEffective, by=list(df$my_tiles), FUN=my_IR_computation)
  colnames(agg_df) <- c("Pred_quantile", "Mean_Ret")
  ## Histogram plotting of the return average per prediction quantile
  g <- ggplot(agg_df, aes(x=Pred_quantile, y=Mean_Ret)) +
    geom_histogram(stat = "identity")
  print(g)
}
Predictionbullish <- prediction_results$DEPrediction >= 0
NextReturnbullish <- prediction_results$DEEffective >= 0

Predictionbearish <- prediction_results$DEPrediction <= 0
NextReturnbearish <- prediction_results$DEEffective <= 0

print("Long confusion matrix")
print(confusionMatrix(Predictionbullish, NextReturnbullish))

print("Short confusion matrix")
print(confusionMatrix(Predictionbearish, NextReturnbearish))

print("Correlation GB unidirection prediction")
print(cor(prediction_results$GBPrediction, prediction_results$GBEffective))

Predictionbullish <- prediction_results$GBPrediction >= 0
NextReturnbullish <- prediction_results$GBEffective >= 0

Predictionbearish <- prediction_results$GBPrediction <= 0
NextReturnbearish <- prediction_results$GBEffective <= 0

print("Long confusion matrix")
print(confusionMatrix(Predictionbullish, NextReturnbullish))

```

```

print("Short confusion matrix")
print(confusionMatrix(Predictionbearish, NextReturnbearish))

print("Correlation JP unidirection prediction")
print(cor(prediction_results$JPPrediction, prediction_results$JPEffective))
for (my_quantile in c(0.1, 1/5, 1/3)){
  df <- prediction_results[,c("JPPrediction", "JPEffective")]
  df$my_tiles <- with(df, cut(JPPrediction,
                             breaks=quantile(JPPrediction, probs=seq(0,1, by=my_quantile)),
                             include.lowest=TRUE))
  agg_df <- aggregate(df$JPEffective, by=list(df$my_tiles), FUN=my_IR_computation)
  colnames(agg_df) <- c("Pred_quantile", "Mean_Ret")
  ## Histogram plotting of the return average per prediction quantile
  g <- ggplot(agg_df, aes(x=Pred_quantile, y=Mean_Ret)) +
    geom_histogram(stat = "identity")
  print(g)
}

Predictionbullish <- prediction_results$JPPrediction >= 0
NextReturnbullish <- prediction_results$JPEffective >= 0

Predictionbearish <- prediction_results$JPPrediction <= 0
NextReturnbearish <- prediction_results$JPEffective <= 0

print("Long confusion matrix")
print(confusionMatrix(Predictionbullish, NextReturnbullish))

print("Short confusion matrix")
print(confusionMatrix(Predictionbearish, NextReturnbearish))
}

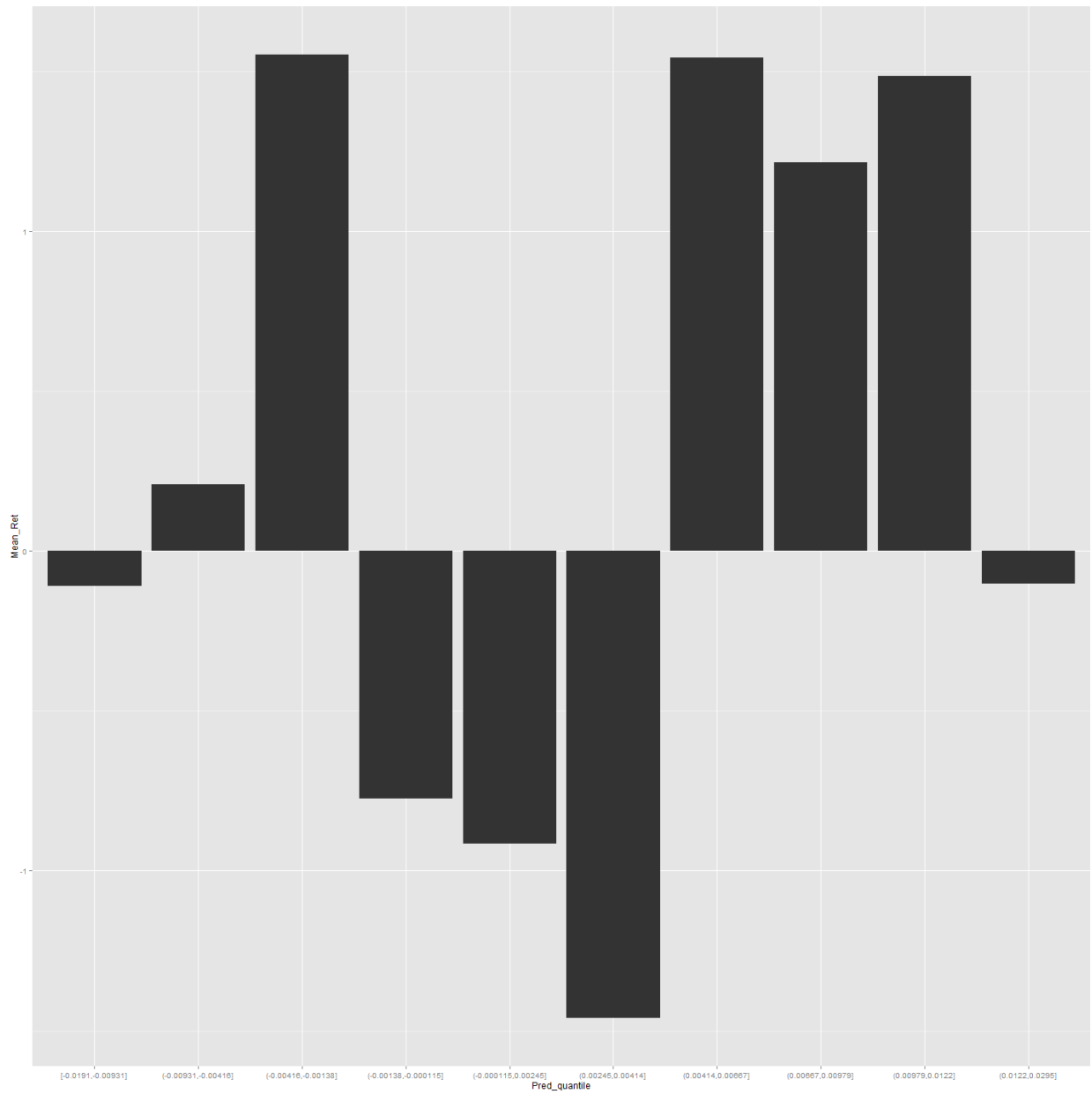
```

```

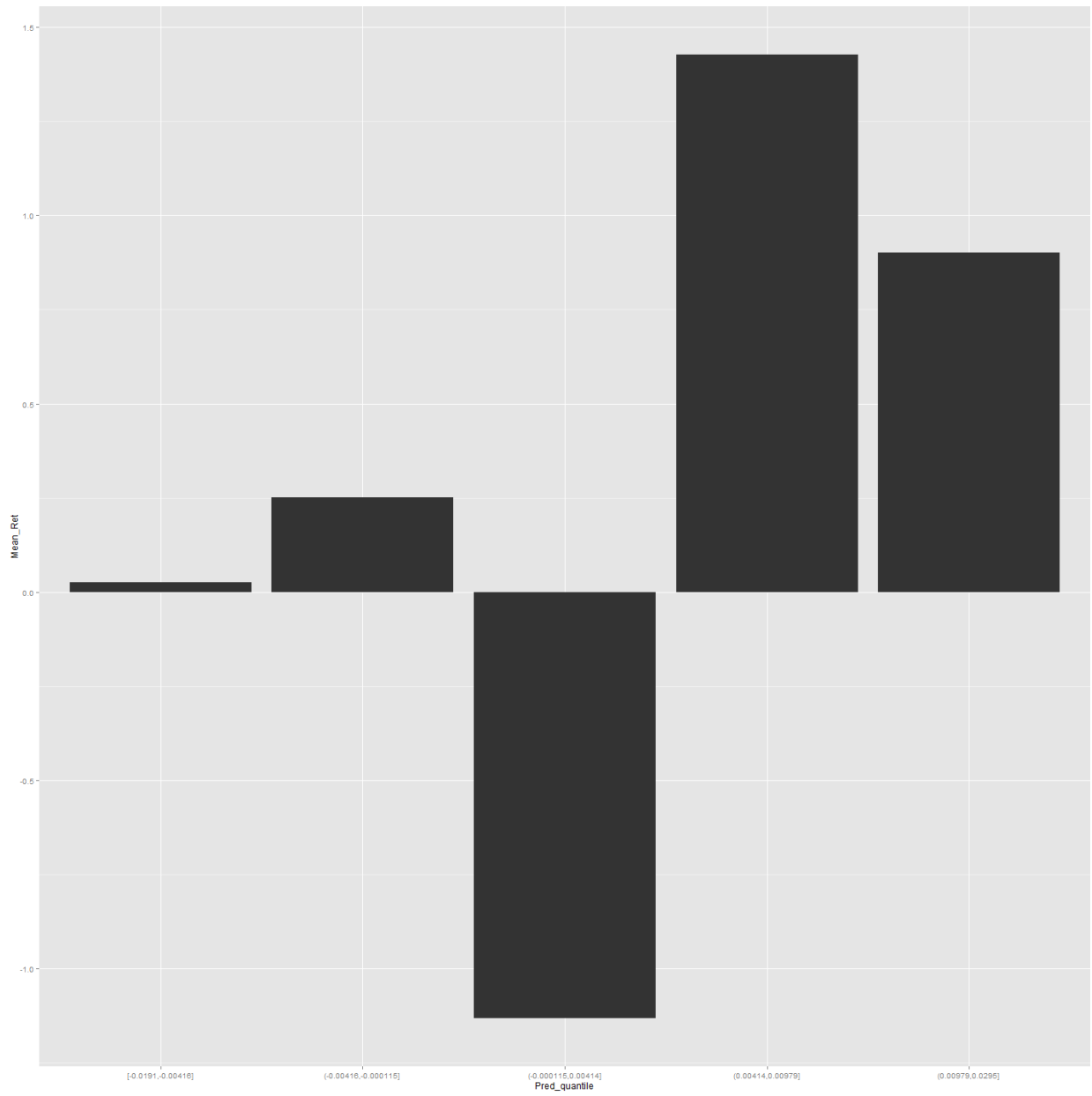
## [1] "Correlation US unidirection prediction"
## [1] 0.0880182

## Warning: Stacking not well defined when ymin != 0

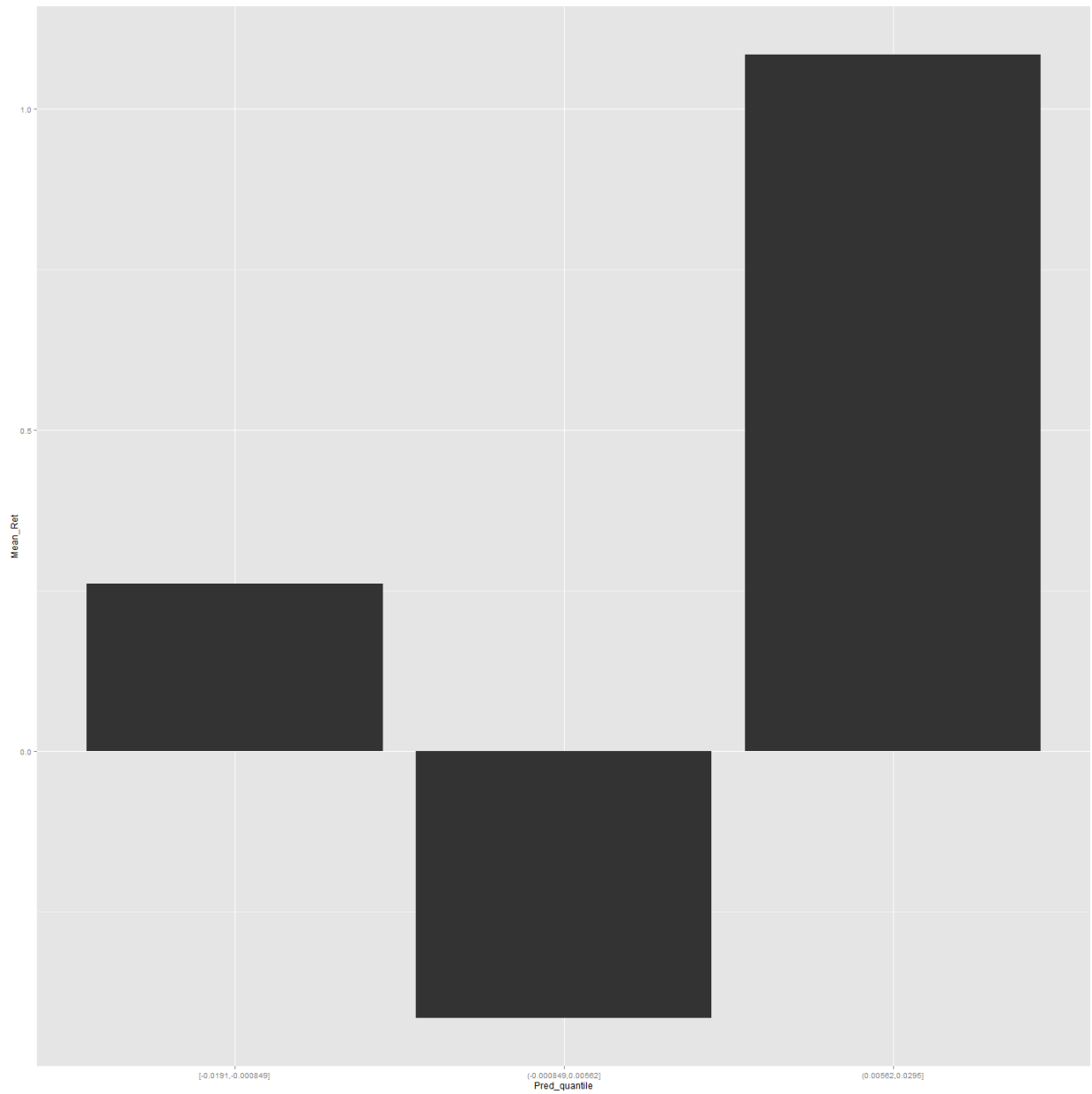
```



## Warning: Stacking not well defined when ymin != 0



`## Warning: Stacking not well defined when ymin != 0`



```
## [1] "Long confusion matrix"
## Confusion Matrix and Statistics
##
##           Reference
## Prediction FALSE TRUE
##      FALSE      20   22
##      TRUE       30   32
##
##           Accuracy : 0.5
##           95% CI : (0.4003, 0.5997)
##      No Information Rate : 0.5192
##      P-Value [Acc > NIR] : 0.6884
##
##           Kappa : -0.0075
```

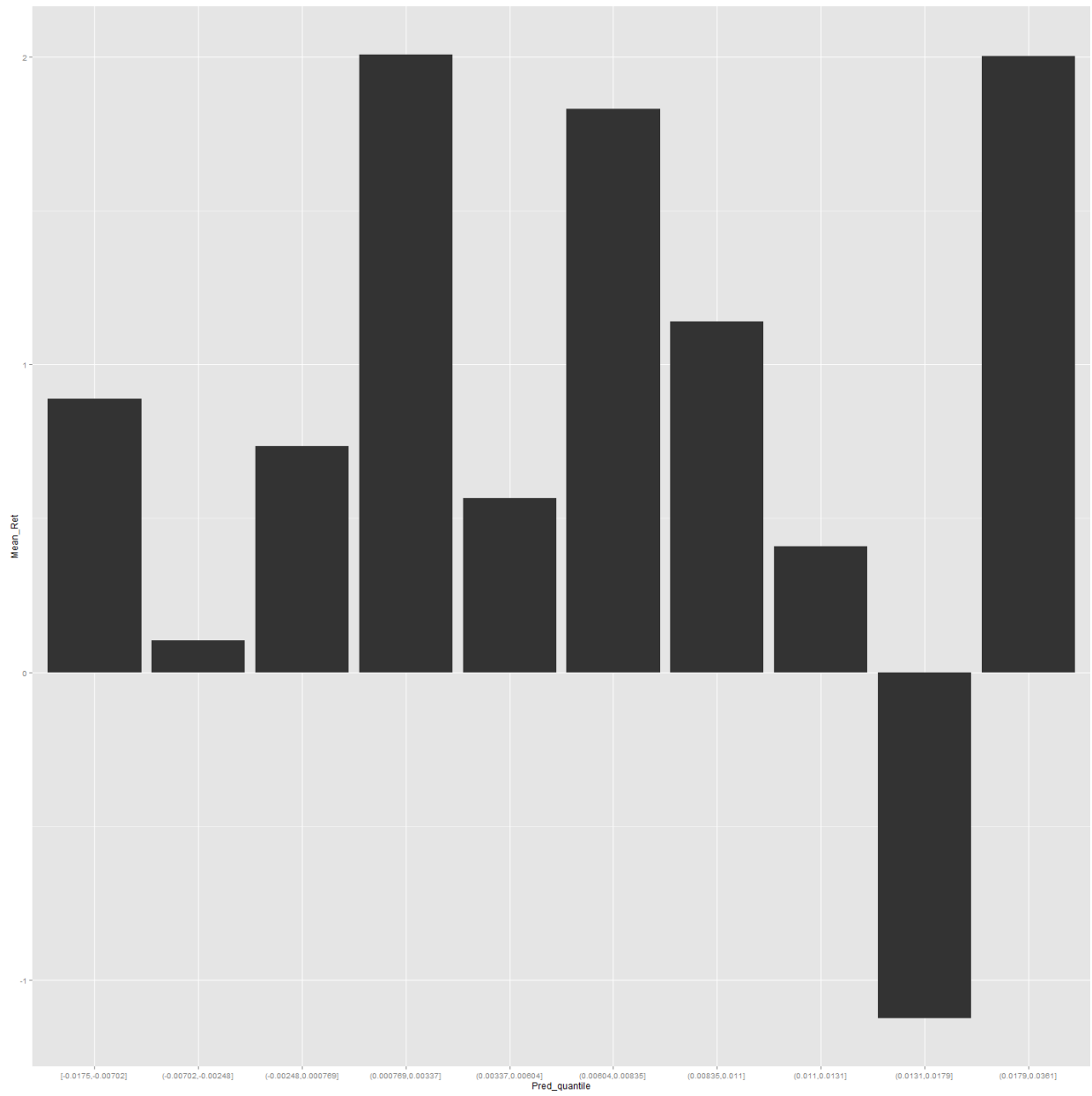
```

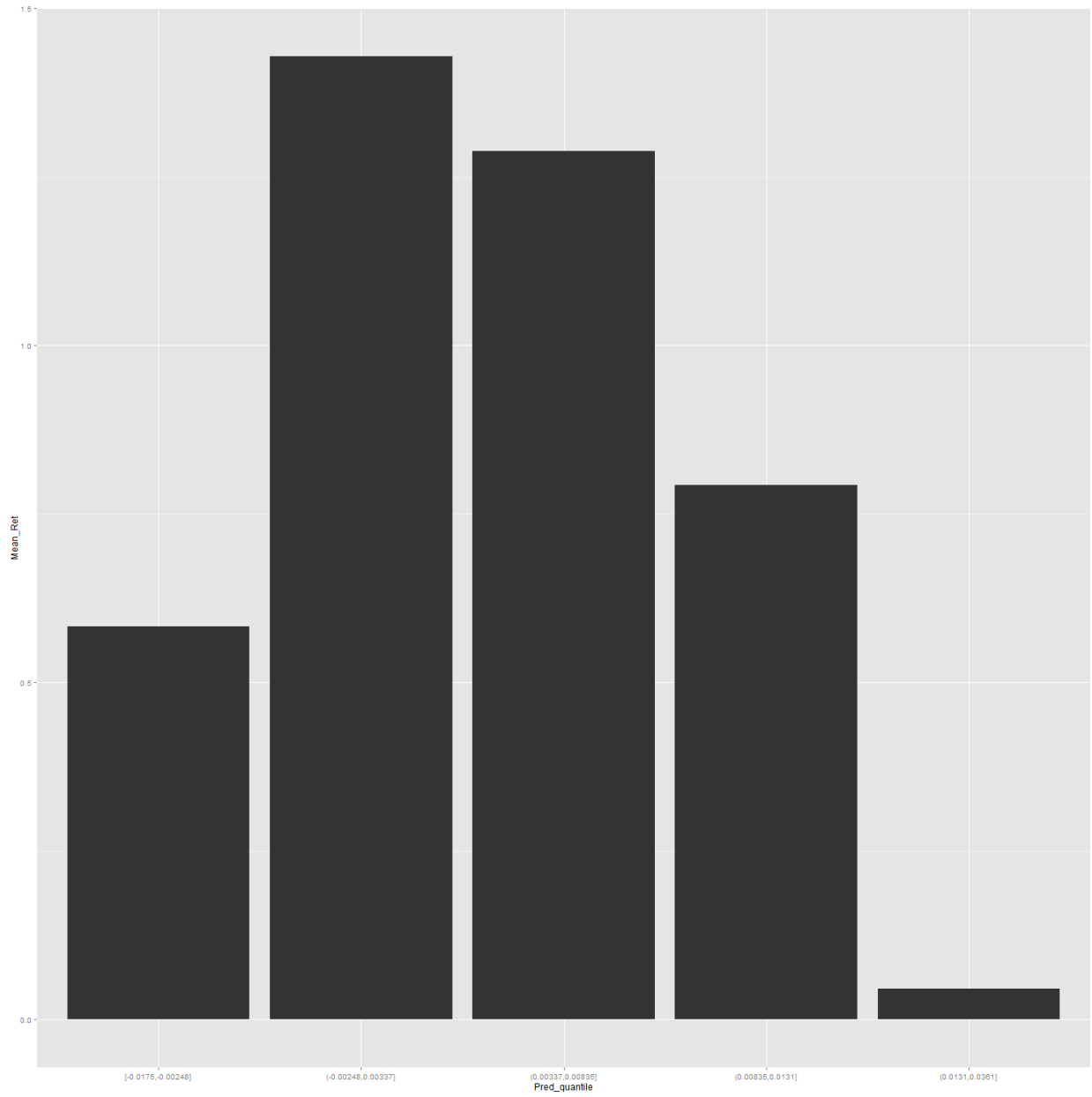
## McNemar's Test P-Value : 0.3317
##
##      Sensitivity : 0.4000
##      Specificity : 0.5926
##      Pos Pred Value : 0.4762
##      Neg Pred Value : 0.5161
##      Prevalence : 0.4808
##      Detection Rate : 0.1923
##      Detection Prevalence : 0.4038
##      Balanced Accuracy : 0.4963
##
##      'Positive' Class : FALSE
##
## [1] "Short confusion matrix"
## Confusion Matrix and Statistics
##
##      Reference
## Prediction FALSE TRUE
##      FALSE      32   30
##      TRUE       22   20
##
##      Accuracy : 0.5
##      95% CI : (0.4003, 0.5997)
##      No Information Rate : 0.5192
##      P-Value [Acc > NIR] : 0.6884
##
##      Kappa : -0.0075
## McNemar's Test P-Value : 0.3317
##
##      Sensitivity : 0.5926
##      Specificity : 0.4000
##      Pos Pred Value : 0.5161
##      Neg Pred Value : 0.4762
##      Prevalence : 0.5192
##      Detection Rate : 0.3077
##      Detection Prevalence : 0.5962
##      Balanced Accuracy : 0.4963
##
##      'Positive' Class : FALSE
##
## [1] "Correlation GB unidirection prediction"
## [1] 0.006945995

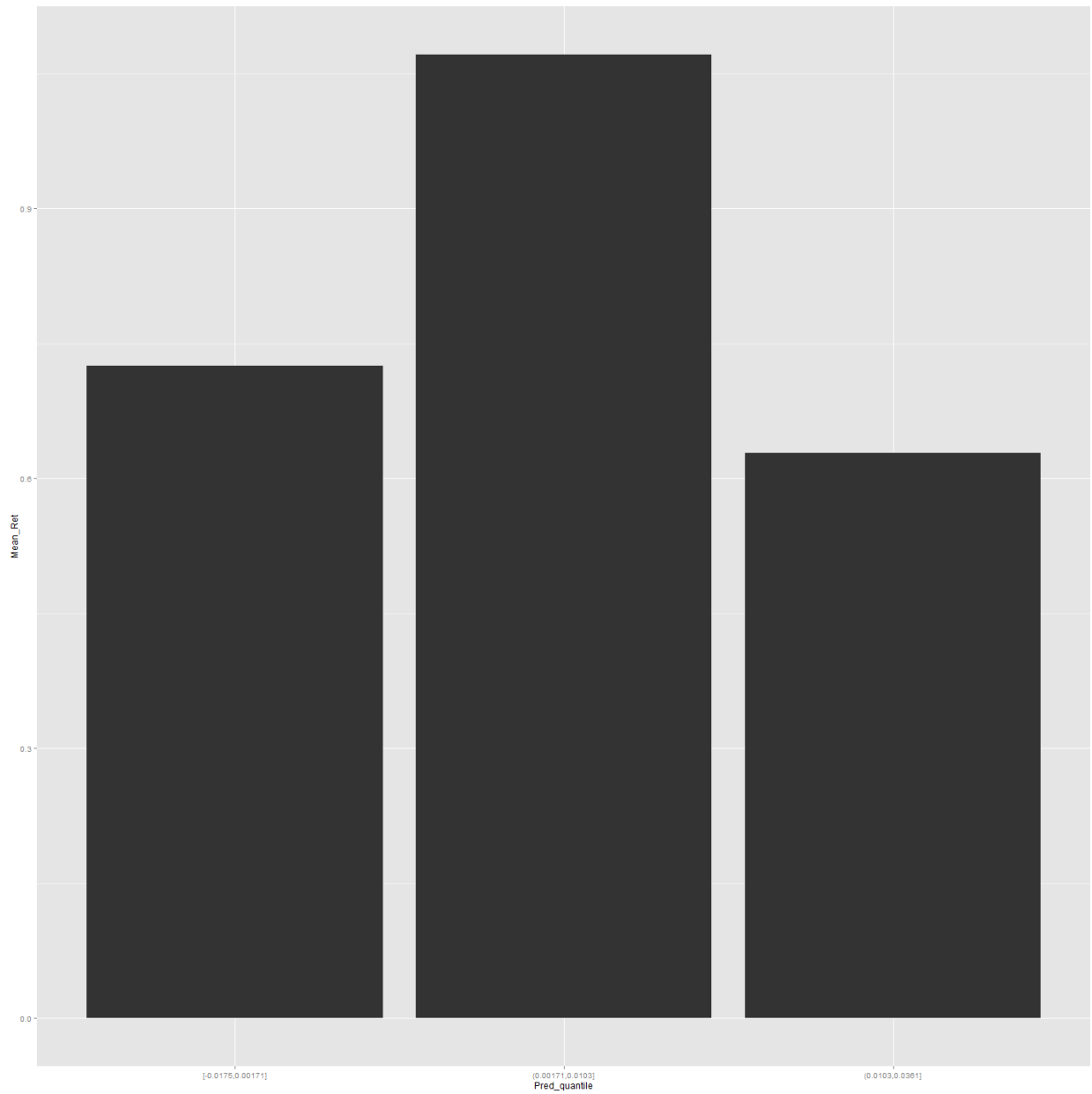
## Warning: Stacking not well defined when ymin != 0

```









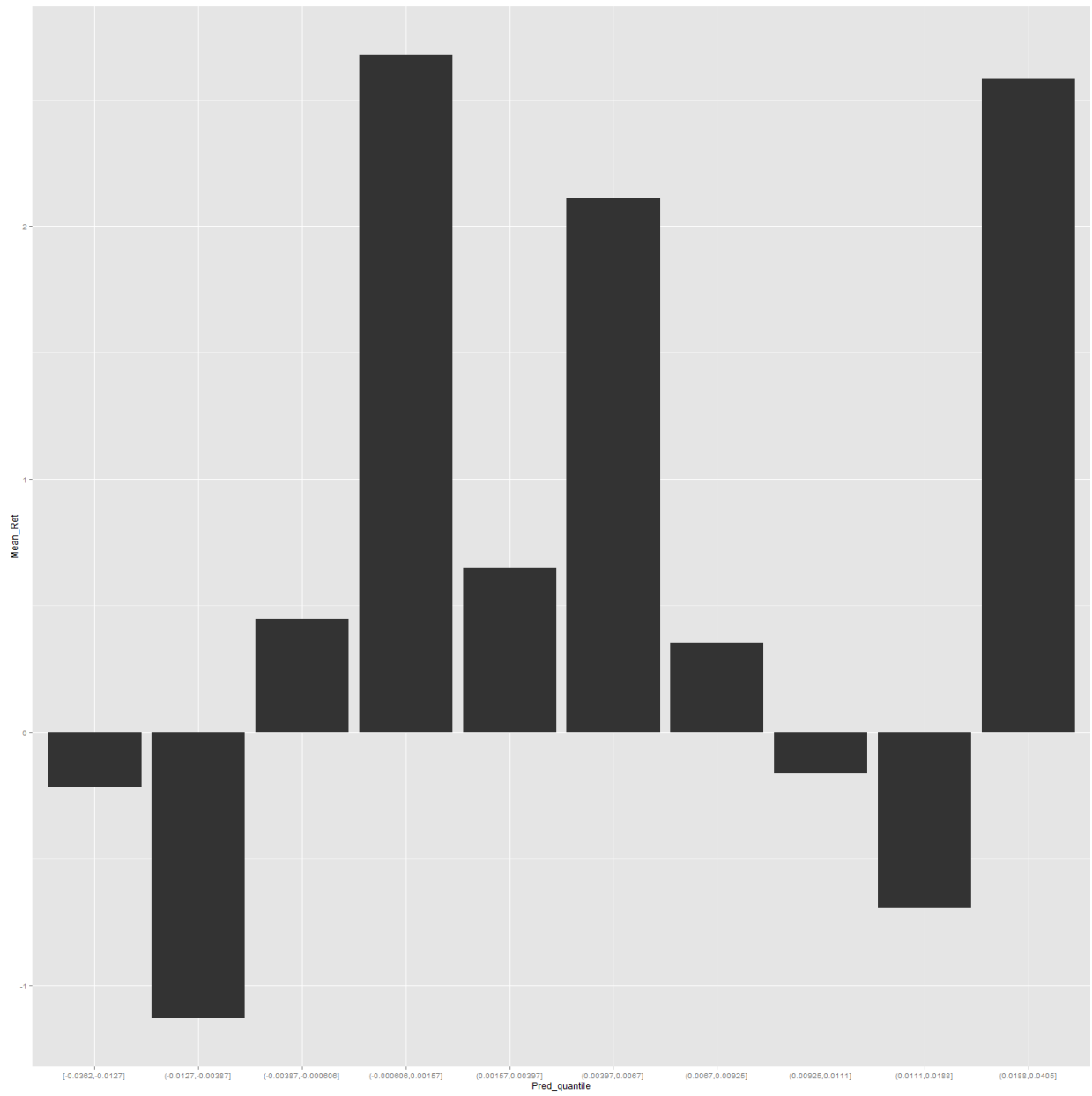
```
## [1] "Long confusion matrix"
## Confusion Matrix and Statistics
##
##           Reference
## Prediction FALSE TRUE
##      FALSE     11  17
##      TRUE      34  42
##
##           Accuracy : 0.5096
##           95% CI : (0.4097, 0.609)
##      No Information Rate : 0.5673
##      P-Value [Acc > NIR] : 0.90048
##
##           Kappa : -0.0457
```

```

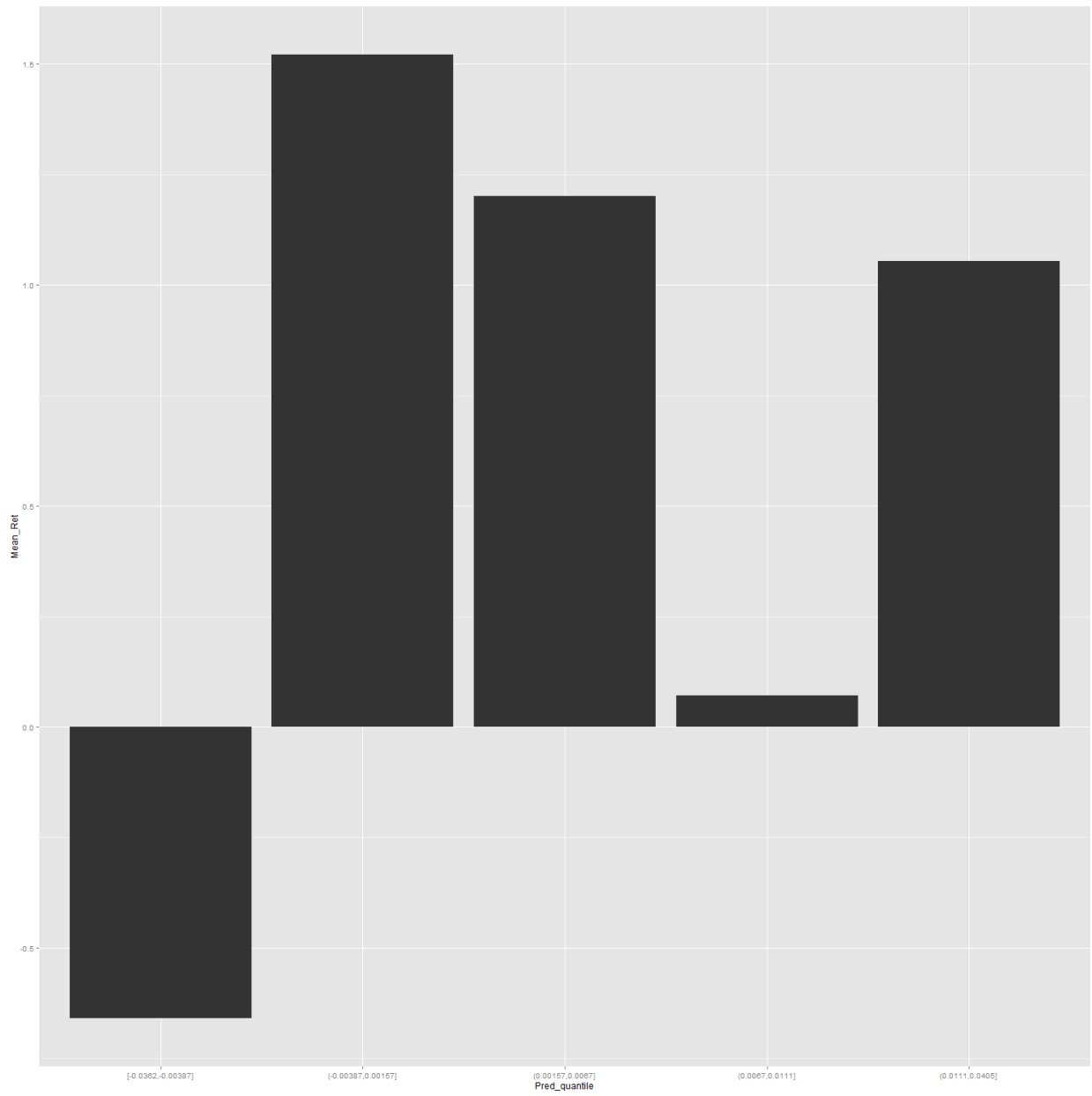
## McNemar's Test P-Value : 0.02506
##
##      Sensitivity : 0.2444
##      Specificity : 0.7119
##      Pos Pred Value : 0.3929
##      Neg Pred Value : 0.5526
##      Prevalence : 0.4327
##      Detection Rate : 0.1058
##      Detection Prevalence : 0.2692
##      Balanced Accuracy : 0.4782
##
##      'Positive' Class : FALSE
##
## [1] "Short confusion matrix"
## Confusion Matrix and Statistics
##
##      Reference
## Prediction FALSE TRUE
##      FALSE      42   34
##      TRUE       17   11
##
##      Accuracy : 0.5096
##      95% CI : (0.4097, 0.609)
##      No Information Rate : 0.5673
##      P-Value [Acc > NIR] : 0.90048
##
##      Kappa : -0.0457
## McNemar's Test P-Value : 0.02506
##
##      Sensitivity : 0.7119
##      Specificity : 0.2444
##      Pos Pred Value : 0.5526
##      Neg Pred Value : 0.3929
##      Prevalence : 0.5673
##      Detection Rate : 0.4038
##      Detection Prevalence : 0.7308
##      Balanced Accuracy : 0.4782
##
##      'Positive' Class : FALSE
##
## [1] "Correlation DE unidirection prediction"
## [1] 0.2137453

## Warning: Stacking not well defined when ymin != 0

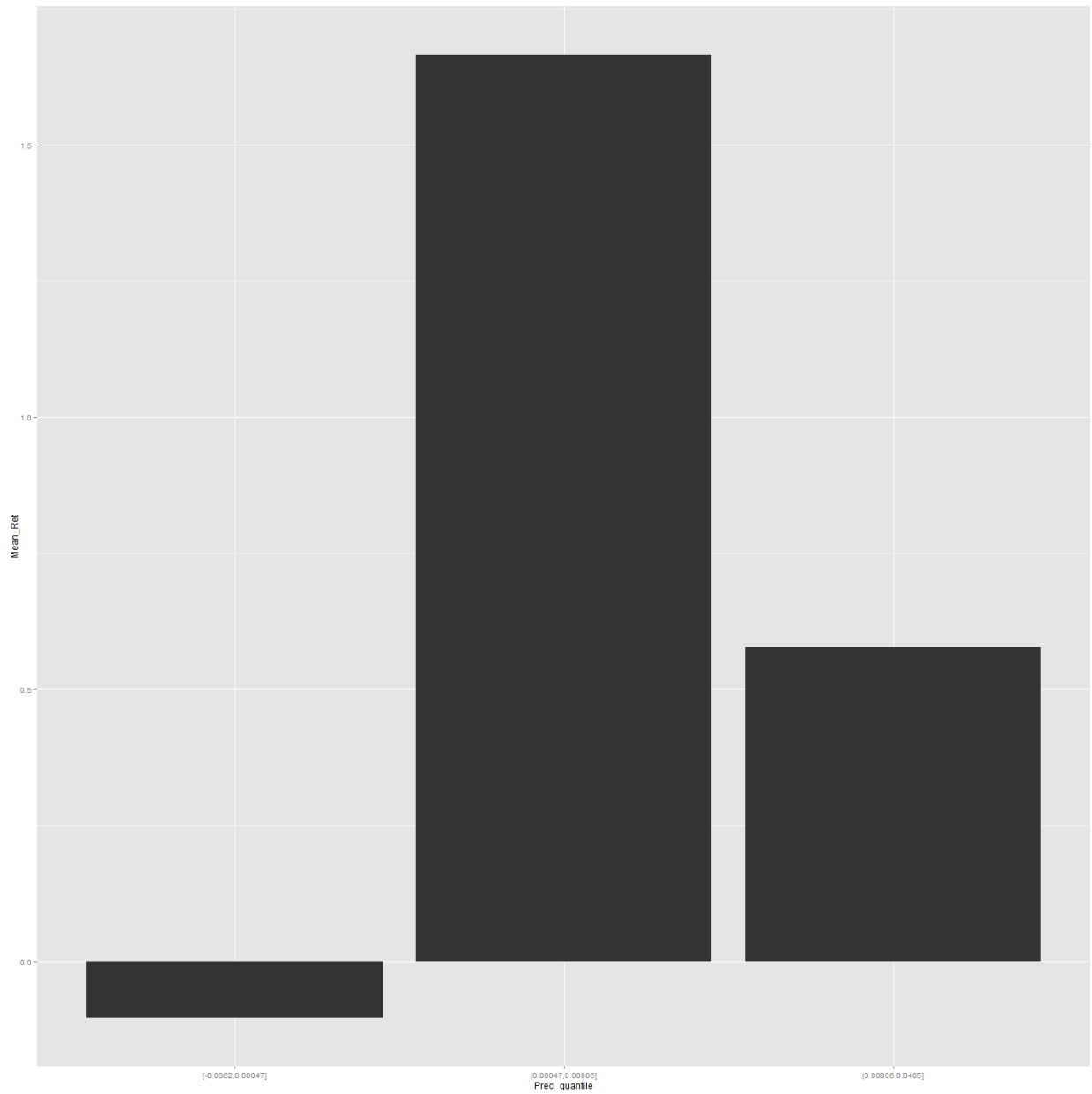
```



## Warning: Stacking not well defined when ymin != 0



`## Warning: Stacking not well defined when ymin != 0`



```
## [1] "Long confusion matrix"
## Confusion Matrix and Statistics
##
##           Reference
## Prediction FALSE TRUE
##      FALSE    15   18
##      TRUE     29   42
##
##           Accuracy : 0.5481
##           95% CI : (0.4474, 0.6459)
##      No Information Rate : 0.5769
##      P-Value [Acc > NIR] : 0.7571
##
##           Kappa : 0.0423
```

```

## McNemar's Test P-Value : 0.1447
##
##      Sensitivity : 0.3409
##      Specificity : 0.7000
##      Pos Pred Value : 0.4545
##      Neg Pred Value : 0.5915
##      Prevalence : 0.4231
##      Detection Rate : 0.1442
##      Detection Prevalence : 0.3173
##      Balanced Accuracy : 0.5205
##
##      'Positive' Class : FALSE
##
## [1] "Short confusion matrix"
## Confusion Matrix and Statistics
##
##      Reference
## Prediction FALSE TRUE
##      FALSE      42   29
##      TRUE       18   15
##
##      Accuracy : 0.5481
##      95% CI : (0.4474, 0.6459)
##      No Information Rate : 0.5769
##      P-Value [Acc > NIR] : 0.7571
##
##      Kappa : 0.0423
## McNemar's Test P-Value : 0.1447
##
##      Sensitivity : 0.7000
##      Specificity : 0.3409
##      Pos Pred Value : 0.5915
##      Neg Pred Value : 0.4545
##      Prevalence : 0.5769
##      Detection Rate : 0.4038
##      Detection Prevalence : 0.6827
##      Balanced Accuracy : 0.5205
##
##      'Positive' Class : FALSE
##
## [1] "Correlation GB unidirection prediction"
## [1] 0.006945995
## [1] "Long confusion matrix"
## Confusion Matrix and Statistics
##
##      Reference
## Prediction FALSE TRUE
##      FALSE      11   17
##      TRUE       34   42
##
##      Accuracy : 0.5096
##      95% CI : (0.4097, 0.609)
##      No Information Rate : 0.5673
##      P-Value [Acc > NIR] : 0.90048

```

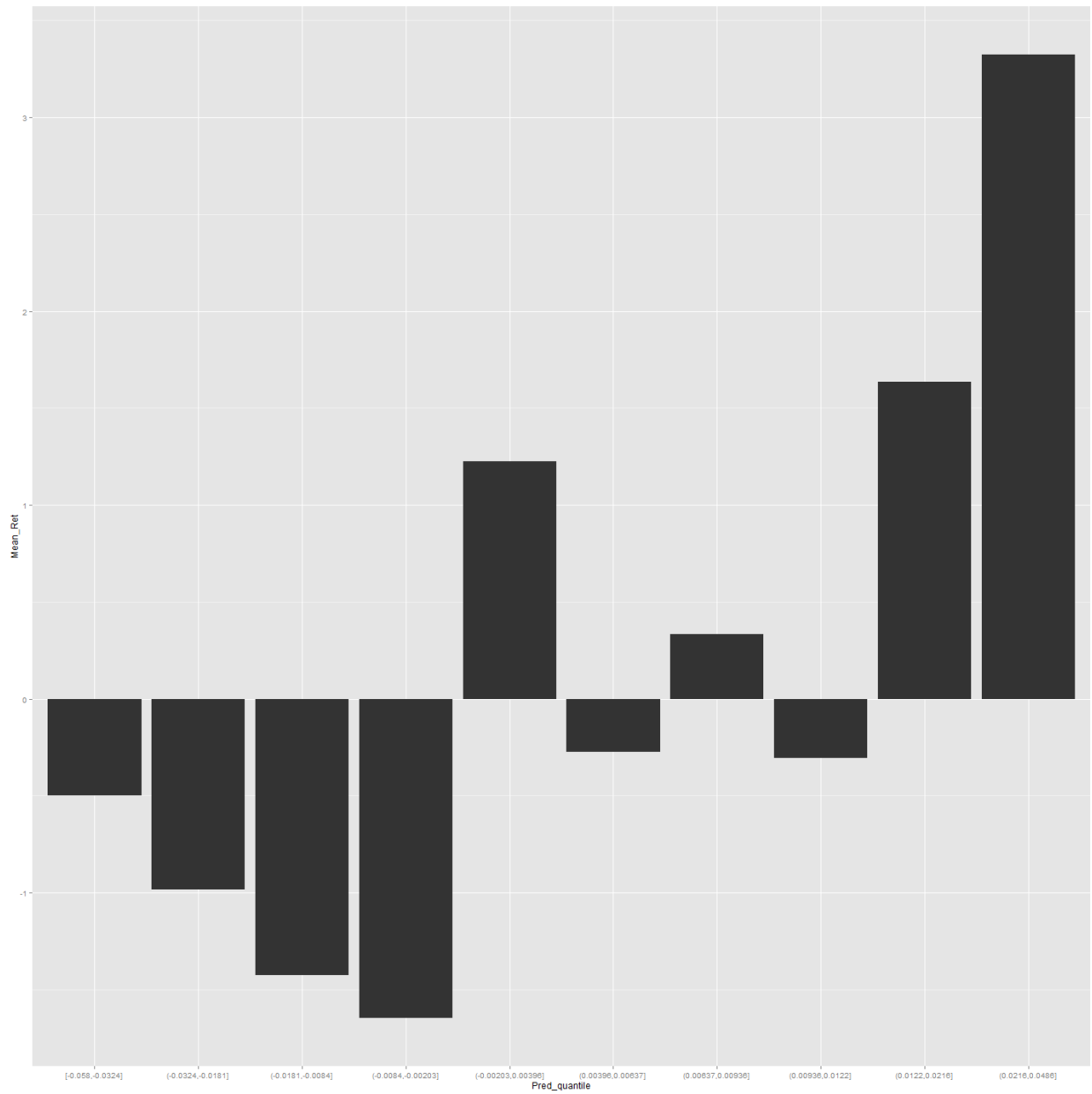


```

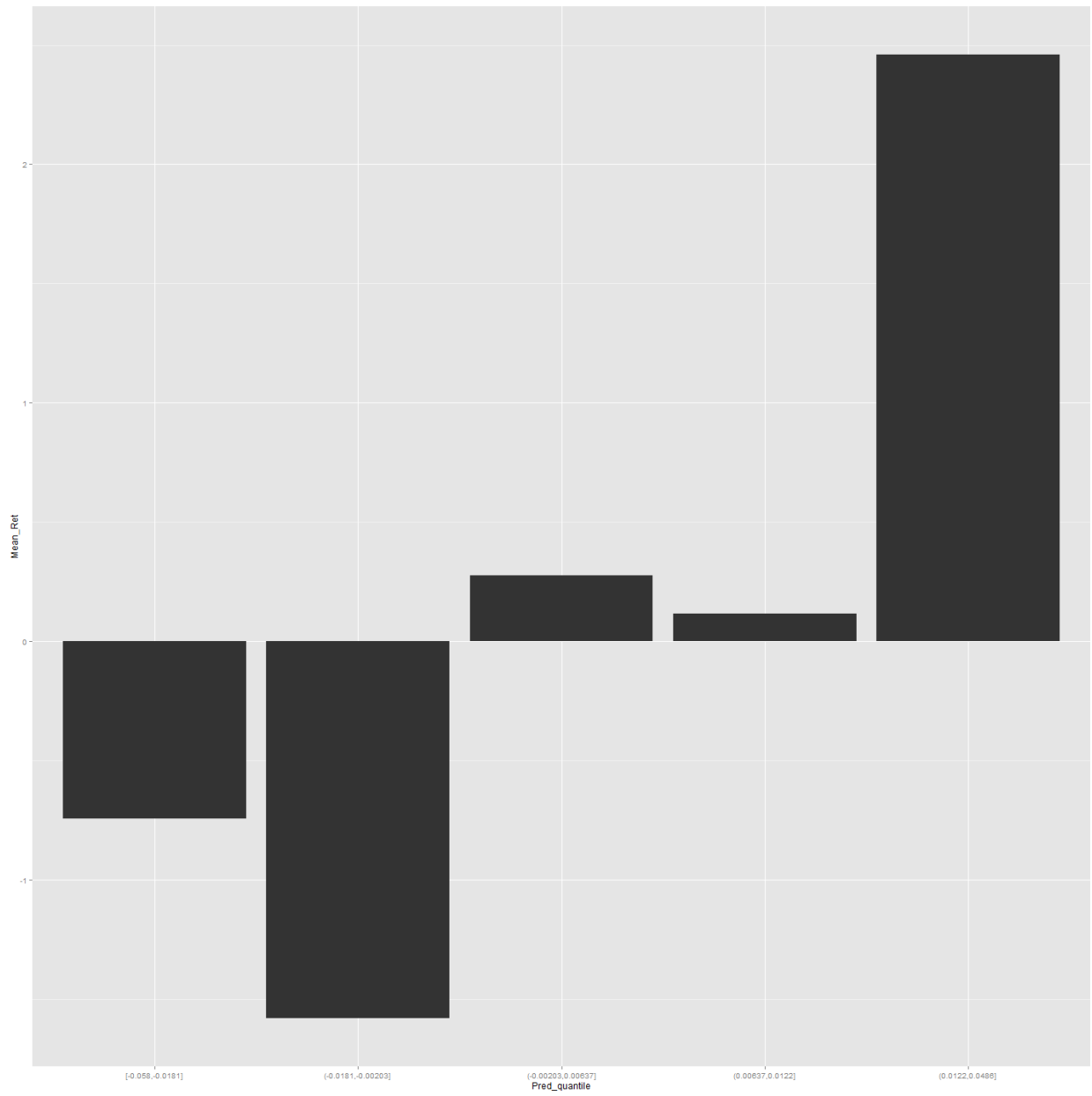
##
##           Kappa : -0.0457
## Mcnemar's Test P-Value : 0.02506
##
##           Sensitivity : 0.2444
##           Specificity : 0.7119
##           Pos Pred Value : 0.3929
##           Neg Pred Value : 0.5526
##           Prevalence : 0.4327
##           Detection Rate : 0.1058
##           Detection Prevalence : 0.2692
##           Balanced Accuracy : 0.4782
##
##           'Positive' Class : FALSE
##
## [1] "Short confusion matrix"
## Confusion Matrix and Statistics
##
##           Reference
## Prediction FALSE TRUE
##           FALSE      42      34
##           TRUE       17      11
##
##           Accuracy : 0.5096
##           95% CI : (0.4097, 0.609)
##           No Information Rate : 0.5673
##           P-Value [Acc > NIR] : 0.90048
##
##           Kappa : -0.0457
## Mcnemar's Test P-Value : 0.02506
##
##           Sensitivity : 0.7119
##           Specificity : 0.2444
##           Pos Pred Value : 0.5526
##           Neg Pred Value : 0.3929
##           Prevalence : 0.5673
##           Detection Rate : 0.4038
##           Detection Prevalence : 0.7308
##           Balanced Accuracy : 0.4782
##
##           'Positive' Class : FALSE
##
## [1] "Correlation JP unidirection prediction"
## [1] 0.2868498

## Warning: Stacking not well defined when ymin != 0

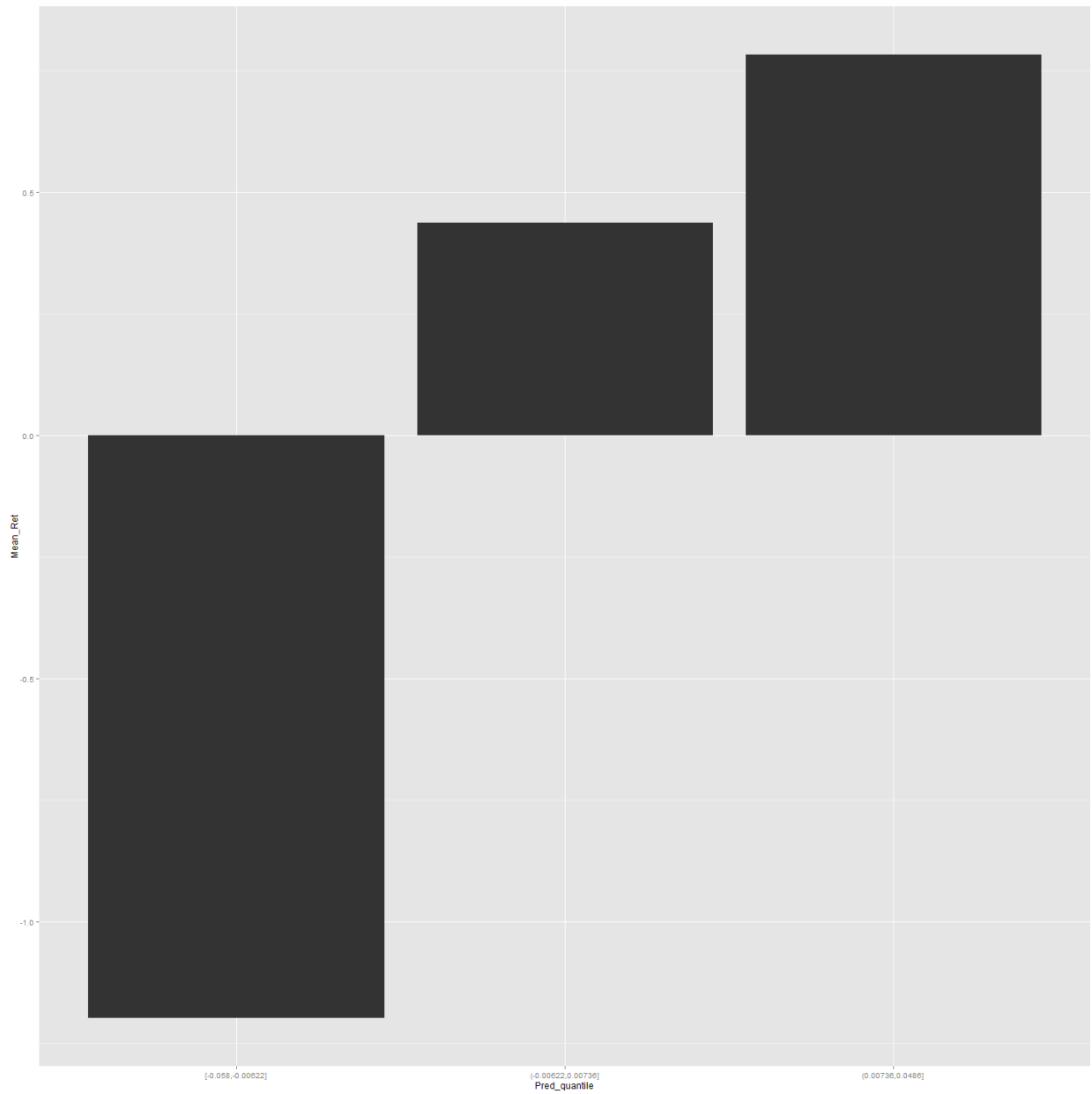
```



## Warning: Stacking not well defined when ymin != 0



## Warning: Stacking not well defined when ymin != 0



```
## [1] "Long confusion matrix"
## Confusion Matrix and Statistics
##
##           Reference
## Prediction FALSE TRUE
##      FALSE      28   18
##      TRUE       19   39
##
##           Accuracy : 0.6442
##           95% CI : (0.5443, 0.7357)
##      No Information Rate : 0.5481
##      P-Value [Acc > NIR] : 0.0298
##
##           Kappa : 0.2805
```

```

## McNemar's Test P-Value : 1.0000
##
##      Sensitivity : 0.5957
##      Specificity : 0.6842
##      Pos Pred Value : 0.6087
##      Neg Pred Value : 0.6724
##      Prevalence : 0.4519
##      Detection Rate : 0.2692
##      Detection Prevalence : 0.4423
##      Balanced Accuracy : 0.6400
##
##      'Positive' Class : FALSE
##
## [1] "Short confusion matrix"
## Confusion Matrix and Statistics
##
##      Reference
## Prediction FALSE TRUE
##      FALSE      38   20
##      TRUE       18   28
##
##      Accuracy : 0.6346
##      95% CI : (0.5345, 0.7269)
##      No Information Rate : 0.5385
##      P-Value [Acc > NIR] : 0.03013
##
##      Kappa : 0.2627
## McNemar's Test P-Value : 0.87113
##
##      Sensitivity : 0.6786
##      Specificity : 0.5833
##      Pos Pred Value : 0.6552
##      Neg Pred Value : 0.6087
##      Prevalence : 0.5385
##      Detection Rate : 0.3654
##      Detection Prevalence : 0.5577
##      Balanced Accuracy : 0.6310
##
##      'Positive' Class : FALSE
##

```

## Variable importance visualization over time

```

my_pairs = list(c("US","DE"),c("US","GB"),c("US","JP"))

my_threshold <- 100
for (my_pair in my_pairs){

  # SaveDataFrame(my_total_df,outputDataPathMonth,paste(my_pair[1], my_pair[2],"variable_importance_spr

  filename <- paste(my_pair[1], my_pair[2],"variable_importance_spread_results_month_2007.rds",sep="")
}

```

```

my_total_df <- tryCatch( readRDS(paste(outputDataPathStrategyMonth,filename, sep = "")),
                        error = function(e) {NULL})
if (!is.null(my_total_df)){

  # we only keep the first 100 most important predictors by importance count over time
  # my_filtered_df <- my_total_df[(sort(rowSums(my_total_df[, -1])), index.return=TRUE, decreasing = TR
  my_index <- rowSums(my_total_df[, -1])!=0
  my_total_df <- my_total_df[my_index,]
  my_filtered_df <- my_total_df[(sort(rowSums(my_total_df[, -1])!=0), index.return=TRUE, decreasing = T

  my_filtered_df$Feature <- as.factor(my_filtered_df$Feature)
  my_filtered_df <- transform(my_filtered_df, Feature=reorder(Feature,(sort(rowSums(my_filtered_df[, -1]
  my_filtered_df <- transform(my_filtered_df, Feature=reorder(Feature,rowSums(my_filtered_df[, -1])))

  my_filtered_df.m <- melt(my_filtered_df,id="Feature")
  colnames(my_filtered_df.m) <- c("Feature","Dates","Weight")
  # my_filtered_df.m <- ddply(my_filtered_df.m, .(variable), transform)
  # my_filtered_df.m <- ddply(my_filtered_df.m, .(variable), transform, rescale = rescale(value))
  p <- ggplot(my_filtered_df.m, aes(Dates, Feature)) + geom_tile(aes(fill = Weight),colour = "white")
    scale_fill_gradient(low = "white", high = "steelblue")
  print(p)
  # ExportPlot(p,outputDataPathStrategyMonth,paste0(my_pair[2],"heatmap"), width=10, height=15)
  print("Done")
}
}

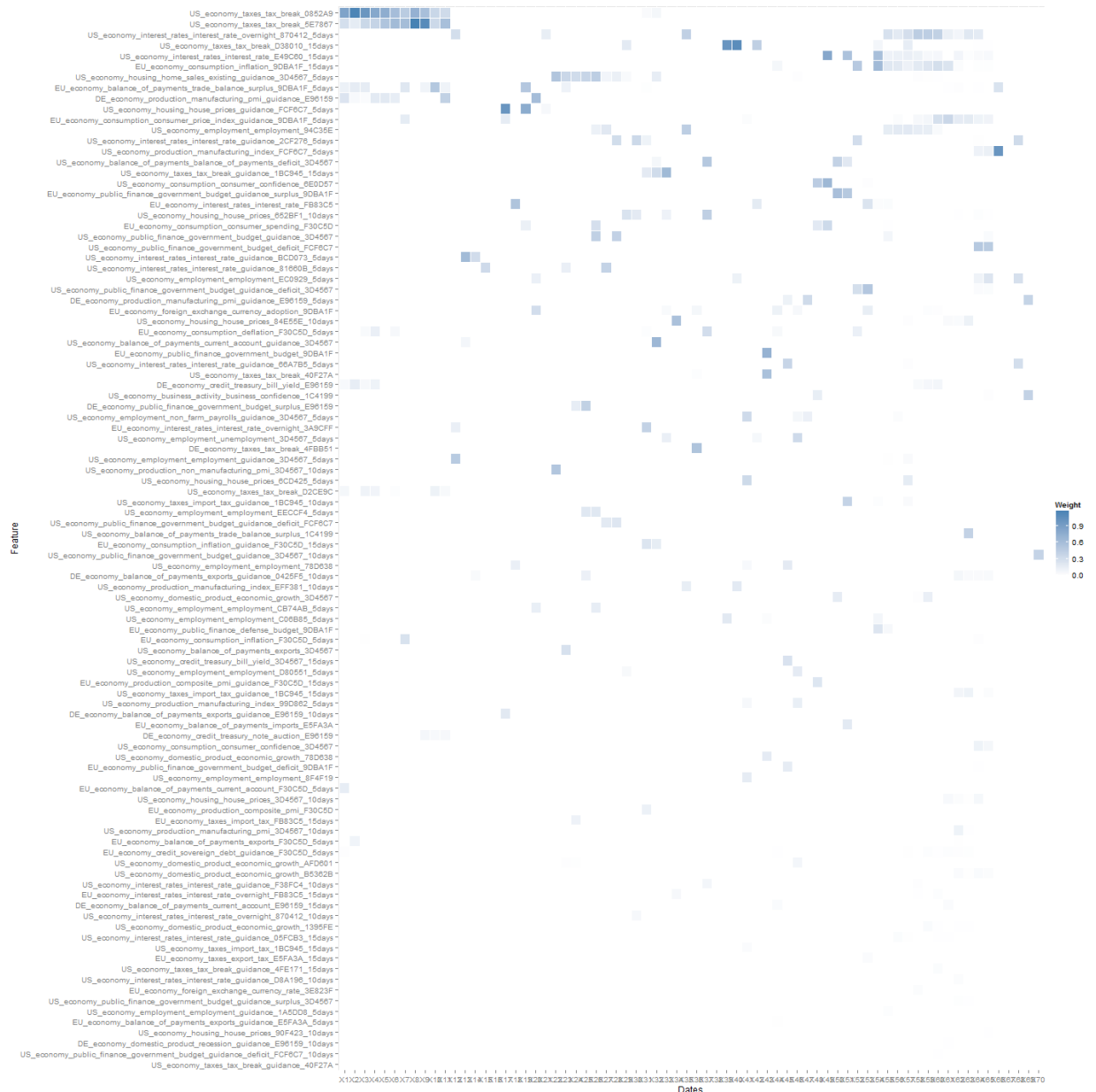
```

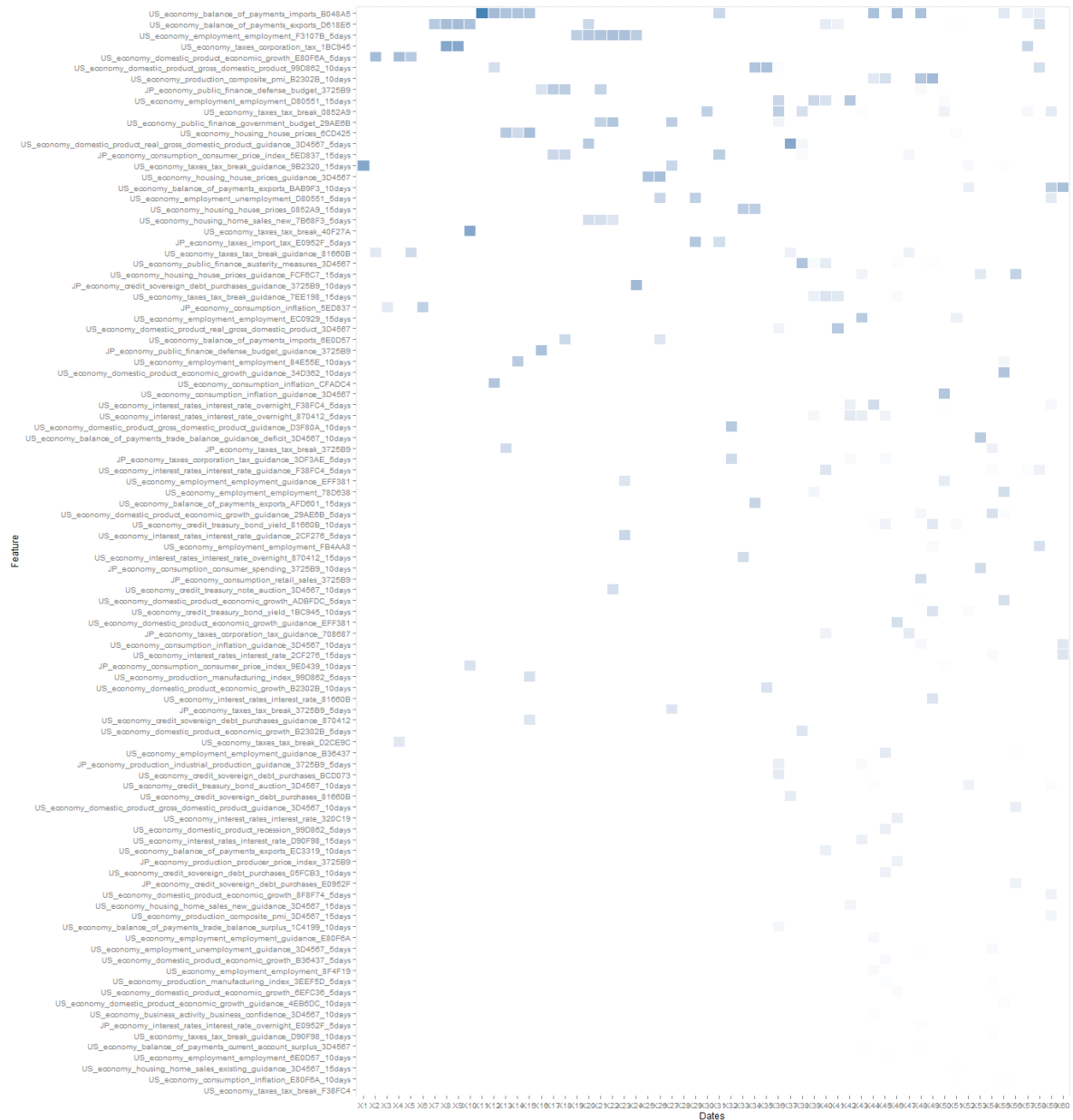
```
## [1] "Done"
```

```

## Warning in gzfile(file, "rb"): cannot open compressed
## file 'W:/OutputData/sduprey/NAR-271/Month_2007/
## USGBvariable_importance_spread_results_month_2007.rds', probable reason 'No
## such file or directory'

```





## [1] "Done"