

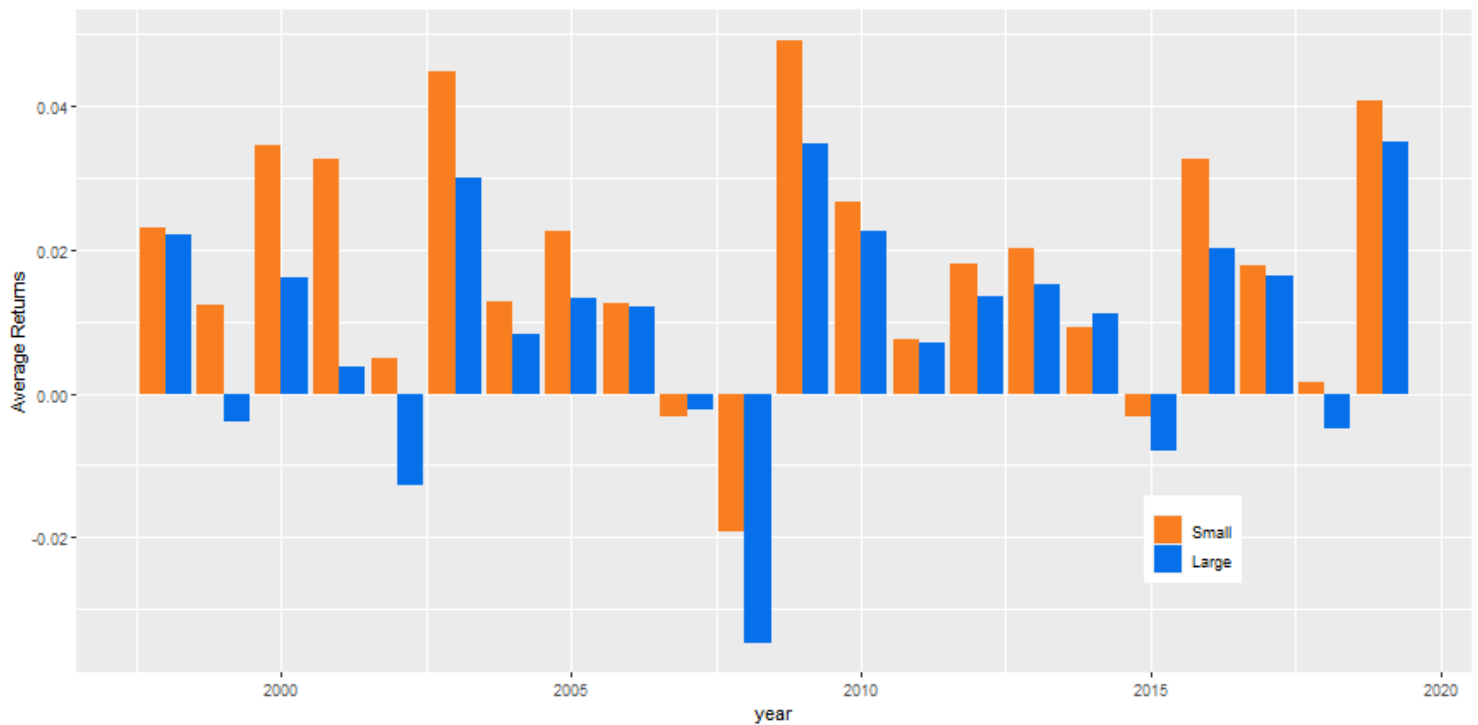
Datasets

```
load(here("Machine Learning for Factor Investing", "data_ml.RData"))
```

Example Factor, Size

```
data_ml %>%  
  group_by(date) %>% # group by date  
  mutate(large = Mkt_Cap_12M_Usd > median(Mkt_Cap_12M_Usd)) %>% # Creates the cap sort  
  ungroup() %>% # ungroup  
  mutate(year = lubridate::year(date)) %>% # Creates a year variable  
  group_by(year, large) %>% # Analyze by year & cap  
  summarize(avg_return = mean(R1M_Usd)) %>% # avg return by year & cap  
  ggplot(aes(x = year, y = avg_return, fill = large)) + # plot!  
  geom_col(position = "dodge") + # bars side-to-side  
  theme(legend.position = c(0.8, 0.2)) + # legend location  
  coord_fixed(124) + # x/y aspect ration  
  theme(legend.title = element_blank()) +  
  scale_fill_manual(values = c("#F87E1F", "#0570EA"), name = "", # colors  
                    labels = c("Small", "Large")) +  
  ylab("Average Returns") +  
  theme(legend.text = element_text(size=9))
```

``summarise()`` regrouping output by 'year' (override with `` .groups `` argument)



Factors

Size

SMB = small firms minus large firms

Value

HM = high minus low: undervalued minus 'growth' firms

Momentum

WML winners minus losers

Profitability

RMW = robust minus weak profits

profitability is measured as (revenues - (cost and expenses)) / equity

Investment

CMA conservative minus aggressive

Investment is measured via the growth of total assets (divided by total assets).

Low 'risk'

BAB betting against beta

(simple vol, market beta, idiosyncratic vol, etc)

Kenneth French Factor Library

Example Factor Model

```
min_date <- "1963-07-31"; max_date <- "2020-06-30"

temp <- tempfile()

KF_website <- "http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/"
KF_file <- "ftp/F-F_Research_Data_5_Factors_2x3_CSV.zip"

link <- paste0(KF_website, KF_file)

download.file(link, temp, quiet = T)

FF_factors <- read_csv(unz(temp, "F-F_Research_Data_5_Factors_2x3.CSV"),
                      skip = 3) %>% # Check the number of lines to skip!
  rename(date = X1, MKT_RF = `Mkt-RF`) %>% # Change the name of first columns
  mutate_at(vars(-date), as.numeric) %>% # Convert values to number
  mutate(date = ymd(parse_date_time(date, "%Y%m"))) %>% # Date in right format
  mutate(date = rollback(date + months(1))) # End of month date
```

Warning: Missing column names filled in: 'X1' [1]

```
-- Column specification -----
cols(
  X1 = col_character(),
  `Mkt-RF` = col_character(),
```

```
SMB = col_character(),
HML = col_character(),
RMW = col_character(),
CMA = col_character(),
RF = col_character()
)

Warning: 1 parsing failure.
row col expected actual file
693 -- 7 columns 1 columns <connection>

Warning: Problem with `mutate()` input `MKT_RF`.
i NAs introduced by coercion
i Input `MKT_RF` is `.`.Primitive("as.double")(MKT_RF)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `SMB`.
i NAs introduced by coercion
i Input `SMB` is `.`.Primitive("as.double")(SMB)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `HML`.
i NAs introduced by coercion
i Input `HML` is `.`.Primitive("as.double")(HML)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `RMW`.
i NAs introduced by coercion
i Input `RMW` is `.`.Primitive("as.double")(RMW)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `CMA`.
i NAs introduced by coercion
i Input `CMA` is `.`.Primitive("as.double")(CMA)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `RF`.
i NAs introduced by coercion
i Input `RF` is `.`.Primitive("as.double")(RF)`.

Warning in mask$eval_all_mutate(dots[[i]]): NAs introduced by coercion

Warning: Problem with `mutate()` input `date`.
i 58 failed to parse.
i Input `date` is `ymd(parse_date_time(date, "%Y%m"))`.

Warning: 58 failed to parse.
```

Table 1: Sample of Monthly Factor Returns.

date	MKT_RF	SMB	HML	RMW	CMA	RF
1963-07-31	-0.0039	-0.0045	-0.0094	0.0066	-0.0115	0.0027
1963-08-31	0.0507	-0.0082	0.0182	0.0040	-0.0040	0.0025
1963-09-30	-0.0157	-0.0048	0.0017	-0.0076	0.0024	0.0027
1963-10-31	0.0253	-0.0130	-0.0004	0.0275	-0.0224	0.0029
1963-11-30	-0.0085	-0.0085	0.0170	-0.0045	0.0222	0.0027
1963-12-31	0.0183	-0.0190	-0.0006	0.0007	-0.0030	0.0029

```
FF_factors <- FF_factors %>% mutate(MKT_RF = MKT_RF / 100,
                                   SMB = SMB / 100,
                                   HML = HML / 100,
                                   RMW = RMW / 100,
                                   CMA = CMA / 100,
                                   RF = RF / 100) %>%
  filter(date >= min_date, date <= max_date)
```

```
knitr::kable(head(FF_factors), booktabs = T,
              caption = "Sample of Monthly Factor Returns.")
```

```
FF_Avg_Returns <- FF_factors %>%
  mutate(date = year(date)) %>%
  gather(key = factor, value = value, - date) %>%
  group_by(date, factor) %>%
  summarise(value = mean(value))
```

`summarise()` regrouping output by 'date' (override with `.groups` argument)

```
FF_Avg_Returns %>%
  ggplot(aes(x = date, y = value, color = factor)) +
  geom_line() + coord_fixed(500)
```

```
FF_factors %>%
  gather(key = factor, value = return, - date) %>%
  filter(factor != 'RF') %>%
  ggplot(aes(return, group = factor)) +
  geom_density(aes(fill = factor, alpha = .25))
```

```
FF_Avg_Returns %>%
  filter(factor != 'RF') %>%
  ggplot(aes(value, group = factor)) +
  geom_density(aes(fill = factor, alpha = .25))
```

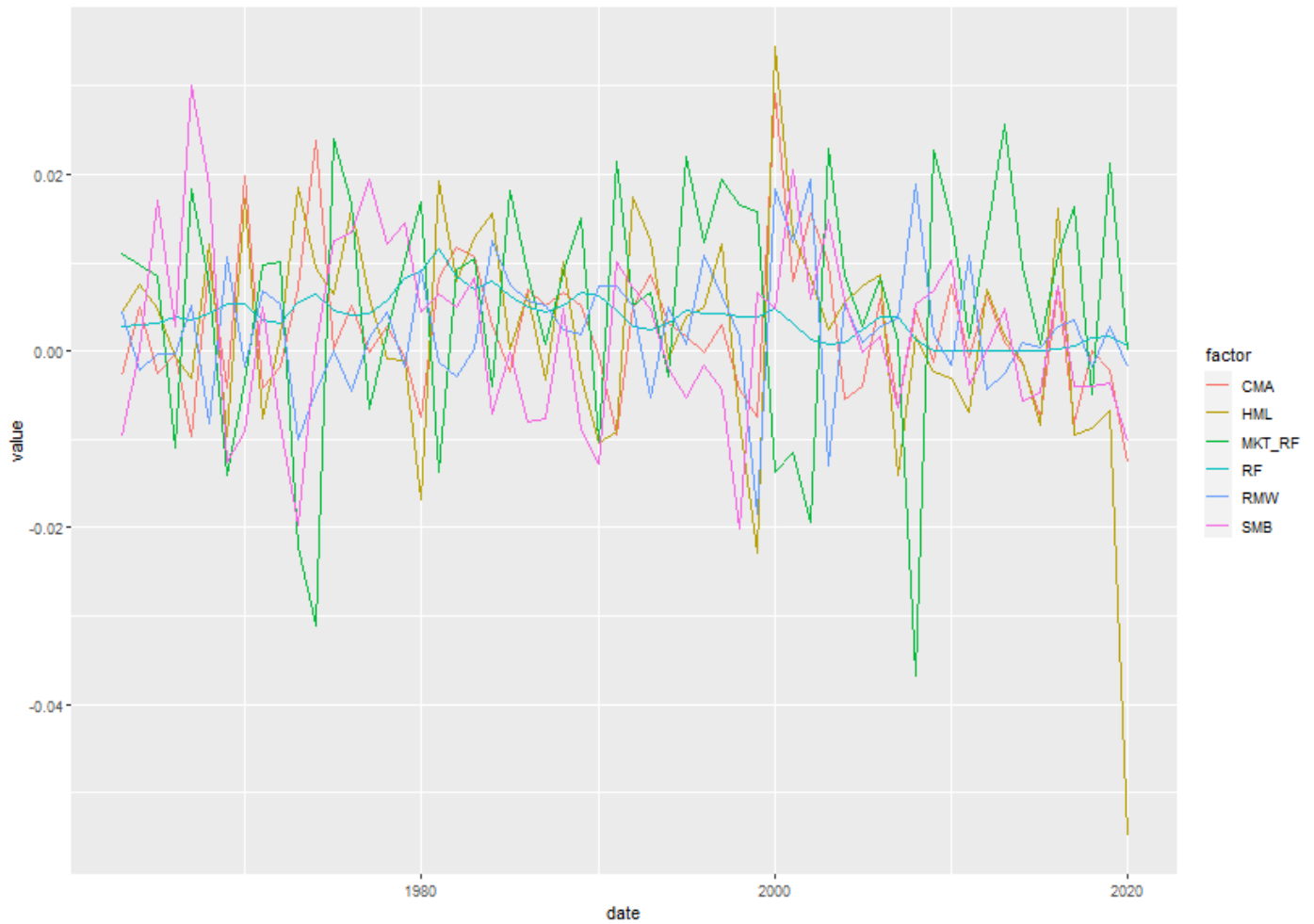


Figure 1: Factor Returns

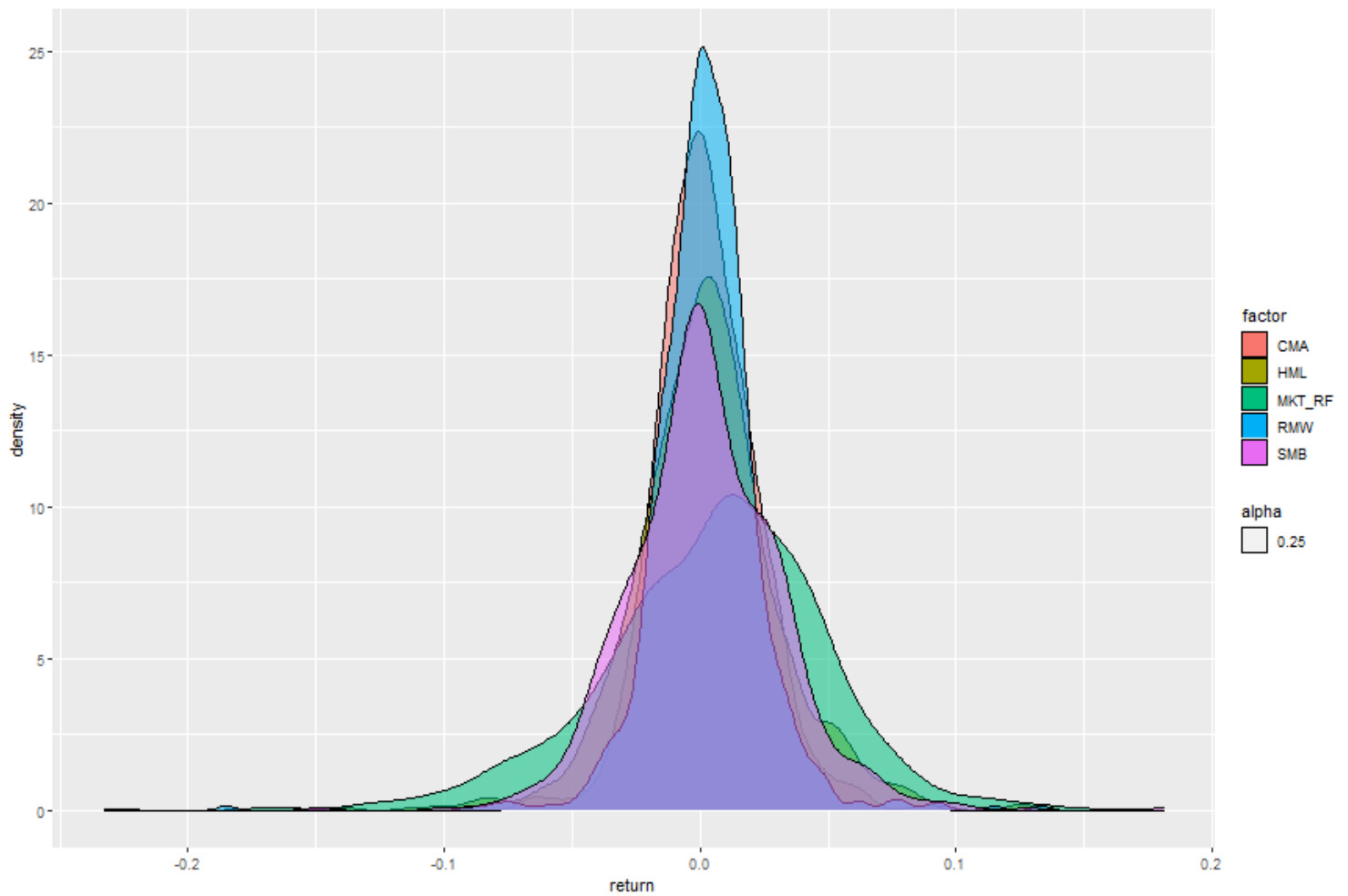


Figure 2: Return Densities

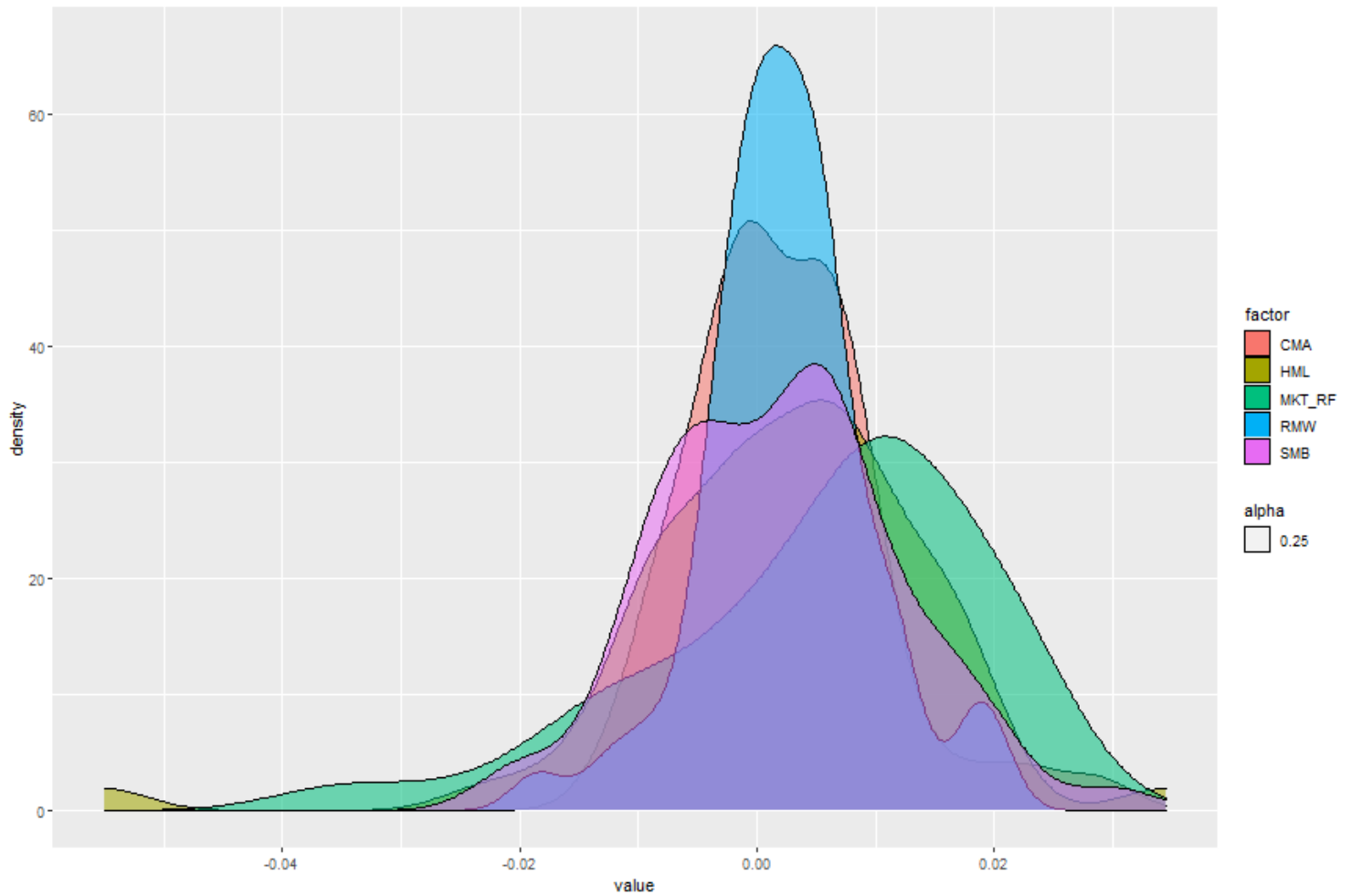


Figure 3: Yearly Avg Return Densities


```
FF_Cum>Returns <- FF_factors %>%  
  gather(key = factor, value = value, -date) %>%  
  group_by(factor) %>%  
  mutate(lag_ret = lag(value)) %>%  
  mutate(return = cumprod(1 + ifelse(is.na(lag_ret), 0, lag_ret)))  
  
FF_Cum>Returns %>%  
  ggplot(aes(date, return, group = factor)) +  
  geom_line(aes(col = factor))
```

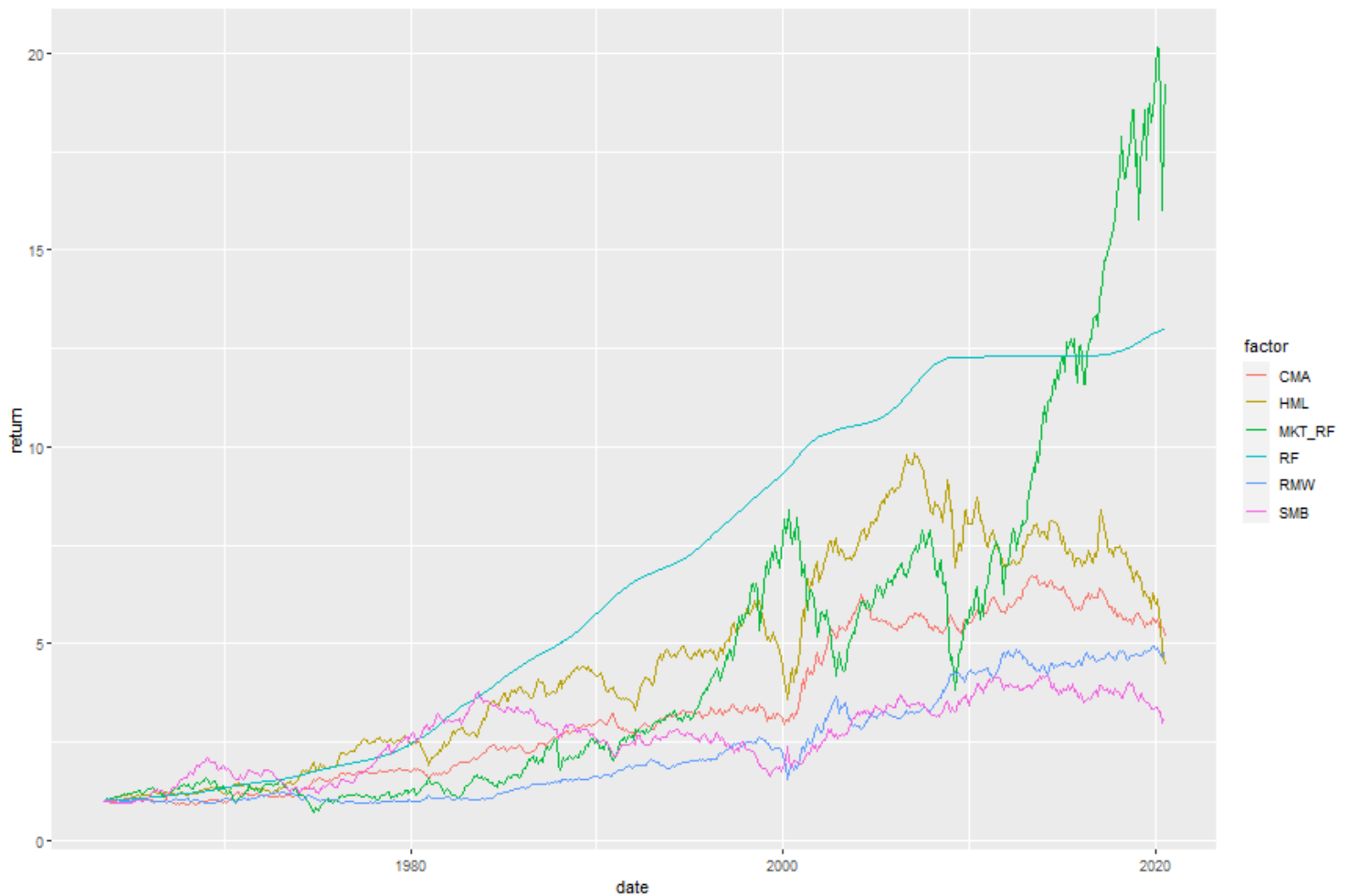


Figure 4: Growth of \$1 by factor

Fama-Macbeth Regressions

```
separation_date <- as_date("2014-01-15")

training_sample <- filter(data_ml, date < separation_date)
testing_sample <- filter(data_ml, date > separation_date)

stock_ids <- levels(as.factor(data_ml$stock_id)) # list of all stock ids

stock_days <- data_ml %>%
  group_by(stock_id) %>%
  summarise(nb = n())

`summarise()` ungrouping output (override with `.groups` argument)

stock_ids_short <- stock_ids[which(stock_days$nb == max(stock_days$nb))] # keep only stocks with max days

# single stock example

stock_idenfier <- 3

data_ml %>%
  filter(date == as_date("2006-06-30") & stock_id == 3)

# A tibble: 1 x 99
  stock_id date      Advt_12M_Usd Advt_3M_Usd Advt_6M_Usd Asset_Turnover Bb_Yld
    <int> <date>          <dbl>      <dbl>      <dbl>          <dbl> <dbl>
1       3 2006-06-30      0.08        0.08        0.09          0.04  0.78
# ... with 92 more variables: Bv <dbl>, Capex_Ps_Cf <dbl>, Capex_Sales <dbl>,
# Cash_Div_Cf <dbl>, Cash_Per_Share <dbl>, Cf_Sales <dbl>, Debtequity <dbl>,
# Div_Yld <dbl>, Dps <dbl>, Ebit_Bv <dbl>, Ebit_NoA <dbl>, Ebit_Oa <dbl>,
# Ebit-Ta <dbl>, Ebitda_Margin <dbl>, Eps <dbl>, Eps_Basic <dbl>,
# Eps_Basic_Gr <dbl>, Eps_Contin_Oper <dbl>, Eps_Dil <dbl>, Ev <dbl>,
# Ev_Ebitda <dbl>, Fa_Ci <dbl>, Fcf <dbl>, Fcf_Bv <dbl>, Fcf_Ce <dbl>,
# Fcf_Margin <dbl>, Fcf_NoA <dbl>, Fcf_Oa <dbl>, Fcf-Ta <dbl>, Fcf_Tbv <dbl>,
# Fcf_ToA <dbl>, Fcf_Yld <dbl>, Free_Ps_Cf <dbl>, Int_Rev <dbl>,
# Interest_Expense <dbl>, Mkt_Cap_12M_Usd <dbl>, Mkt_Cap_3M_Usd <dbl>,
# Mkt_Cap_6M_Usd <dbl>, Mom_11M_Usd <dbl>, Mom_5M_Usd <dbl>,
# Mom_Sharp_11M_Usd <dbl>, Mom_Sharp_5M_Usd <dbl>, Nd_Ebitda <dbl>,
# Net_Debt <dbl>, Net_Debt_Cf <dbl>, Net_Margin <dbl>, Netdebtyield <dbl>,
# Ni <dbl>, Ni_Avail_Margin <dbl>, Ni_Oa <dbl>, Ni_ToA <dbl>, Noa <dbl>,
# Oa <dbl>, Ocf <dbl>, Ocf_Bv <dbl>, Ocf_Ce <dbl>, Ocf_Margin <dbl>,
# Ocf_NoA <dbl>, Ocf_Oa <dbl>, Ocf-Ta <dbl>, Ocf_Tbv <dbl>, Ocf_ToA <dbl>,
# Op_Margin <dbl>, Op_Prt_Margin <dbl>, Oper_Ps_Net_Cf <dbl>, Pb <dbl>,
# Pe <dbl>, Ptx_Mgn <dbl>, Recurring_Earning_Total_Assets <dbl>,
# Return_On_Capital <dbl>, Rev <dbl>, Roa <dbl>, Roc <dbl>, Roce <dbl>,
```

```
# Roe <dbl>, Sales_Ps <dbl>, Share_Turn_12M <dbl>, Share_Turn_3M <dbl>,
# Share_Turn_6M <dbl>, Ta <dbl>, Tev_Less_Mktcap <dbl>, Tot_Debt_Rev <dbl>,
# Total_Capital <dbl>, Total_Debt <dbl>, Total_Debt_Capital <dbl>,
# Total_Liabilities_Total_Assets <dbl>, Vol1Y_Usd <dbl>, Vol3Y_Usd <dbl>,
# R1M_Usd <dbl>, R3M_Usd <dbl>, R6M_Usd <dbl>, R12M_Usd <dbl>
```

```
stock_returns <- data_ml %>%
  filter(stock_id == stock_idenfier) %>%
  select(date, stock_id, Return = R1M_Usd) %>%
  group_by(stock_id) %>%
  mutate(Return = lag(Return)) %>%
  ungroup()
```

```
stock_returns %>%
  filter(date == as_date("2006-06-30") & stock_id == 3)
```

```
# A tibble: 1 x 3
  date      stock_id Return
<date>      <int>   <dbl>
1 2006-06-30         3     NA
```

```
factor_data <- left_join(stock_returns, FF_factors, by = "date") %>%
  select(date, stock_id, MKT_RF, SMB, HML, RMW, CMA, RF, Return)
```

```
factor_loading <-
  coef(summary(lm(formula = "Return ~ MKT_RF + SMB + HML + RMW + CMA", data = factor_data))) %>%
  as.data.frame() %>%
  select(Value = Estimate) %>%
  rownames_to_column("Factor") %>%
  mutate(stock_id = stock_idenfier) %>%
  spread(key = "Factor", value = "Value") %>%
  select(MKT_RF, SMB, HML, RMW, CMA)
```

```
factor_data <- cbind(factor_loading, stock_returns) %>%
  filter(!is.na(Return))
```

```
nb_factors <- 5 # Number of factors
```

```
data_FM <- left_join(data_ml %>% # Join the 2 datasets
  dplyr::select(date, stock_id, R1M_Usd) %>% # (with returns...
  filter(stock_id %in% stock_ids_short), # ... over some stocks)
  FF_factors,
  by = "date") %>%
  group_by(stock_id) %>% # Grouping
  mutate(R1M_Usd = lag(R1M_Usd)) %>% # Lag returns
  ungroup() %>%
```

Table 2: Coefficients

	Constant	MKT_RF	SMB	HML	RMW	CMA
3	-0.0017438	0.8092717	0.8280240	0.8479085	0.1198440	-0.2522913
4	0.0037247	0.3073030	0.2619254	-0.1392922	0.4354316	0.4485567
7	0.0050755	0.5203728	0.5246247	0.0376542	0.3136473	0.3198131
9	0.0044285	0.7516452	0.6174593	1.0164648	-0.0597775	-0.0562163
16	0.0010675	1.1996284	-0.1769292	1.3980331	0.1910664	-0.6164365
22	0.0019074	0.5925792	0.5670595	0.3446145	0.4738824	0.1648655

```

na.omit() %>% # Remove missing points
spread(key = stock_id, value = R1M_Usd)

models <- lapply(paste0("`", stock_ids_short,
  "` ~ MKT_RF + SMB + HML + RMW + CMA'"),
  function(f){ lm(as.formula(f), data = data_FM,
    na.action="na.exclude") %>%
    summary() %>% # Gather the output
    "$"(coef) %>% # Keep only coefs
    data.frame() %>% # Convert to dataframe
    dplyr::select(Estimate)}
  )

betas <- matrix(unlist(models), ncol = nb_factors + 1, byrow = T) %>% # Extract the betas
  data.frame(row.names = stock_ids_short) # Format: row names

stopifnot(nrow(betas) == length(stock_ids_short))

colnames(betas) <- c("Constant", "MKT_RF", "SMB", "HML", "RMW", "CMA") # Format: col names

knitr::kable(head(betas), caption = "Coefficients")

factor_loadings <- betas %>%
  dplyr::select(-Constant) %>%
  data.frame()

stock_returns <- data_FM %>%
  dplyr::select(-MKT_RF, -SMB, -HML, -RMW, -CMA, -RF)

factor_returns <- stock_returns %>%
  dplyr::select(-date) %>%
  data.frame(row.names = stock_returns$date) %>%
  t()

```

```

stopifnot(nrow(factor_returns) == nrow(factor_loadings))

FM_data <- cbind(factor_loadings, factor_returns)

models <- lapply(paste("~",
                      stock_returns$date, "~",
                      ' ~ MKT_RF + SMB + HML + RMW + CMA', sep = ""),
                function(f){ lm(as.formula(f), data = FM_data) %>% # Call lm(.)
                              summary() %>% # Gather the ou
                              "$"(coef) %>% # Keep only the
                              data.frame() %>% # Convert to da
                              dplyr::select(Estimate)} # Keep only est

)

gammas <- matrix(unlist(models), ncol = nb_factors + 1, byrow = T) %>% # Switch to datafram
  data.frame(row.names = stock_returns$date) # & set row na
colnames(gammas) <- c("Constant", "MKT_RF", "SMB", "HML", "RMW", "CMA") # Set col names

gammas %>% # Take gammas:
  # The first row is omitted because the first row of returns is undefined
  dplyr::select(MKT_RF, SMB, HML) %>% # Select 3 factors
  bind_cols(date = data_FM$date) %>% # Add date
  gather(key = factor, value = gamma, -date) %>% # Put in tidy shape
  ggplot(aes(x = date, y = gamma, color = factor)) + # Plot
  geom_line() + facet_grid( factor~. ) + # Lines & facets
  scale_color_manual(values=c("#F87E1F", "#0570EA", "#F81F40")) # Colors

```

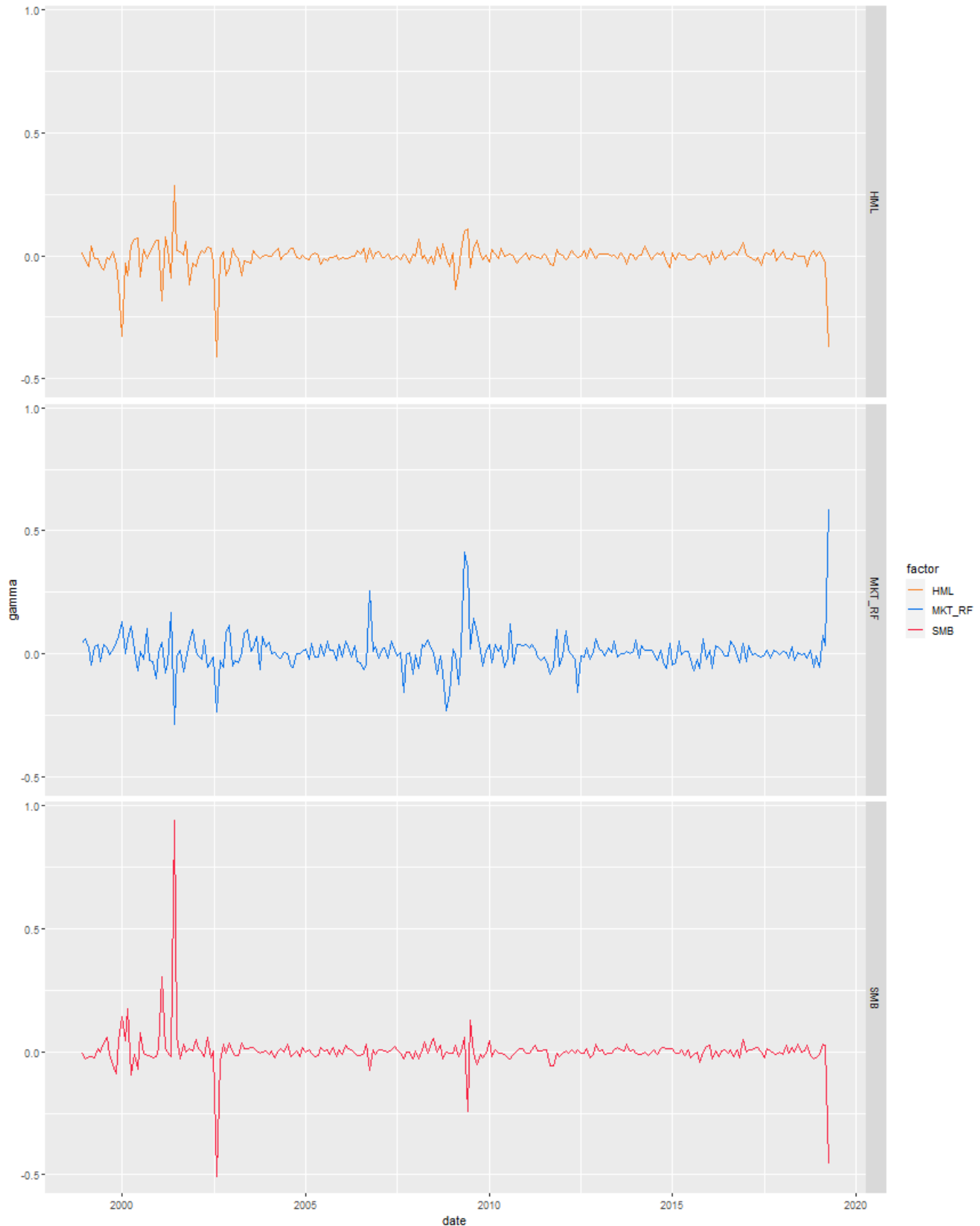


Figure 5: Gammas

Factor Competition

```
factors <- c("MKT_RF", "SMB", "HML", "RMW", "CMA")

models <- lapply(paste(factors, ' ~ MKT_RF + SMB + HML + RMW + CMA-', factors),
  function(f){ lm(as.formula(f), data = FF_factors) %>%
    summary() %>% # Call lm(.)
    "$"(coef) %>% # Gather the output
    data.frame() %>% # Keep only the coefs
    filter(rownames(.) == "(Intercept)") %>% # Convert to dataframe
    dplyr::select(Estimate, `Pr...t...`) %>% # Keep only the Intercept
    ) # Keep the coef & p-value
alphas <- matrix(unlist(models), ncol = 2, byrow = T) %>% # Switch from list to dataframe
  data.frame(row.names = factors)
# alphas # To see the alphas (optional)

results <- matrix(NA, nrow = length(factors), ncol = length(factors) + 1) # Coefs
signif <- matrix(NA, nrow = length(factors), ncol = length(factors) + 1) # p-values

for(j in 1:length(factors)){
  form <- paste(factors[j],
    ' ~ MKT_RF + SMB + HML + RMW + CMA-', factors[j]) # Build model
  fit <- lm(form, data = FF_factors) %>% summary() # Estimate model
  coef <- fit$coefficients[,1] # Keep coefficient
  p_val <- fit$coefficients[,4] # Keep p-values
  results[j, -(j+1)] <- coef # Fill matrix
  signif[j, -(j+1)] <- p_val
}

signif[is.na(signif)] <- 1 # Kick out NAs

results <- results %>% round(3) %>% data.frame() # Basic formatting

results[signif<0.001] <- paste(results[signif<0.001], " (***)") # 3 star signif

results[signif>0.001&signif<0.01] <- # 2 star signif
  paste(results[signif>0.001&signif<0.01], " (**)")

results[signif>0.01&signif<0.05] <- # 1 star signif
  paste(results[signif>0.01&signif<0.05], " (*)")

results <- cbind(factors, results) # Add dep. variable

colnames(results) <- c("Dep. Variable", "Intercept", factors) # Add column names
```

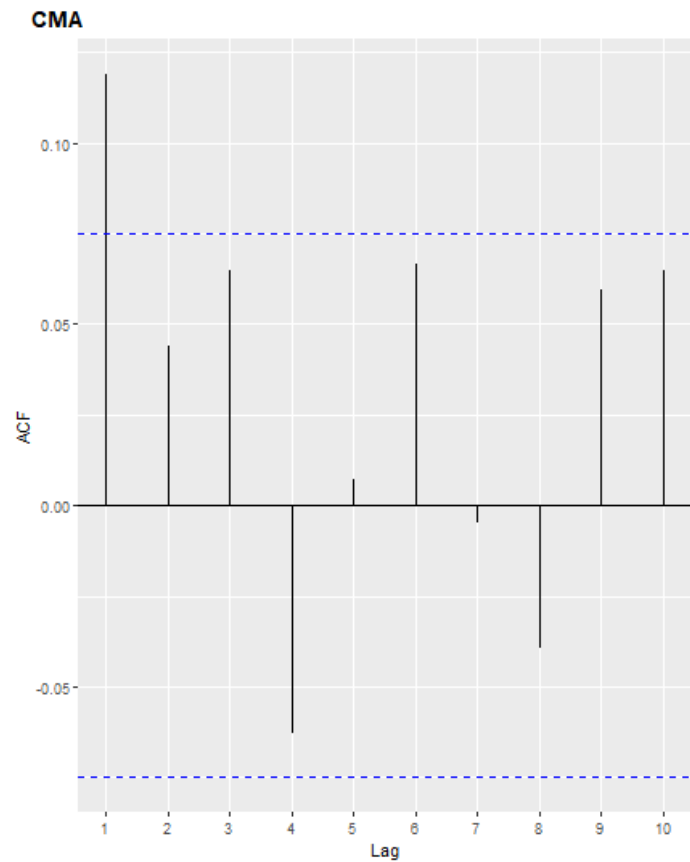
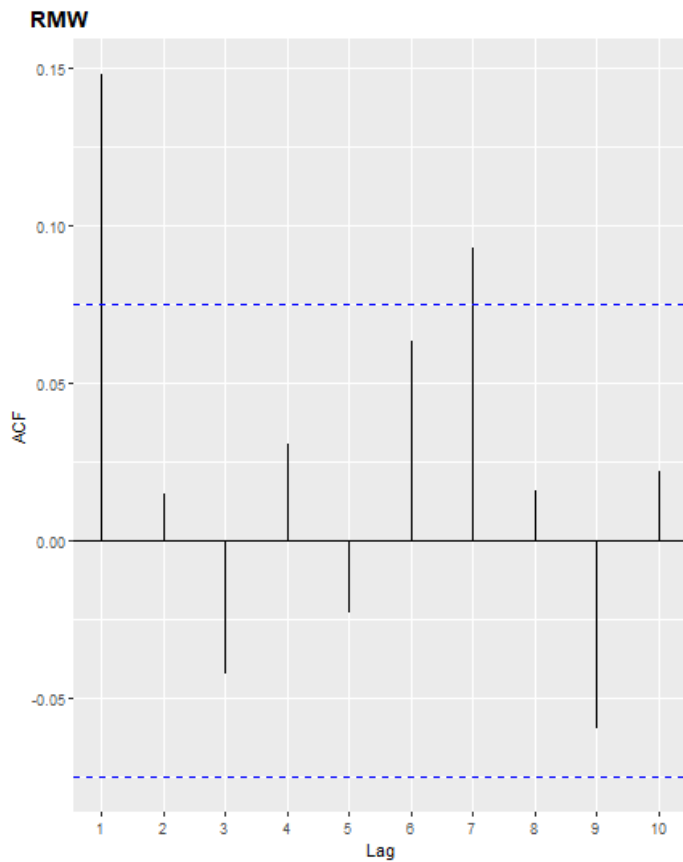
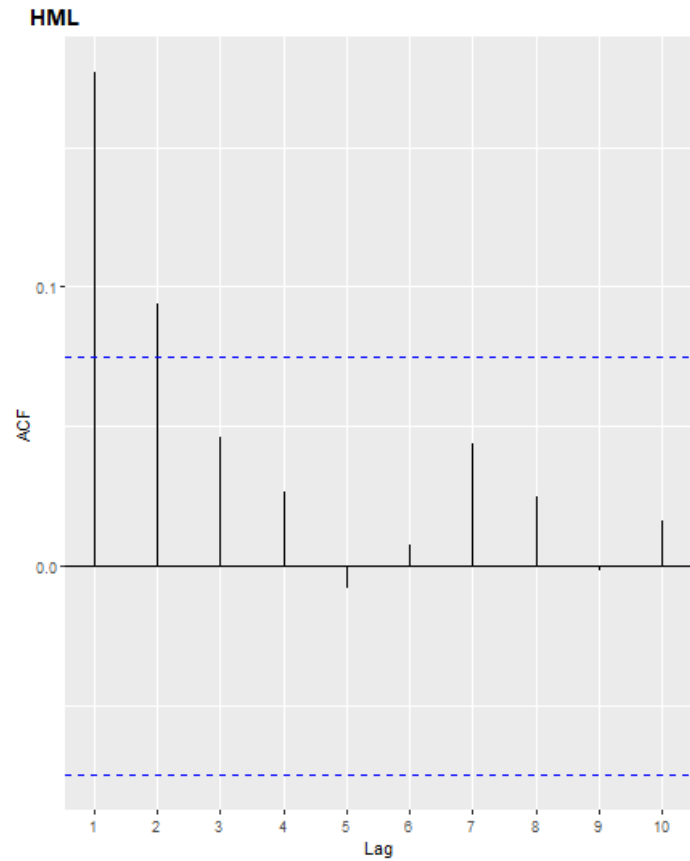
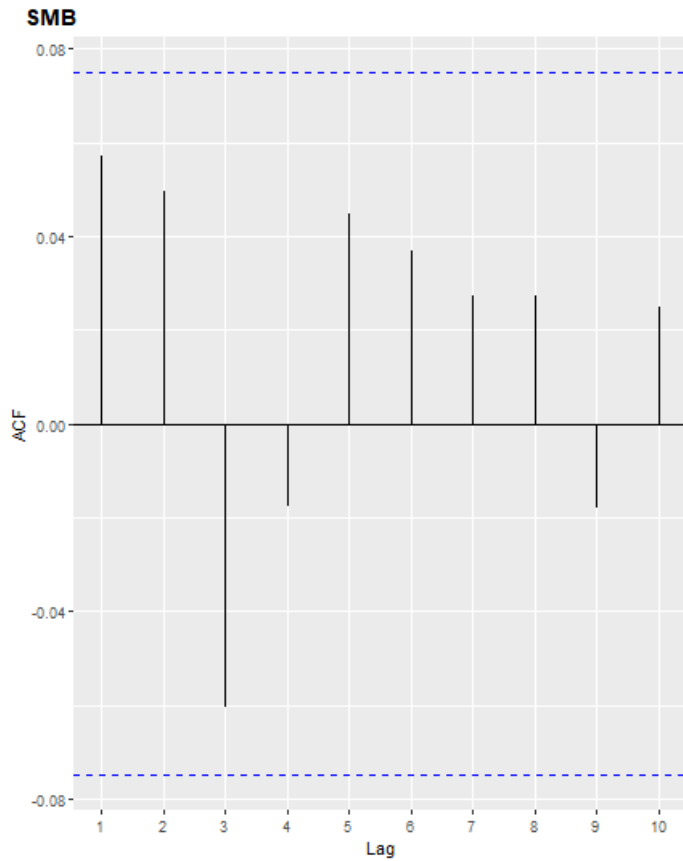
Table 3: Factor competition among the Fama and French (2015) five factors.

Dep. Variable	Intercept	MKT_RF	SMB	HML	RMW	CMA
MKT_RF	0.008 (***)	NA	0.287 (***)	0.143 (*)	-0.326 (***)	-0.951 (***)
SMB	0.003 (*)	0.143 (***)	NA	0.104 (*)	-0.423 (***)	-0.149
HML	-0.001	0.04 (*)	0.059 (*)	NA	0.172 (***)	1.027 (***)
RMW	0.004 (***)	-0.084 (***)	-0.22 (***)	0.158 (***)	NA	-0.286 (***)
CMA	0.003 (***)	-0.115 (***)	-0.036	0.441 (***)	-0.133 (***)	NA

Momentum, timing and ESG

```
acf_SMB <- ggAcf(FF_factors$SMB, lag.max = 10) + labs(title = "")
acf_HML <- ggAcf(FF_factors$HML, lag.max = 10) + labs(title = "")
acf_RMW <- ggAcf(FF_factors$RMW, lag.max = 10) + labs(title = "")
acf_CMA <- ggAcf(FF_factors$CMA, lag.max = 10) + labs(title = "")

plot_grid(acf_SMB, acf_HML, acf_RMW, acf_CMA, # plot
          labels = c("SMB", "HML", "RMW", "CMA"))
```

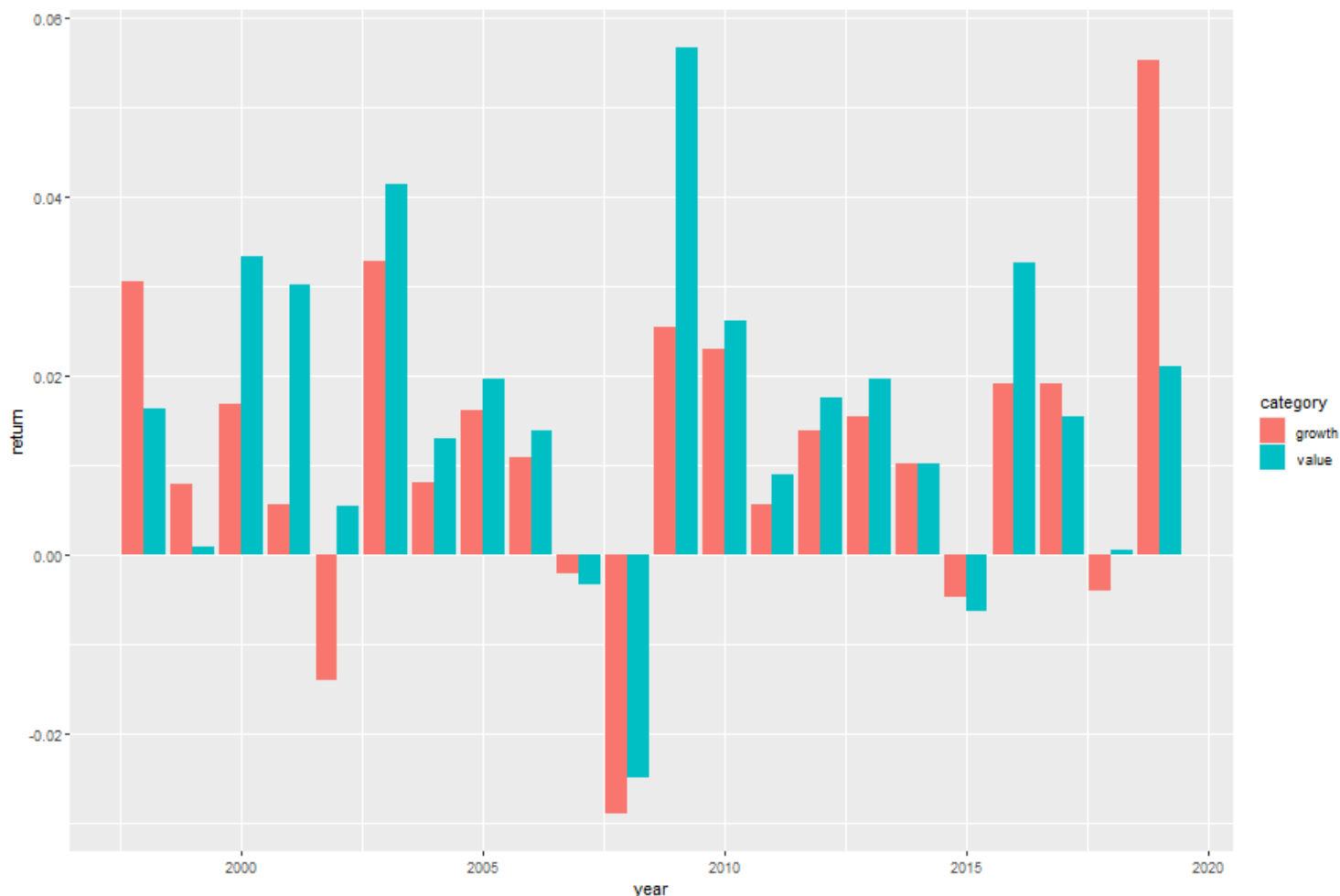



Exercises

1.) Compute annual returns of the growth vs value portfolios, that is, the average return of firms with above median price-to-book ratio (*Pb*).

```
data_ml %>%
  group_by(date) %>%
  mutate(category = ifelse(Pb > median(Pb), "growth", "value")) %>%
  ungroup() %>%
  mutate(year = lubridate::year(date)) %>%
  group_by(year, category) %>%
  summarise(return = mean(R1M_Usd)) %>%
  ggplot(aes(year, return, fill = category)) +
  geom_col(position = "dodge")
```

`summarise()` regrouping output by 'year' (override with `.groups` argument)



2.) Same exercise, but compare the monthly returns and plot the value (through time) of the corresponding portfolios.

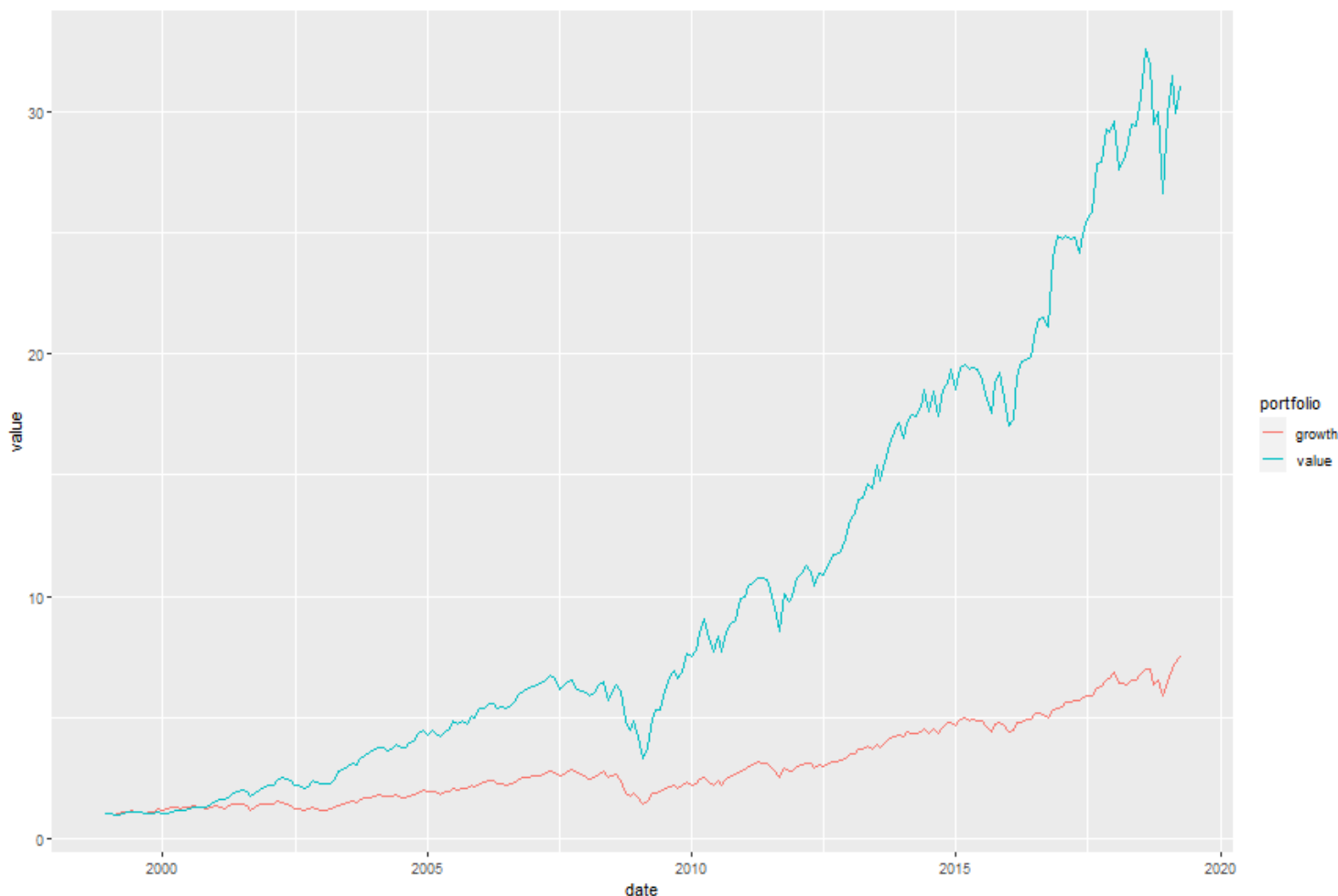
```
monthly_return <- data_ml %>%
  group_by(date) %>%
  mutate(growth = Pb > median(Pb)) %>%
  ungroup() %>%
  group_by(date, growth) %>%
  summarise(return = mean(R1M_Usd)) %>%
  spread(key = growth, value = return) %>%
  ungroup()
```

`summarise()` regrouping output by 'date' (override with `.groups` argument)

```
colnames(monthly_return)[2:3] <- c("value", "growth")
```

```
monthly_return %>%
```

```
mutate(growth = cumprod(1 + growth), value = cumprod(1 + value)) %>%
gather(key = portfolio, value = value, -date) %>%
ggplot(aes(x = date, y = value, color = portfolio)) +
  geom_line()
```



3.) Instead of a unique threshold, compute simply sorted portfolios based on quartiles of market capitalization.

Compute their annual returns and plot them.

```
data_ml %>%
  group_by(date) %>%
  mutate(capitalization = ntile(Mkt_Cap_3M_Usd, 4)) %>%
  ungroup() %>%
  mutate(year = lubridate::year(date)) %>%
  group_by(year, capitalization) %>%
  summarise(return = mean(R1M_Usd)) %>%
```

```
mutate(capitalization = factor(capitalization, levels = 1:4, labels = c("small", "medium", "large", "x-large")) +  
ggplot(aes(year, return, fill = capitalization)) +  
  geom_col(position = "dodge")
```

`summarise()` regrouping output by 'year' (override with `.groups` argument)

