FINM4411

Most of the early code here is not pertinent and just used for other regressions and to load stuff. Skip to **SKIP TO HERE** for more pertinent stuff.

Java Installation (used for Powerpoint interface)

```
require(rJava)
## Loading required package: rJava
system("java -version")
```

Package Installation, not pertinent

```
#install.packages('quantmod')
#install.packages('xts')
#install.packages("ggplot2")
#install.packages("data.table")
#install.packages("gridExtra")
#install.packages("knitr")
#install.packages("zoo")
#install.packages("psych")
#install.packages("fBasics") ## online download, needed to find summary
statistics
#install.packages("kableExtra") ## online download, needed for Kable function
#install.packages("tidyverse") ## online download, needed for pull() function
#install.packages("lubridate")
#install.packages("lfe")
#install.packages("reshape2")
#install.packages("boot")
#install.packages("dplyr")
#install.packages("RCurl")
#install.packages("packrat")
#install.packages("rsconnect")
#install.packages("officer")
#install.packages("rvg")
#install.packages("scales")
#install.packages("magrittr")
#install.packages("devtools")
#install.packages("here")
#install.packages("glue")
#install.packages("viridis")
#install.packages("xlsx")
#install.packages("caret")
```

Library call, not pertinent

```
library(xts) ## 'r load-packages' command used above so Library is called
during Knit process
library(quantmod)
library(data.table)
library(ggplot2)
library(gridExtra)
library(fBasics)
library(knitr)
library(zoo)
library(psych)
library(kableExtra)
library(tidyverse)
library(lubridate)
library(lfe)
library(reshape2)
library(boot)
library(dplyr)
library(RCurl)
library(packrat)
library(rsconnect)
library(officer)
library(rvg)
                     # for formatting numbers
library(scales)
library(magrittr)
                     # for the %>% operator
library(devtools)
library(rJava)
library(customLayout)
library(here)
library(glue)
library(xlsx)
Pregin data download
Preqin1 <- read.csv('Preqin1 Values Only.csv', row.names = NULL,</pre>
stringsAsFactors = FALSE)
## above code saves Pregin1 Values Only.csv data into the variable 'Pregin1'
OLS regression of Final Close Size - US MN vs Net IRR
mod <- lm(Preqin1$NET.IRR.... ~ Preqin1$FINAL.CLOSE.SIZE..USD.MN., na.exclude
= TRUE)
## performs linear regression between Final Close Size - US MN vs Net IRR
summary(mod)
##
## Call:
## lm(formula = Preqin1$NET.IRR.... ~ Preqin1$FINAL.CLOSE.SIZE..USD.MN.,
       na.exclude = TRUE)
##
```

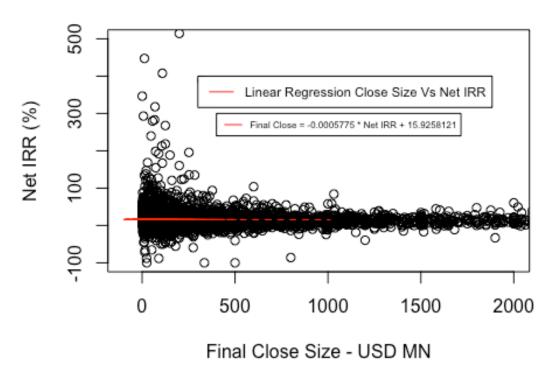
##

Residuals:

```
Min 10 Median
                       30
## -115.9
           -9.5
                -3.3
                        4.9 997.2
##
## Coefficients:
##
                                  Estimate Std. Error t value
## (Intercept)
                                  15.925812
                                            0.408506
                                                       38.99
## Pregin1$FINAL.CLOSE.SIZE..USD.MN. -0.000578
                                             0.000258
                                                       -2.24
                                            Pr(>|t|)
                                  ## (Intercept)
## Preqin1$FINAL.CLOSE.SIZE..USD.MN.
                                              0.025 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 27.9 on 5532 degrees of freedom
## Multiple R-squared: 0.000904, Adjusted R-squared: 0.000723
## F-statistic: 5 on 1 and 5532 DF, p-value: 0.0253
```

Below code creates scatter plot of Fund Final Close Size vs IRR

Correlation of Final Close Size and Net IRR (%)



Pregin data download using different data set

```
Preqin2<- read.csv('Preqin2 Values Only.csv', row.names = NULL,
stringsAsFactors = FALSE)
## above code saves Preqin2 Values Only.csv data into the variable 'Preqin2'</pre>
```

Sorts data by IRR, below shows the top 6 performing funds in the data set by Net IRR. As shown most funds are quite small with \$322 Million beng the largest. All but 2 are venture capital firms. Surprisingly, 3 out of the 6 have vintage years before 2000 which is interesting how they maintained such returns for so long. Another interesting thing is that all but 2 of them are currently liquidated. Also, all but 1 fund were/are located in the US.

```
TopIRR <- Preqin2[order(Preqin2$NET.IRR....),]
#tail(TopIRR) use this command to view the data but it doesn't Knit to pdf
well
```

Splitting data and for ease of use and finding basic mean IRR of US data

```
Geo<-split(Preqin2,Preqin2$GEOGRAPHIC.FOCUS)
USmean <- mean(Geo$US$NET.IRR....)</pre>
```

Creating dummy variables Vintage year using funds incepted before 2000 as Vin (Vintage)

```
Condense <- as.data.frame(cbind(Preqin2$VINTAGE...INCEPTION.YEAR,
Preqin2$NET.IRR...))
colnames(Condense) <- c("Date","IRR")
#creates data set with only Vintage and net IRR

for(i in 1:length(Condense$Date))
{
    if(Condense[i,"Date"]<2000 & Condense[i,"Date"]>1995)
    {
        Condense$Vin[i]=1
     }
     else
    {
        Condense$Vin[i]=0
     }
}
```

Creating dummy variables for Geographic Focus i.e. U.S. = 1, EUR = 0, ASIA = 0 for US data. Repeat process for different Geographic Focus areas.

```
Condense2 <- as.data.frame(cbind(Preqin2$GEOGRAPHIC.FOCUS,</pre>
Preqin2$NET.IRR....))
colnames(Condense2) <- c("GeoFocus","IRR")</pre>
#creates data set with only Geographic focus and net IRR data
#Below code creates US dummy variable
for(i in 1:length(Condense2$GeoFocus))
    if(Condense2[i, "GeoFocus"] == "US")
      Condense2$US[i]=1
    }
    else
      Condense2$US[i]=0
}
#Below code creates Asia dummy variable
for(i in 1:length(Condense2$GeoFocus))
{
    if(Condense2[i, "GeoFocus"] == "Asia")
      Condense2$Asia[i]=1
    }
    else
      Condense2$Asia[i]=0
```

```
#Below code creates India dummy variable
for(i in 1:length(Condense2$GeoFocus))
    if(Condense2[i, "GeoFocus"] == "India")
      Condense2$India[i]=1
    else
    {
      Condense2$India[i]=0
}
#Below code creates Australia dummy variable
for(i in 1:length(Condense2$GeoFocus))
    if(Condense2[i, "GeoFocus"] == "Australia")
      Condense2$Australia[i]=1
    }
    else
    {
      Condense2$Australia[i]=0
}
#Below code creates Africa dummy variable
for(i in 1:length(Condense2$GeoFocus))
    if(Condense2[i, "GeoFocus"] == "Africa")
      Condense2$Africa[i]=1
    }
    else
      Condense2$Africa[i]=0
}
#Below code creates Europe dummy variable
for(i in 1:length(Condense2$GeoFocus))
    if(Condense2[i, "GeoFocus"] == "Europe")
      Condense2$Europe[i]=1
    else
```

```
Condense2$Europe[i]=0
}

for(i in 1:length(Condense2$GeoFocus))
{
    if(Condense2[i,"GeoFocus"] == "Brazil")
    {
        Condense2$Brazil[i]=1
    }
    else
    {
        Condense2$Brazil[i]=0
    }
}
```

Multiple Regression

```
ols_no_int <- lm(Condense2$IRR ~ Condense2$US + Condense2$Asia +
Condense2$India + Condense2$Australia + Condense2$Africa + Condense2$Brazil -
# -1 means exclude intercept
summary(ols_no_int)
##
## Call:
## lm(formula = Condense2$IRR ~ Condense2$US + Condense2$Asia +
      Condense2$India + Condense2$Australia + Condense2$Africa +
##
      Condense2$Brazil - 1)
##
## Residuals:
     Min
             10 Median
                           30
                                 Max
## -103.5
           -2.6
                   8.1
                         16.9 1013.0
##
## Coefficients:
                      Estimate Std. Error t value
                                                             Pr(>|t|)
##
## Condense2$US
                                    0.729
                                           15.587
## Condense2$Asia
                        13.047
                                    1.890
                                            6.90
                                                      0.000000000057 ***
## Condense2$India
                                                              0.00036 ***
                        17.716
                                    4.963
                                             3.57
                                                      0.0000031082287 ***
## Condense2$Australia
                        18.287
                                    3.917
                                            4.67
## Condense2$Africa
                        9.635
                                    7.019
                                             1.37
                                                              0.16989
## Condense2$Brazil
                        13.530
                                    6.676
                                            2.03
                                                              0.04275 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 30.6 on 5415 degrees of freedom
## Multiple R-squared: 0.0916, Adjusted R-squared: 0.0906
## F-statistic: 91 on 6 and 5415 DF, p-value: <0.0000000000000000
```

T-Test to determine if there is a significant difference in the net IRR's of the U.S. and Australia

```
t.test(c(Geo$US$NET.IRR....),c(Geo$Australia$NET.IRR....))

##

## Welch Two Sample t-test

##

## data: c(Geo$US$NET.IRR....) and c(Geo$Australia$NET.IRR....)

## t = -1.171, df = 73.88, p-value = 0.246

## alternative hypothesis: true difference in means is not equal to 0

## 95 percent confidence interval:

## -7.29636 1.89632

## sample estimates:

## mean of x mean of y

## 15.5865 18.2866

## Performs Welch Two Sample t-test
```

Cleaned Pregin Data

```
Preqinclean<- read.csv('Calerecent.csv', row.names = NULL, stringsAsFactors =
FALSE)
## above code saves Calerecent.csv data into the variable 'Preqinclean'

Preqin_mod <- as.data.frame(cbind(Preqinclean$NET.IRR...,
Preqinclean$FUND.SIZE..USD.MN., Preqinclean$Market.rate...Y.N...,
Preqinclean$Impact...Y.N..., Preqinclean$STRATEGY,
Preqinclean$PRIMARY.REGION.FOCUS, Preqinclean$VINTAGE...INCEPTION.YEAR))
colnames(Preqin_mod) <- c("IRR", "Fundsize", "Market_Rate?", "Impact?",
"Strategy", "RegionFocus", "Vintage")
#creates data set with only Market Rate and net IRR data</pre>
```

Creating Dummy variables

```
#Below code creates market return dummy variable
for(i in 1:length(Preqin_mod$`Market_Rate?`))
{
    if(Preqin_mod[i,"Market_Rate?"] == "Y")
    {
        Preqin_mod$MarketDUM[i]=1
    }
    else
    {
        Preqin_mod$MarketDUM[i]=0
    }
}
head(Preqin_mod)
```

```
IRR Fundsize Market Rate? Impact?
                                             Strategy
                                                        RegionFocus Vintage
## 1 7.88
                167
                                Υ
                                        N Early Stage North America
                                                                        2007
                                Υ
## 2 21.79
                175
                                        N Early Stage North America
                                                                        2011
                               Υ
## 3 11.51
                217
                                        N Early Stage North America
                                                                        2015
## 4 11.6
              203.4
                               Υ
                                        Ν
                                               Growth North America
                                                                        2015
                               Υ
## 5 11.4
                475
                                        Ν
                                               Buyout North America
                                                                        1993
## 6 -5.9
                               Υ
                530
                                        Ν
                                               Buyout North America
                                                                        1998
##
     MarketDUM
## 1
## 2
             1
## 3
             1
             1
## 4
             1
## 5
## 6
             1
#Below code creates impact dummy variable where dummy = 1 for Y
for(i in 1:length(Preqinclean$Impact...Y.N...))
{
    if(Preqin_mod[i,"Impact?"] == "Y")
      Preqin_mod$ImpactDUM[i]=1
    }
    else
      Preqin mod$ImpactDUM[i]=0
}
#Creating more dummy variables for Strategy i.e. Early stage = 1, Growth = 0,
Buyout = 0 for US data. Repeat process for different categorical strategies.
#Below code strategy dummy variables
for(i in 1:length(Preqin_mod$Strategy))
{
    if(Preqin_mod[i, "Strategy"] == "Early Stage")
      Preqin mod$Early StageDUM[i]=1
    }
    else
      Preqin_mod$Early_StageDUM[i]=0
    }
}
for(i in 1:length(Preqin mod$Strategy))
    if(Preqin_mod[i, "Strategy"] == "Growth")
    {
      Preqin_mod$GrowthDUM[i]=1
```

```
else
      Preqin_mod$GrowthDUM[i]=0
}
for(i in 1:length(Preqin_mod$Strategy))
    if(Preqin_mod[i,"Strategy"] == "Buyout")
      Preqin_mod$BuyoutDUM[i]=1
    else
      Preqin_mod$BuyoutDUM[i]=0
}
for(i in 1:length(Preqin_mod$Strategy))
    if(Preqin_mod[i, "Strategy"] == "Fund of Funds")
      Preqin_mod$Fund_of_FundsDUM[i]=1
    else
      Preqin_mod$Fund_of_FundsDUM[i]=0
}
for(i in 1:length(Preqin_mod$Strategy))
    if(Preqin_mod[i,"Strategy"] == "Venture (General)")
      Preqin_mod$Venture_GeneralDUM[i]=1
    }
    else
      Preqin_mod$Venture_GeneralDUM[i]=0
}
for(i in 1:length(Preqin_mod$Strategy))
{
    if(Preqin_mod[i,"Strategy"] == "Early Stage: Seed")
      Preqin_mod$Early_Stage_SeedDUM[i]=1
```

```
else
    {
      Preqin_mod$Early_Stage_SeedDUM[i]=0
}
for(i in 1:length(Preqin_mod$Strategy))
    if(Preqin_mod[i, "Strategy"] == "Co-Investment")
    {
      Preqin_mod$Co_InvestmentDUM[i]=1
    else
    {
      Preqin_mod$Co_InvestmentDUM[i]=0
    }
}
#Below code creates North America dummy variable
for(i in 1:length(Preqin mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "North America")
    {
      Preqin_mod$North_America_DUM[i]=1
    }
    else
      Preqin_mod$North_America_DUM[i]=0
    }
}
#Below code creates Europe dummy variable
for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "Europe")
    {
      Preqin_mod$Europe_DUM[i]=1
    }
    else
      Preqin_mod$Europe_DUM[i]=0
    }
}
#Below code creates Asia dummy variable
for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "Asia")
```

```
Preqin_mod$Asia_DUM[i]=1
    }
    else
    {
      Preqin_mod$Asia_DUM[i]=0
}
#Below code creates Diversified Multi-Regional dummy variable
for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "Diversified Multi-Regional")
      Preqin_mod$Diversified_Multi_Regional_DUM[i]=1
    else
    {
      Preqin_mod$Diversified_Multi_Regional_DUM[i]=0
}
#Below code creates Americas dummy variable
for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "Americas")
    {
      Preqin_mod$Americas_DUM[i]=1
    else
    {
      Preqin mod$Americas DUM[i]=0
}
#Below code creates Africa dummy variable
for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i, "RegionFocus"] == "Africa")
    {
      Preqin_mod$Africa_DUM[i]=1
    else
      Preqin_mod$Africa_DUM[i]=0
}
for(i in 1:length(Preqin_mod$RegionFocus))
```

```
if(Preqin_mod[i,"RegionFocus"] == "Middle East & Israel")
{
    Preqin_mod$Middle_East_and_Israel_DUM[i]=1
}
else
{
    Preqin_mod$Middle_East_and_Israel_DUM[i]=0
}
}

for(i in 1:length(Preqin_mod$RegionFocus))
{
    if(Preqin_mod[i,"RegionFocus"] == "Australasia")
    {
        Preqin_mod$Australasia_DUM[i]=1
    }
    else
    {
        Preqin_mod$Australasia_DUM[i]=0
    }
}
```

Below code creates a matrix of dummy variables with 1969 - 2017 as columns and the length of the total dataset as rows.

```
min <- min(Preginclean$VINTAGE...INCEPTION.YEAR)</pre>
k <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1</pre>
n <- length(Preqinclean$NET.IRR....)</pre>
IRR <- matrix(NA, nrow=n, ncol=k)</pre>
for(j in 1:k){
for(i in 1:n)
    if(Preginclean$VINTAGE...INCEPTION.YEAR[i] == sum((min-1)+j))
      IRR[i,j]=1
    }
    else
      IRR[i,j]=0
}
}
Vintage_all <- IRR[,1:k]</pre>
colnames(Vintage_all) <- c(1969:2017)</pre>
Vintage <- Vintage all[,c(1,3,4,8:49)]
# above code removes vintage years with no data
```

SKIP TO HERE Multiple Regression for Dummy variables

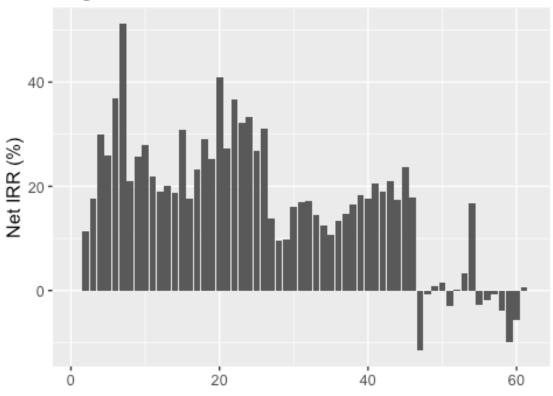
```
ols no int1 <- lm(Preginclean$NET.IRR.... ~ Preginclean$FUND.SIZE..USD.MN. +
Vintage + Preqin_mod$ImpactDUM + Preqin_mod$Early_StageDUM +
Pregin mod$GrowthDUM + Pregin mod$BuyoutDUM + Pregin mod$Fund of FundsDUM +
Preqin mod$Venture GeneralDUM + Preqin mod$Early Stage SeedDUM +
Pregin mod$Co InvestmentDUM + Pregin mod$North America DUM +
Preqin mod$Europe DUM + Preqin mod$Asia DUM +
Pregin mod$Diversified Multi Regional DUM + Pregin mod$Americas DUM +
Preqin_mod$Africa_DUM + Preqin_mod$Middle_East_and_Israel_DUM - 1)
summary(ols_no_int1)
##
## Call:
## lm(formula = Preqinclean$NET.IRR.... ~ Preqinclean$FUND.SIZE..USD.MN. +
       Vintage + Preqin_mod$ImpactDUM + Preqin_mod$Early_StageDUM +
##
       Preqin mod$GrowthDUM + Preqin mod$BuyoutDUM +
##
Pregin mod$Fund of FundsDUM +
       Preqin mod$Venture GeneralDUM + Preqin mod$Early Stage SeedDUM +
##
       Pregin mod$Co InvestmentDUM + Pregin mod$North America DUM +
##
##
       Preqin mod$Europe DUM + Preqin mod$Asia DUM +
Preqin_mod$Diversified_Multi_Regional_DUM +
       Pregin mod$Americas DUM + Pregin mod$Africa DUM +
Preqin mod$Middle East and Israel DUM -
##
       1)
##
## Residuals:
##
      Min
              10 Median
                            3Q
                                  Max
## -109.2
            -9.4
                   -1.7
                           5.3 975.4
##
## Coefficients:
##
                                               Estimate Std. Error t value
## Preqinclean$FUND.SIZE..USD.MN.
                                               -0.000406
                                                           0.000269
                                                                      -1.51
## Vintage1969
                                               11.284486 27.437140
                                                                       0.41
## Vintage1971
                                               17.673024 27.402551
                                                                       0.64
                                               29.883942 19.524361
## Vintage1972
                                                                       1.53
## Vintage1976
                                                                       0.94
                                               25.901488 27.432697
                                               36.789712 27.433862
                                                                       1.34
## Vintage1977
## Vintage1978
                                               51.140781 19.556103
                                                                       2.62
                                               20.894998 19.545099
## Vintage1979
                                                                       1.07
## Vintage1980
                                               25.768722
                                                          9.674835
                                                                       2.66
## Vintage1981
                                               28.042823 12.657122
                                                                       2.22
## Vintage1982
                                               21.810299
                                                           9.658786
                                                                       2.26
## Vintage1983
                                               18.920370
                                                           7.569921
                                                                       2.50
## Vintage1984
                                               20.130086
                                                           6.503893
                                                                       3.10
## Vintage1985
                                                                       2.75
                                               18.802448
                                                           6.830149
## Vintage1986
                                               30.741341
                                                                       4.94
                                                           6.223076
                                                                       3.07
## Vintage1987
                                               17.673100
                                                           5.750311
## Vintage1988
                                               23.173859
                                                           5.568281
                                                                       4.16
```

```
## Vintage1989
                                                 29.088538
                                                              5.335792
                                                                           5.45
## Vintage1990
                                                 25.154619
                                                              5.225110
                                                                           4.81
## Vintage1991
                                                 40.938513
                                                              6.194072
                                                                           6.61
## Vintage1992
                                                 27.362590
                                                              4.819256
                                                                           5.68
## Vintage1993
                                                 36.609729
                                                              4.650684
                                                                           7.87
## Vintage1994
                                                 32.283172
                                                              4.518855
                                                                           7.14
## Vintage1995
                                                 33.360303
                                                              4.391574
                                                                           7.60
## Vintage1996
                                                 26.892273
                                                              4.351530
                                                                           6.18
## Vintage1997
                                                 30.995553
                                                              4.039081
                                                                           7.67
## Vintage1998
                                                 13.873730
                                                              3.826120
                                                                           3.63
## Vintage1999
                                                  9.667120
                                                              3.885471
                                                                           2.49
## Vintage2000
                                                              3.724456
                                                                           2.66
                                                  9.891320
## Vintage2001
                                                 16.009512
                                                              3.847051
                                                                           4.16
## Vintage2002
                                                 17.072216
                                                              4.069279
                                                                           4.20
                                                 17.110711
                                                                           4.19
## Vintage2003
                                                              4.080265
## Vintage2004
                                                 14.462197
                                                              3.865120
                                                                           3.74
## Vintage2005
                                                 12.447448
                                                              3.675453
                                                                           3.39
## Vintage2006
                                                 10.609798
                                                              3.630801
                                                                           2.92
## Vintage2007
                                                 13.305622
                                                              3.587058
                                                                           3.71
## Vintage2008
                                                 14.634103
                                                              3.611065
                                                                           4.05
## Vintage2009
                                                 16.523450
                                                              3.996982
                                                                           4.13
## Vintage2010
                                                                           4.73
                                                 18.197432
                                                              3.849290
                                                                           4.72
## Vintage2011
                                                 17.581732
                                                              3.721743
## Vintage2012
                                                 20.474217
                                                              3.732381
                                                                           5.49
## Vintage2013
                                                 18.882504
                                                              3.682767
                                                                           5.13
## Vintage2014
                                                 21.069317
                                                              3.631465
                                                                           5.80
## Vintage2015
                                                 17.437361
                                                              3.629764
                                                                           4.80
## Vintage2016
                                                 23.740167
                                                              3.596752
                                                                           6.60
## Vintage2017
                                                 17.936597
                                                                           4.81
                                                              3.725838
## Pregin mod$ImpactDUM
                                                -11.433887
                                                              6.062649
                                                                          -1.89
## Preqin mod$Early StageDUM
                                                              1.679796
                                                                          -0.45
                                                 -0.755503
## Preqin_mod$GrowthDUM
                                                                           0.48
                                                  0.825214
                                                              1.705758
## Pregin mod$BuyoutDUM
                                                  1.476570
                                                              1.213141
                                                                           1.22
## Preqin_mod$Fund_of_FundsDUM
                                                 -2.930747
                                                              1.351989
                                                                          -2.17
## Preqin_mod$Venture_GeneralDUM
                                                  0.173512
                                                              1.394699
                                                                           0.12
## Preqin_mod$Early_Stage_SeedDUM
                                                  3.393560
                                                              3.218596
                                                                           1.05
## Preqin mod$Co InvestmentDUM
                                                 16.637105
                                                              3.147412
                                                                           5.29
## Preqin_mod$North_America_DUM
                                                 -2.756373
                                                              3.192758
                                                                          -0.86
## Pregin mod$Europe DUM
                                                 -1.929736
                                                              3.253272
                                                                          -0.59
## Preqin_mod$Asia_DUM
                                                 -0.681247
                                                              3.395362
                                                                          -0.20
## Preqin_mod$Diversified_Multi_Regional_DUM
                                                 -3.839031
                                                              4.892482
                                                                          -0.78
## Preqin_mod$Americas_DUM
                                                 -9.965697
                                                              4.291868
                                                                          -2.32
## Pregin mod$Africa DUM
                                                 -5.742518
                                                              5.432838
                                                                          -1.06
## Preqin mod$Middle East and Israel DUM
                                                  0.596627
                                                              4.636518
                                                                           0.13
                                                          Pr(>|t|)
##
## Preqinclean$FUND.SIZE..USD.MN.
                                                            0.13063
                                                           0.68088
## Vintage1969
## Vintage1971
                                                            0.51899
## Vintage1972
                                                            0.12593
## Vintage1976
                                                            0.34512
```

```
## Vintage1977
                                                          0.17997
## Vintage1978
                                                          0.00895 **
## Vintage1979
                                                          0.28509
                                                          0.00776 **
## Vintage1980
## Vintage1981
                                                          0.02676 *
                                                          0.02398 *
## Vintage1982
## Vintage1983
                                                          0.01247 *
                                                          0.00198 **
## Vintage1984
## Vintage1985
                                                          0.00593 **
## Vintage1986
                                              0.0000008050909960
                                                          0.00213 **
## Vintage1987
                                              0.0000320686696775
## Vintage1988
## Vintage1989
                                              0.0000000521273469
## Vintage1990
                                              0.0000015181520002
## Vintage1991
                                              0.0000000000423078
## Vintage1992
                                              0.0000000143551236
## Vintage1993
                                              0.00000000000000042
                                              0.0000000000010253 ***
## Vintage1994
## Vintage1995
                                              0.00000000000000356
                                              0.0000000006879094 ***
## Vintage1996
                                              0.0000000000000197 ***
## Vintage1997
                                                          0.00029 ***
## Vintage1998
## Vintage1999
                                                          0.01288 *
## Vintage2000
                                                          0.00794 **
## Vintage2001
                                              0.0000321049192452 ***
## Vintage2002
                                               0.0000276746150428
                                              0.0000279020973037 ***
## Vintage2003
                                                          0.00018 ***
## Vintage2004
                                                                  ***
## Vintage2005
                                                          0.00071
                                                          0.00349 **
## Vintage2006
## Vintage2007
                                                          0.00021
## Vintage2008
                                              0.0000513654840943
## Vintage2009
                                              0.0000361906718468
## Vintage2010
                                              0.0000023307261142
## Vintage2011
                                              0.0000023701087761
                                              0.0000000430810318 ***
## Vintage2012
## Vintage2013
                                              0.0000003041891307
## Vintage2014
                                              0.0000000069285996
## Vintage2015
                                              0.0000015971637571
                                              0.0000000000448935 ***
## Vintage2016
## Vintage2017
                                              0.0000015186734923 ***
## Pregin mod$ImpactDUM
                                                          0.05935
## Preqin mod$Early StageDUM
                                                          0.65290
## Pregin mod$GrowthDUM
                                                          0.62856
## Pregin mod$BuyoutDUM
                                                          0.22360
## Preqin_mod$Fund_of_FundsDUM
                                                          0.03022 *
## Preqin_mod$Venture_GeneralDUM
                                                          0.90100
## Preqin mod$Early Stage SeedDUM
                                                          0.29177
## Pregin mod$Co InvestmentDUM
                                              0.0000001299357425
## Pregin mod$North America DUM
                                                          0.38800
```

```
## Pregin mod$Europe DUM
                                                   0.55309
## Pregin mod$Asia DUM
                                                   0.84099
## Preqin mod$Diversified Multi Regional DUM
                                                   0.43268
## Preqin mod$Americas DUM
                                                   0.02027 *
## Preqin mod$Africa DUM
                                                   0.29056
## Preqin_mod$Middle_East_and_Israel_DUM
                                                   0.89762
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 27.2 on 5428 degrees of freedom
## Multiple R-squared: 0.286, Adjusted R-squared: 0.278
ols.beta1 = matrix(NA, length(ols no int1$coefficients), 1)
# below code used to make the chart of the coefficients
for(i in 1:length(ols no int1$coefficients)){
ols.beta1[i] = ols_no_int1[1]$coefficients[i]
}
plot.ols.beta1 = data.frame( y = ols.beta1, x =
c(1:length(ols_no_int1$coefficients)))
ggplot(plot.ols.beta1, aes(x, y)) + geom_col() + theme(text =
element text(size=12)) + ylab("Net IRR (%)") + xlab("") +
ggtitle("Regression Coefficients")
```

Regression Coefficients



creating a matrix for excel

Multiple Regression for Dummy variables with insignifant variables removed.

```
Vintage <- Vintage_all[,c(10,12:49)]
# above code removes insignificant Vintage Dummy variables from model 1

ols_no_int2 <- lm(Preqinclean$NET.IRR.... ~ Vintage + Preqin_mod$ImpactDUM +
Preqin_mod$Fund_of_FundsDUM + Preqin_mod$Co_InvestmentDUM +
Preqin_mod$Americas_DUM - 1)

# So far I have removed the insignificant vintage year dummy variables. Also
I have removed: Early stage, Growth, Buyout, Venture General, Early Stage
Seed strategies as well as Diversified Regionals and Africa as primary region
focus area dummy variables.

summary(ols_no_int2)

##
## Call:
## Tm(formula = Preqinclean$NET.IRR.... ~ Vintage + Preqin_mod$ImpactDUM +
## Preqin_mod$Fund_of_FundsDUM + Preqin_mod$Co_InvestmentDUM +</pre>
```

```
Preqin_mod$Americas_DUM - 1)
##
##
## Residuals:
               1Q Median
                             3Q
##
      Min
                                    Max
## -108.6
            -9.4
                    -1.9
                            5.3 975.5
##
## Coefficients:
##
                                 Estimate Std. Error t value
Pr(>|t|)
                                               19.244
## Vintage1978
                                   48.550
                                                         2.52
0.01167
                                               9.071
## Vintage1980
                                   23.610
                                                         2.60
0.00928
                                   25.099
                                               12.185
                                                         2.06
## Vintage1981
0.03946
                                   19.711
                                               9.072
                                                         2.17
## Vintage1982
0.02985
                                   16.780
## Vintage1983
                                                6.804
                                                         2.47
0.01368
## Vintage1984
                                   17.829
                                                5.555
                                                         3.21
0.00134
                                   16.905
                                                5.939
                                                         2.85
## Vintage1985
0.00444
                                   28.532
                                                5.238
                                                         5.45
## Vintage1986
0.00000005329476938
## Vintage1987
                                   15.511
                                               4.668
                                                         3.32
0.00090
## Vintage1988
                                   21.186
                                               4.537
                                                         4.67
0.00000308913808779
                                   26.982
                                               4.150
                                                         6.50
## Vintage1989
0.00000000008692261
                                   23.237
                                               4.013
                                                         5.79
## Vintage1990
0.00000000742801260
## Vintage1991
                                   38.948
                                                5.239
                                                         7.43
0.0000000000012089
                                                         7.24
                                   25.449
                                                3.515
## Vintage1992
0.00000000000050774
                                   34.626
                                                        10.48 <
## Vintage1993
                                                3.303
0.00000000000000000
                                   30.449
                                                        10.01 <
## Vintage1994
                                                3.043
0.0000000000000000
                                   31.365
                                                2.921
                                                        10.74 <
## Vintage1995
0.00000000000000000
## Vintage1996
                                   24.977
                                                2.886
                                                         8.65 <
0.00000000000000000
                                   29.057
                                                2.278
                                                        12.75 <
## Vintage1997
0.0000000000000000
## Vintage1998
                                   11.889
                                                2.004
                                                         5.93
0.00000000318616543
## Vintage1999
                                    7.660
                                                2.017
                                                         3.80
```

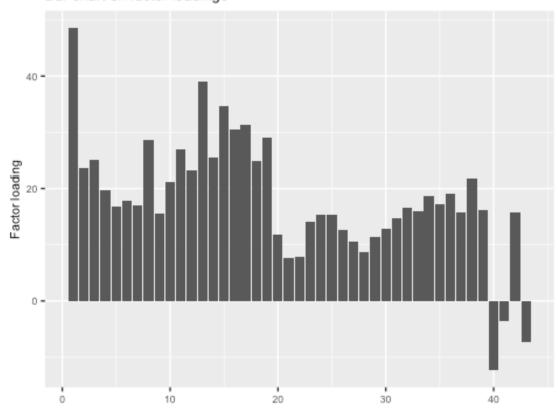
0.00015	7.000	4 744	4 64	
## Vintage2000	7.889	1.711	4.61	
0.0000411966716234	14 001	2 052	6 92	
## Vintage2001	14.001	2.052	6.82	
0.00000000000987810	15 270	2 204	C 41	
## Vintage2002	15.279	2.384	6.41	
0.0000000015919632	15 215	2 442	6 27	
## Vintage2003	15.315	2.443	6.27	
0.0000000038817913	12 671	2 007	6 07	
## Vintage2004	12.671	2.087	6.07	
0.0000000136601078	10 601	1 725	c 11	
## Vintage2005	10.601	1.735	6.11	
0.0000000105257089	9 605	1 522	Г 6Г	
## Vintage2006 0.00000001659747374	8.605	1.522	5.65	
	11 470	1 500	7.61	
## Vintage2007 0.00000000000003300	11.479	1.509	7.61	
	12 900	1 566	8.17	
## Vintage2008	12.800	1.566	8.17	
0.0000000000000037	14.787	2.292	6.45	
## Vintage2009 0.00000000012142755	14./0/	2.292	0.45	
	16.610	2.069	8.03	
## Vintage2010 0.00000000000000119	10.010	2.009	0.05	
## Vintage2011	15.883	1.763	9.01 <	
0.000000000000000000000000000000000000	13.883	1.703	9.01 \	
## Vintage2012	18.680	1.780	10.49 <	
0.000000000000000000000000000000000000	10.000	1.700	10.45 (
## Vintage2013	17.178	1.723	9.97 <	
0.0000000000000000000000000000000000000	17.170	1.723	J.J/ \	
## Vintage2014	19.173	1.585	12.09 <	
0.0000000000000000000000000000000000000	13.173	1.505	12.05	
## Vintage2015	15.748	1.617	9.74 <	
0.0000000000000000000000000000000000000	13.710	1.017	3.71	
## Vintage2016	21.804	1.544	14.12 <	
0.0000000000000000000000000000000000000	21.001	1.5	11112 \	
## Vintage2017	16.062	1.761	9.12 <	
0.0000000000000000000000000000000000000		_,,,_		
## Preqin_mod\$ImpactDUM	-12.308	5.976	-2.06	
0.03948		21270	_,,,	
## Preqin mod\$Fund of FundsDUM	-3.596	0.974	-3.69	
0.00023	2122		2102	
## Preqin_mod\$Co_InvestmentDUM	15.662	3.001	5.22	
0.0000018607751614				
## Preqin mod\$Americas DUM	-7.335	2.925	-2.51	
0.01218				
##				
## Vintage1978	*			
## Vintage1980	**			
## Vintage1981	*			
## Vintage1982	*			
U				

```
## Vintage1983
## Vintage1984
## Vintage1985
## Vintage1986
## Vintage1987
## Vintage1988
## Vintage1989
## Vintage1990
## Vintage1991
                                ***
## Vintage1992
## Vintage1993
## Vintage1994
## Vintage1995
                                ***
## Vintage1996
                                ***
## Vintage1997
                                ***
## Vintage1998
                                ***
## Vintage1999
## Vintage2000
                                ***
## Vintage2001
## Vintage2002
## Vintage2003
## Vintage2004
                                ***
## Vintage2005
                                ***
## Vintage2006
                                ***
## Vintage2007
## Vintage2008
                                ***
## Vintage2009
## Vintage2010
## Vintage2011
## Vintage2012
                                ***
## Vintage2013
                                ***
## Vintage2014
                                ***
## Vintage2015
                                ***
## Vintage2016
## Vintage2017
## Preqin_mod$ImpactDUM
## Preqin_mod$Fund_of_FundsDUM ***
## Preqin_mod$Co_InvestmentDUM ***
## Preqin mod$Americas DUM
## ---
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
## Residual standard error: 27.2 on 5446 degrees of freedom
## Multiple R-squared: 0.284, Adjusted R-squared: 0.278
## F-statistic: 50.2 on 43 and 5446 DF, p-value: <0.00000000000000000
ols.beta2 = matrix(NA, length(ols_no_int2$coefficients), 1)
# below code used to make the chart of the coeffients
for(i in 1:length(ols_no_int2$coefficients)){
```

```
ols.beta2[i] = ols_no_int2[1]$coefficients[i]
}

plot.ols.beta2 = data.frame( y = ols.beta2, x =
c(1:length(ols_no_int2$coefficients)))
ggplot(plot.ols.beta2, aes(x, y)) + geom_col() + theme(text =
element_text(size=8)) + ylab("Factor loading") + xlab("") +
ggtitle("Bar chart on factor loadings")
```

Bar chart on factor loadings



Multiple Regression for Dummy variables with all variables removed besides those of Signif. codes: 0 '***'.

```
Vintage <- Vintage_all[,c(18,20:30,33:37,39:49)]
# above code creates a subset of vintage years which had a p-value of almost
0 (*** Significance code from model 1 regression1)

ols_no_int3 <- lm(Preqinclean$NET.IRR.... ~ Vintage +
Preqin_mod$Co_InvestmentDUM - 1)

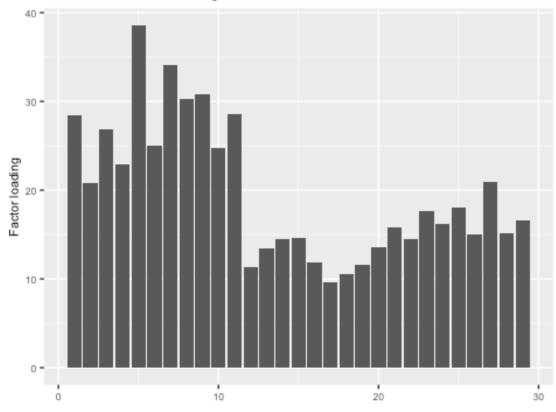
summary(ols_no_int3)

##
## Call:</pre>
```

```
## lm(formula = Preqinclean$NET.IRR.... ~ Vintage +
Preqin_mod$Co_InvestmentDUM -
##
       1)
##
## Residuals:
      Min
               1Q Median
                             3Q
##
                                    Max
## -111.4
            -8.4
                    -1.1
                            8.0
                                 975.4
##
## Coefficients:
                                Estimate Std. Error t value
##
Pr(>|t|)
                                                5.30
## Vintage1986
                                    28.40
                                                         5.36
0.000000085639456
                                    20.79
                                                4.59
                                                         4.53
## Vintage1988
0.000005966157386
                                    26.81
                                                4.20
                                                         6.39
## Vintage1989
0.00000000180427
                                    22.92
                                                4.06
## Vintage1990
                                                         5.65
0.000000016950283
                                    38.55
                                                5.30
                                                         7.28
## Vintage1991
0.000000000000385
                                    25.03
                                                3.55
                                                         7.04
## Vintage1992
0.000000000002093
                                    34.15
                                                3.34
                                                        10.23 <
## Vintage1993
0.00000000000000000
## Vintage1994
                                    30.27
                                                3.08
                                                         9.84 <
0.00000000000000002
## Vintage1995
                                    30.77
                                                2.95
                                                        10.43 <
0.0000000000000000
                                    24.68
                                                2.92
                                                         8.46 <
## Vintage1996
0.00000000000000000
                                    28.60
                                                        12.43 <
## Vintage1997
                                                2.30
0.00000000000000000
## Vintage1998
                                    11.38
                                                2.02
                                                         5.62
0.000000019792234
                                                2.07
                                                         6.46
                                    13.37
## Vintage2001
0.000000000113084
                                                2.40
                                                         6.03
                                    14.51
## Vintage2002
0.000000001714762
                                    14.57
                                                2.46
                                                         5.92
## Vintage2003
0.000000003452008
                                    11.80
                                                2.10
                                                         5.62
## Vintage2004
0.000000019670831
## Vintage2005
                                     9.62
                                                1.74
                                                         5.54
0.000000032384118
                                    10.50
                                                1.51
                                                         6.96
## Vintage2007
0.00000000003879
## Vintage2008
                                    11.61
                                                1.56
                                                         7.45
0.000000000000106
## Vintage2009
                                    13.52
                                                2.30
                                                         5.88
```

0.000000004447018	15 02	2.00	7.60	
## Vintage2010	15.82	2.08	7.60	
0.000000000000035	1/ 51	1 76	0 26 /	
## Vintage2011 0.00000000000000002	14.51	1.76	8.26 <	
## Vintage2012	17.61	1.78	9.87 <	
0.000000000000000000000000000000000000	17.01	1.78	<i>3.</i> 67 \	
## Vintage2013	16.18	1.73	9.36 <	
0.000000000000000000000000000000000000	10.10	1.75	J. 30 X	
## Vintage2014	18.10	1.59	11.42 <	
0.000000000000000000000000000000000000	10.10	1.55	44.76 \	
## Vintage2015	15.03	1.63	9.24 <	
0.0000000000000000000000000000000000000	13.03	2.05	- J	
## Vintage2016	20.96	1.55	13.52 <	
0.0000000000000000000000000000000000000			,	
## Vintage2017	15.12	1.77	8.55 <	
0.0000000000000000000000000000000000000		,,		
## Preqin_mod\$Co_InvestmentDUM	16.62	3.02	5.50	
0.00000040323022			-	
##				
## Vintage1986	***			
## Vintage1988	***			
## Vintage1989	***			
## Vintage1990	***			
## Vintage1991	***			
## Vintage1992	***			
## Vintage1993	***			
## Vintage1994	***			
## Vintage1995	***			
## Vintage1996	***			
## Vintage1997	***			
## Vintage1998	***			
## Vintage2001	***			
## Vintage2002	***			
## Vintage2003	***			
## Vintage2004	***			
## Vintage2005	***			
## Vintage2007	***			
## Vintage2008	***			
## Vintage2009	***			
## Vintage2010	***			
## Vintage2011	***			
## Vintage2012	***			
## Vintage2013	***			
## Vintage2014	***			
## Vintage2015	***			
## Vintage2016	***			
## Vintage2017	***			
<pre>## Preqin_mod\$Co_InvestmentDUM</pre>	***			
##				

Bar chart on factor loadings



Checking the average of Impact and Non Impact to make sure our regression is correct.

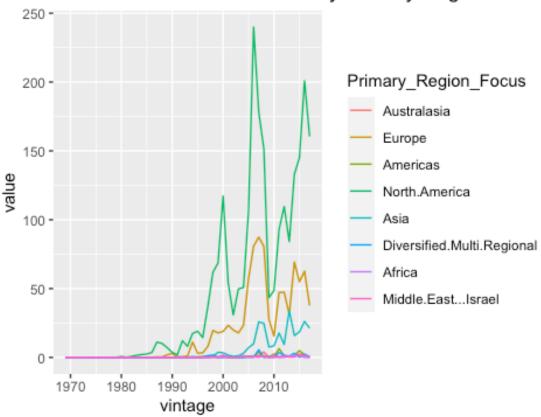
```
Impact.idx <- which(Preqinclean$Impact...Y.N...=="Y")
meanImpactIRR <- mean(Preqinclean$NET.IRR....[Impact.idx])
print(meanImpactIRR)</pre>
```

```
## [1] 3.01048
# mean of Impact IRR
nonImpact.idx <- which(Preginclean$Impact...Y.N...=="N")</pre>
mean nonImpactIRR <- mean(Preqinclean$NET.IRR....[nonImpact.idx])</pre>
print(mean nonImpactIRR)
## [1] 15.6843
# mean of non Impact IRR (every fund that is not impact)
print(mean_nonImpactIRR - meanImpactIRR)
## [1] 12.6738
# difference between the two.
# In most of our regressions the impact dummy was around -12. Therefore the
impact dummy must mean how much less IRR impact funds get compared to
standard funds on average
# Now lets try the same thing with "Buyout" as a strategy to make sure we
have our bearings for that correct also
Buyout.idx <- which(Preqinclean$STRATEGY=="Buyout")</pre>
meanBuyoutIRR <- mean(Preqinclean$NET.IRR....[Buyout.idx])</pre>
print(meanBuyoutIRR)
## [1] 16.5112
# mean of Buyout IRR
nonBuyout.idx <- which(Preginclean$STRATEGY!="Buyout")</pre>
meannonBuyoutIRR <- mean(Preqinclean$NET.IRR....[nonBuyout.idx])</pre>
print(meannonBuyoutIRR)
## [1] 15.1863
# mean of non Buyout IRR
print(meanBuyoutIRR - meannonBuyoutIRR)
## [1] 1.32488
# again so the Dummy's we are getting for the Buyout Dum are around 1.5 which
probably means that the buyout dummy's (or any strategy dummy's refer to just
the premium or reduction in IRR if your fund is a Buyout fund compared to any
other fund that is not Buyout)
# The final check I will do is for the Vintage year Dummy variables.
vintage.idx <- which(Preqinclean$VINTAGE...INCEPTION.YEAR==2017)</pre>
meanvintageIRR <- mean(Preqinclean$NET.IRR....[vintage.idx])</pre>
print(meanvintageIRR)
```

```
## [1] 15.7332
\# above is the mean IRR for the vintage year of x.
# I'll check below
allvintage.idx <- which(Preqinclean$VINTAGE...INCEPTION.YEAR!=2017)
meanallvintageIRR <- mean(Preqinclean$NET.IRR....[allvintage.idx])</pre>
print(meanallvintageIRR)
## [1] 15.6313
print(meanvintageIRR - meanallvintageIRR)
## [1] 0.101903
Vintage Geo map code.
min <- min(Preginclean$VINTAGE...INCEPTION.YEAR)</pre>
n <- max(Preginclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(Preqinclean$PRIMARY.REGION.FOCUS))</pre>
vin sum <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(Preqinclean$PRIMARY.REGION.FOCUS)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
  {
    vin sum[i,j]=
sum(Preqinclean$FUND.SIZE..USD.MN.[intersect(which(Preqinclean$VINTAGE...INCE
PTION.YEAR==(1968+i)
,which(Preginclean$PRIMARY.REGION.FOCUS==pick[j]))])/1000
}
# above code creates a two variable (i,j) loop to populate the vintage year
summation matrix. The vin_sum[i,j] refers to the matrix being populated. The
loop uses various which() functions which pick out certain variables. The
first which loop for i and j = 1 picks the list of African funds (by Primary
region focus) that were stared in 1969. Once this list of creates it is run
through the second which loop that picks the corresponding fund sizes. The
loop then runs again for i and j for all vintage year dates and primary
region focus areas.
colnames(vin sum) <- c(pick[1:length(x)])</pre>
rownames(vin sum) <- c(1969:2017)</pre>
# Plot of the market size by Primary region focus - not cumulative.
```

```
df <- data.frame(vintage = 1969:2017, vin_sum)
df <- reshape2::melt(df, id.vars = 'vintage',variable.name =
'Primary_Region_Focus')
plot1 <- ggplot(df, aes(vintage,value)) + geom_line(aes(colour =
Primary_Region_Focus)) + ggtitle("Total New Market Issuances by Primary
Region Focus USD bln(.)") + theme(legend.position = "none" +
scale_colour_manual(values=c(pick[1:k])))
plot1</pre>
```

Total New Market Issuances by Primary Region Focus



Export to Powerpoint

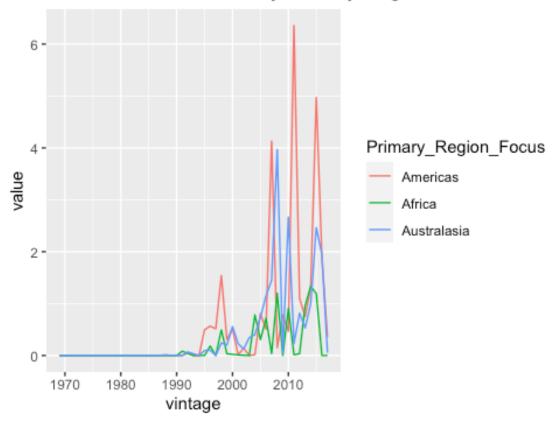
```
p_dml <- rvg::dml(ggobj = plot1)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "1.pptx"
```

```
)
```

Vintage Geo map code restricted to smaller focus regions.

```
restricted <- c("Americas", "Africa", "Australasia")</pre>
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)</pre>
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1</pre>
k <- length(unique(restricted))</pre>
vin_sum <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(restricted)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
  {
    vin sum[i,j]=
sum(Preginclean$FUND.SIZE..USD.MN.[intersect(which(Preginclean$VINTAGE...INCE
PTION.YEAR==(1968+i))
,which(Preqinclean$PRIMARY.REGION.FOCUS==pick[j]))])/1000
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin_sum) <- c(1969:2017)</pre>
# Plot of the market size by Primary region focus - not cumulative.
df <- data.frame(vintage = 1969:2017, vin sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Primary Region Focus')
plot2 <- ggplot(df, aes(vintage, value)) + geom line(aes(colour =</pre>
Primary_Region_Focus)) + ggtitle("New Market Issuances by Primary Region
Focus - Restricted - USD bln(.)") + theme(legend.position = "none" +
scale colour manual(values=c(pick[1:k])))
plot2
```

New Market Issuances by Primary Region Focus - Restri



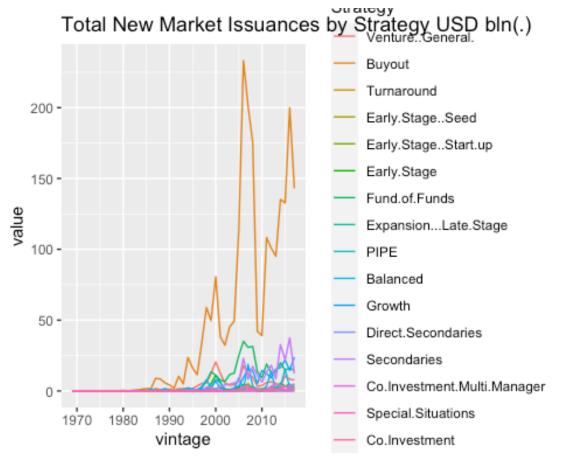
Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot2)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "2.pptx"
    )
)
```

Vintage Strategy map code.

```
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(Preqinclean$STRATEGY))
vin_sum <- matrix(NA, nrow=n, ncol=k)
# base matrix for later population</pre>
```

```
x <- unique(Preginclean$STRATEGY)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
  {
    vin sum[i,j]=
sum(Preqinclean$FUND.SIZE..USD.MN.[intersect(which(Preqinclean$VINTAGE...INCE
PTION.YEAR==(1968+i)) ,which(Preqinclean$STRATEGY==pick[j]))])/1000
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin_sum) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name = 'Strategy')</pre>
plot3 <- ggplot(df, aes(vintage,value)) + geom_line(aes(colour = Strategy)) +</pre>
ggtitle("Total New Market Issuances by Strategy USD bln(.)") +
theme(legend.position = "none" + scale colour manual(values=c(pick[1:k])))
plot3
```



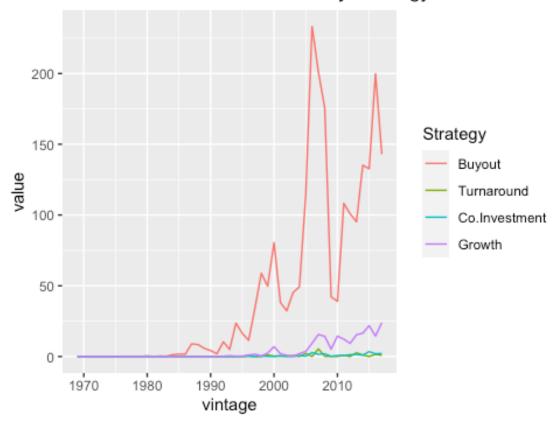
Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot3)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "3.pptx"
    )
)
```

Vintage Strategy map code restricted.

```
restricted <- c("Buyout", "Co-Investment", "Growth", "Turnaround")</pre>
min <- min(Preginclean$VINTAGE...INCEPTION.YEAR)</pre>
n <- max(Preginclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(restricted))</pre>
vin_sum <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(restricted)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
    vin sum[i,j]=
sum(Preginclean$FUND.SIZE..USD.MN.[intersect(which(Preginclean$VINTAGE...INCE
PTION.YEAR==(1968+i)) ,which(Preqinclean$STRATEGY==pick[j]))])/1000
}
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin sum) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name = 'Strategy')</pre>
plot4 <- ggplot(df, aes(vintage,value)) + geom line(aes(colour = Strategy)) +</pre>
ggtitle("Total New Market Issuances by Strategy - Restricted - USD bln(.)") +
theme(legend.position = "none" + scale_colour_manual(values=c(pick[1:k])))
plot4
```

Total New Market Issuances by Strategy - Restricted - I



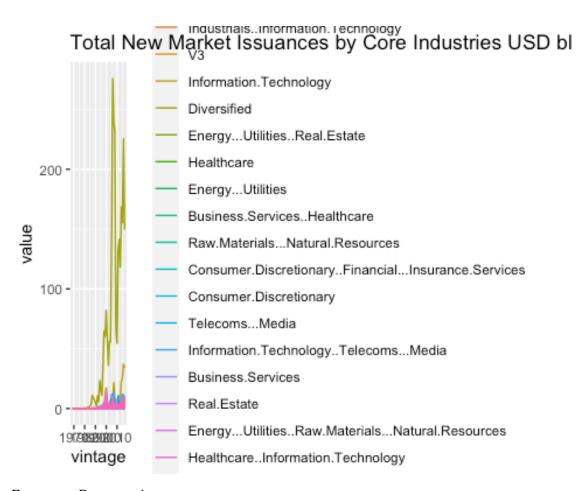
Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot4)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and Location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "4.pptx"
    )
)
```

Vintage Core Industries map code.

```
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(Preqinclean$CORE.INDUSTRIES))
vin_sum <- matrix(NA, nrow=n, ncol=k)
# base matrix for later population</pre>
```

```
x <- unique(Preginclean$CORE.INDUSTRIES)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
 {
    vin_sum[i,j]=
sum(Preqinclean$FUND.SIZE..USD.MN.[intersect(which(Preqinclean$VINTAGE...INCE
PTION.YEAR==(1968+i)) ,which(Preqinclean$CORE.INDUSTRIES==pick[j]))])/1000
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin_sum) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Core Industries')
plot5 <- ggplot(df, aes(vintage, value)) + geom_line(aes(colour =</pre>
Core_Industries)) + ggtitle("Total New Market Issuances by Core Industries
USD bln(.)") + theme(legend.position = "none" +
scale_colour_manual(values=c(pick[1:k])))
plot5
```



Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot5)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "5.pptx"
    )
)
```

Vintage Core Industries map code. Restricted

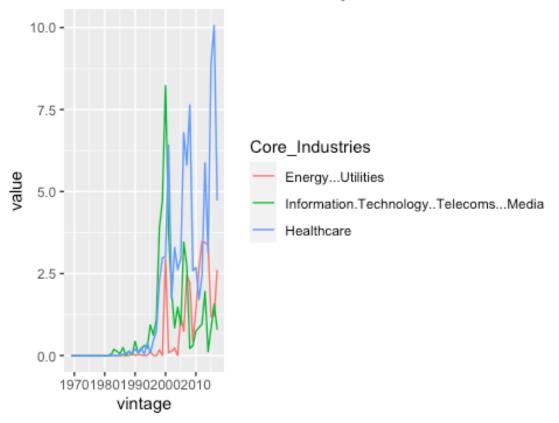
```
restricted <- c("Healthcare", "Energy & Utilities", "Information Technology,
Telecoms & Media")

min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)

n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(restricted))</pre>
```

```
vin_sum <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(restricted)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
    vin_sum[i,j]=
sum(Preqinclean$FUND.SIZE..USD.MN.[intersect(which(Preqinclean$VINTAGE...INCE
PTION.YEAR==(1968+i)) ,which(Preqinclean$CORE.INDUSTRIES==pick[j]))])/1000
}
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin_sum) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Core Industries')
plot6 <- ggplot(df, aes(vintage, value)) + geom_line(aes(colour =</pre>
Core Industries)) + ggtitle("Total New Market Issuances by Core Industries -
Restricted - USD bln(.)") + theme(legend.position = "none" +
scale colour manual(values=c(pick[1:k])))
plot6
```

Total New Market Issuances by Core Industries - Restr



Export to Powerpoint

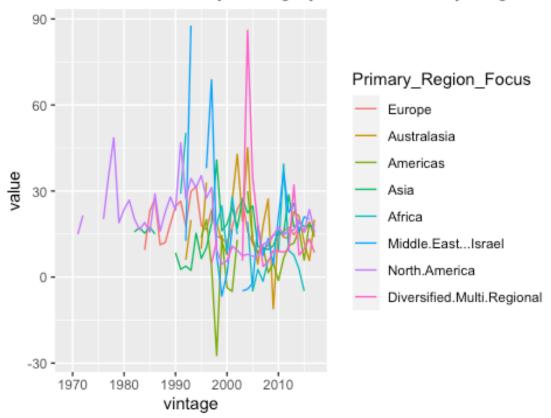
```
p_dml <- rvg::dml(ggobj = plot6)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "6.pptx"
    )
)
```

Vintage Geo map code by IRR. Unrestricted.

```
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(unique(Preqinclean$PRIMARY.REGION.FOCUS))
vin_mean <- matrix(NA, nrow=n, ncol=k)
# base matrix for later population</pre>
```

```
x <- unique(Preqinclean$PRIMARY.REGION.FOCUS)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
  {
    vin_mean[i,j]=
mean(Preqinclean$NET.IRR....[intersect(which(Preqinclean$VINTAGE...INCEPTION.
YEAR==(1968+i)) ,which(Preqinclean$PRIMARY.REGION.FOCUS==pick[j]))])
}
colnames(vin_mean) <- c(pick[1:length(x)])</pre>
rownames(vin_mean) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_mean)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Primary_Region_Focus')
plot7 <- ggplot(df, aes(vintage, value)) + geom_line(aes(colour =</pre>
Primary_Region_Focus)) + ggtitle("Mean Fund IRR by Vintage year and Primary
Region Focus") + theme(legend.position = "none" +
scale_colour_manual(values=c(pick[1:k])))
plot7
## Warning: Removed 119 row(s) containing missing values (geom path).
```

Mean Fund IRR by Vintage year and Primary Region Fo



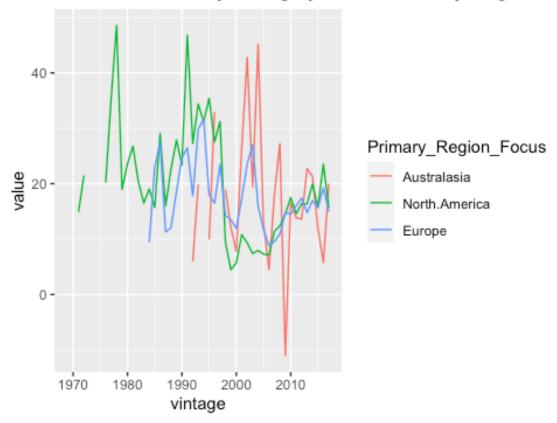
Export to Powerpoint

Vintage Geo map code by IRR.Restricted.

```
restricted <- c("North America", "Australasia", "Europe")
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(restricted)</pre>
```

```
vin_mean <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(restricted)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
    vin_mean[i,j]=
mean(Preqinclean$NET.IRR....[intersect(which(Preqinclean$VINTAGE...INCEPTION.
YEAR==(1968+i)) ,which(Preqinclean$PRIMARY.REGION.FOCUS==pick[j]))])
}
}
colnames(vin_mean) <- c(pick[1:length(x)])</pre>
rownames(vin_mean) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_mean)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Primary_Region_Focus')
plot8 <- ggplot(df, aes(vintage, value)) + geom_line(aes(colour =</pre>
Primary Region Focus)) + ggtitle("Mean Fund IRR by Vintage year and Primary
Region Focus") + theme(legend.position = "none" +
scale colour manual(values=c(pick[1:k])))
plot8
## Warning: Removed 34 row(s) containing missing values (geom_path).
```

Mean Fund IRR by Vintage year and Primary Region Fo



Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot8)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "8.pptx"
    )
)

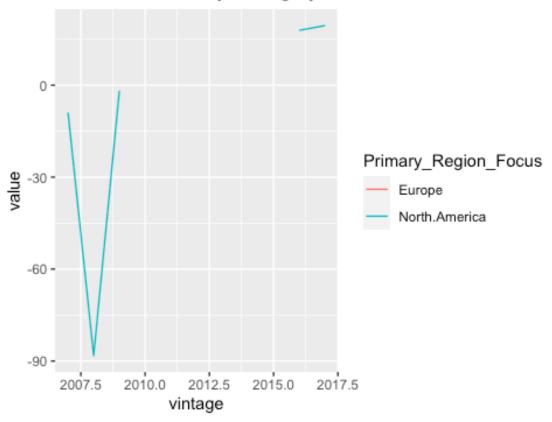
## Warning: Removed 34 row(s) containing missing values (geom_path).
```

Vintage Geo map code by IRR. Restricted. Impact only.

```
restricted <- c("North America", "Europe")
impact_only<- read.csv('impactonly.csv', row.names = NULL, stringsAsFactors =
FALSE)
min <- min(impact_only$VINTAGE...INCEPTION.YEAR)</pre>
```

```
n <- max(impact only$VINTAGE...INCEPTION.YEAR)-min + 1
k <- length(restricted)</pre>
vin mean <- matrix(NA, nrow=n, ncol=k)</pre>
# base matrix for later population
x <- unique(restricted)</pre>
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
    vin mean[i,j]=
mean(impact only$NET.IRR....[intersect(which(impact only$VINTAGE...INCEPTION.
YEAR==(min-1+i)) ,which(impact_only$PRIMARY.REGION.FOCUS==pick[j]))])
}
colnames(vin_mean) <- c(pick[1:length(x)])</pre>
rownames(vin mean) <- c(min:max(impact only$VINTAGE...INCEPTION.YEAR))</pre>
df <- data.frame(vintage = min:max(impact_only$VINTAGE...INCEPTION.YEAR),</pre>
vin mean)
df <- reshape2::melt(df, id.vars = 'vintage', variable.name =</pre>
'Primary Region Focus')
plot9 <-ggplot(df, aes(vintage,value)) + geom line(aes(colour =</pre>
Primary Region Focus)) + ggtitle("Mean Fund IRR by Vintage year -
Restricted") + theme(legend.position = "none" +
scale colour manual(values=c(pick[1:k])))
plot9
## Warning: Removed 4 row(s) containing missing values (geom_path).
```

Mean Fund IRR by Vintage year - Restricted



Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot9)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "9.pptx"
    )
)

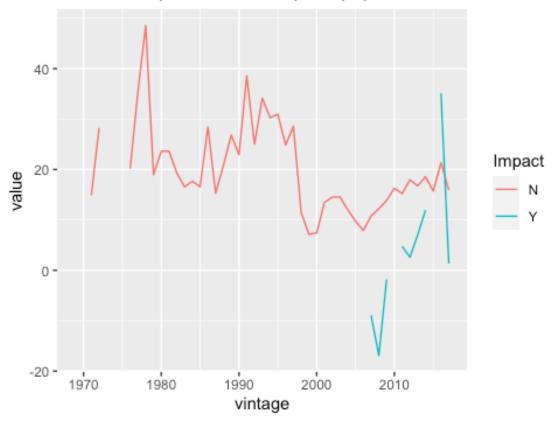
### Warning: Removed 4 row(s) containing missing values (geom_path).
```

Impact vs non Impact.

```
min <- min(Preqinclean$VINTAGE...INCEPTION.YEAR)
n <- max(Preqinclean$VINTAGE...INCEPTION.YEAR)-min + 1
k <- 2
vin_sum <- matrix(NA, nrow=n, ncol=k)</pre>
```

```
# base matrix for later population
x <- c("N","Y")
pick <- sample(x, length(x), replace = FALSE, prob = NULL)</pre>
# above code picks random categorical data from primary region focus without
replacement.
for(j in 1:k){
for(i in 1:n)
  {
    vin sum[i,j]=
mean(Preqinclean$NET.IRR....[intersect(which(Preqinclean$VINTAGE...INCEPTION.
YEAR==(1968+i)) ,which(Preqinclean$Impact...Y.N...==pick[j]))])
}
}
colnames(vin_sum) <- c(pick[1:length(x)])</pre>
rownames(vin_sum) <- c(1969:2017)</pre>
df <- data.frame(vintage = 1969:2017, vin_sum)</pre>
df <- reshape2::melt(df, id.vars = 'vintage', variable.name = 'Impact')</pre>
plot10 <- ggplot(df, aes(vintage, value)) + geom_line(aes(colour = Impact)) +</pre>
ggtitle("Net IRR Impact vs non-Impact (%)") + theme(legend.position = "none"
+ scale colour manual(values=c(pick[1:k])))
plot10
## Warning: Removed 38 row(s) containing missing values (geom_path).
```

Net IRR Impact vs non-Impact (%)



Export to Powerpoint

```
p_dml <- rvg::dml(ggobj = plot10)
# initialize PowerPoint slide ----
officer::read_pptx() %>%
    # add slide ----
    officer::add_slide() %>%
    # specify object and location of object ----
    officer::ph_with(p_dml, ph_location()) %>%
    # export slide -----
base::print(
    target = here::here(
        "10.pptx"
    )
)

## Warning: Removed 38 row(s) containing missing values (geom_path).
```

Estimation.

```
est_tableIN<- read.csv('est_tableIN.csv', row.names = NULL, stringsAsFactors
= FALSE)

colnames(est_tableIN) <- c("Estimate No.", "Fund Size (US mn)", "Vintage")</pre>
```

```
Year","Impact Fund?","Strategy","Prim Focus","Estimate")
coeff_table <- matrix(NA, nrow=length(names(ols_no_int1$coefficients)),</pre>
ncol=4)
# matrix for coefficent names
colnames(coeff_table) <- c("Coefficient Name", "Coefficient Model 1</pre>
Value", "Coefficient Pr(>|t|)", "Signif. code: 0 '***' 0.001 '**' 0.05
(.' 0.1 (' 1")
rownames(coeff table) <- 1:length(names(ols no int1$coefficients))</pre>
coeff_table[,1] <- names(ols_no_int1$coefficients)</pre>
coeff_table[,2] = ols_no_int1$coefficients
coeff table[,3] = summary(ols no int1)$coefficients[,4]
for(i in 1:length(names(ols no int1$coefficients)))
    if((coeff table[i,3]<0.1) && (coeff table[i,3]>0.05) )#regime1end is the
last date where the regime is 1
      coeff_table[i,4]="."
}
for(i in 1:length(names(ols_no_int1$coefficients)))
    if((coeff_table[i,3]<0.05) && (coeff_table[i,3]>0.01) )#regime1end is the
last date where the regime is 1
    {
      coeff_table[i,4]="*"
}
for(i in 1:length(names(ols_no_int1$coefficients)))
    if((coeff_table[i,3]<0.01) && (coeff_table[i,3]>0.001) )#regime1end is
the last date where the regime is 1
      coeff_table[i,4]="**"
    }
}
for(i in 1:length(names(ols no int1$coefficients)))
    if(coeff_table[i,3]<0.001)</pre>
      coeff_table[i,4]="***"
}
```

```
for(i in 1:length(names(ols_no_int1$coefficients)))
{
    if(coeff_table[i,3]>0.1)
    {
      coeff_table[i,4]=""
}
unique(coeff_table[,1])
    [1] "Preginclean$FUND.SIZE..USD.MN."
##
##
    [2] "Vintage1969"
##
    [3] "Vintage1971"
    [4] "Vintage1972"
##
##
    [5] "Vintage1976"
##
    [6] "Vintage1977"
##
    [7] "Vintage1978"
##
    [8]
        "Vintage1979"
##
    [9] "Vintage1980"
## [10] "Vintage1981"
## [11] "Vintage1982"
## [12] "Vintage1983"
## [13]
        "Vintage1984"
## [14] "Vintage1985"
## [15] "Vintage1986"
## [16] "Vintage1987"
## [17]
        "Vintage1988"
## [18] "Vintage1989"
## [19] "Vintage1990"
## [20] "Vintage1991"
## [21] "Vintage1992"
## [22]
       "Vintage1993"
## [23] "Vintage1994"
        "Vintage1995"
## [24]
## [25] "Vintage1996"
## [26] "Vintage1997"
## [27] "Vintage1998"
## [28] "Vintage1999"
## [29] "Vintage2000"
## [30] "Vintage2001"
## [31]
        "Vintage2002"
## [32] "Vintage2003"
## [33] "Vintage2004"
## [34] "Vintage2005"
## [35] "Vintage2006"
## [36] "Vintage2007"
## [37] "Vintage2008"
## [38] "Vintage2009"
## [39] "Vintage2010"
```

```
## [40] "Vintage2011"
## [41] "Vintage2012"
## [42] "Vintage2013"
## [43] "Vintage2014"
## [44] "Vintage2015"
## [45] "Vintage2016"
## [46] "Vintage2017"
## [47] "Preqin_mod$ImpactDUM"
## [48] "Preqin_mod$Early_StageDUM"
## [49] "Preqin_mod$GrowthDUM"
## [50] "Preqin mod$BuyoutDUM"
## [51] "Preqin_mod$Fund_of_FundsDUM"
## [52] "Preqin mod$Venture GeneralDUM"
## [53] "Preqin_mod$Early_Stage_SeedDUM"
## [54] "Preqin mod$Co InvestmentDUM"
## [55] "Preqin_mod$North_America_DUM"
## [56] "Preqin_mod$Europe_DUM"
## [57] "Preqin_mod$Asia_DUM"
## [58] "Preqin_mod$Diversified_Multi_Regional_DUM"
## [59] "Preqin mod$Americas DUM"
## [60] "Preqin_mod$Africa_DUM"
## [61] "Preqin_mod$Middle_East_and_Israel_DUM"
for(i in 1:6)
{
est tableIN[i,7] <-
as.numeric(coeff_table[which(coeff_table[,1]=="Preqinclean$FUND.SIZE..USD.MN.
"),2])*est_tableIN[i,2] +
ols no int1$coefficients[which(names(ols no int1$coefficients)==est tableIN[i
,3])] + if(est_tableIN[i,4]=="Y"){impact <-</pre>
ols no int1$coefficients[which(names(ols no int1$coefficients)=="Preqin mod$I
mpactDUM")] }else{impact=0} +
ols no int1$coefficients[which(names(ols no int1$coefficients)==est tableIN[i
+ [([5,
ols no int1$coefficients[which(names(ols no int1$coefficients)==est tableIN[i
,6])]
}
write.csv(coeff_table, "coeff_table.csv")
#save matrix
write.csv(est tableIN, "Estimation.csv")
Dummy variables for Core Industries.
n <- length(Preqinclean$NET.IRR....)</pre>
```

k <- length(unique(Preqinclean\$CORE.INDUSTRIES))</pre>

core_DUM <- matrix(NA, nrow=n, ncol=k)
base matrix for later population</pre>

```
x <- unique(Preqinclean$CORE.INDUSTRIES)
pick <- sample(x, length(x), replace = FALSE, prob = NULL)
# above code picks random categorical data from primary region focus without
replacement.

for(j in 1:k){
    for(i in 1:n)
    {
        if(Preqinclean$CORE.INDUSTRIES[i] == pick[j])
        {
            core_DUM[i,j]=1
        }
        else
        {
            core_DUM[i,j]=0
        }
}
colnames(core_DUM) <- c(pick[1:length(x)])</pre>
```

Adding numericals

```
numericals <- matrix(NA, nrow=n, ncol=5)</pre>
for(i in 1:n)
{
    if(Preginclean$NET.MULTIPLE..X.[i]=="n/a")
      numericals[i,1]=1
    }
   else
      numericals[i,1]=Preqinclean$NET.MULTIPLE..X.[i]
    }
  if(Preqinclean$RVPI....[i]=="n/a")
      numericals[i,2]=0
   else
      numericals[i,2]=Preqinclean$RVPI....[i]
    }
  if(Preqinclean$DPI....[i]=="n/a")
      numericals[i,3]=0
   else
```

```
{
    numericals[i,3]=Preqinclean$DPI....[i]
}

if(Preqinclean$CALLED....[i]=="n/a")
    {
        numericals[i,4]=100
    }
    else
    {
            numericals[i,4]=Preqinclean$CALLED....[i]
        }

if(Preqinclean$MEDIAN.BENCHMARK.NET.IRR....[i]=="n/a")
        {
            numericals[i,5]=10
        }
        else
        {
            numericals[i,5]=Preqinclean$MEDIAN.BENCHMARK.NET.IRR....[i]
        }
}
```

Making the model dataset

```
Preqin_mod <- cbind(Preqin_mod,Vintage_all,core_DUM)
Preqin_mod <- as.matrix(Preqin_mod)
subset <- c(1:2,9:25,27:28,32:93)
Preqin_mod <- Preqin_mod[,subset]
Preqin_mod <- cbind(Preqin_mod,numericals)
write.csv(Preqin_mod,'Preqin_mod.csv')
write.csv(ols_no_int1$coefficients,'ols_no_int1$coefficients.csv')</pre>
```