

Independent Study

```
library(readxl)
library(openxlsx)
library(tidyr)
library(stringr)
library(tibble)
library(car)
```

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

```
library(showtext)
```

```
## Loading required package: sysfonts
```

```
## Loading required package: showtextdb
```

```
library(dplyr)
```

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

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

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

```
## The following objects are masked from 'package:base':
##      intersect, setdiff, setequal, union
```

```
library(CCP)
library(readxl)
library(ggcorrplot)
```

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

```
library(factoextra)
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WB  
a
```

```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
library(glmnet)
```

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

```
##  
## Attaching package: 'Matrix'
```

```
## The following objects are masked from 'package:tidyverse':  
##  
##     expand, pack, unpack
```

```
## Loaded glmnet 4.1-6
```

```
library(factoextra)  
library(FactoMineR)  
require(FactoMineR)  
require(factoextra)  
setwd("/Users/jettadler/Desktop/Data /Study")
```

```
#Defining Data  
rm(list = ls())  
stl_raw <- read_xls('steel_data.xls')  
al_raw <- read_xls('Al_data.xls')
```

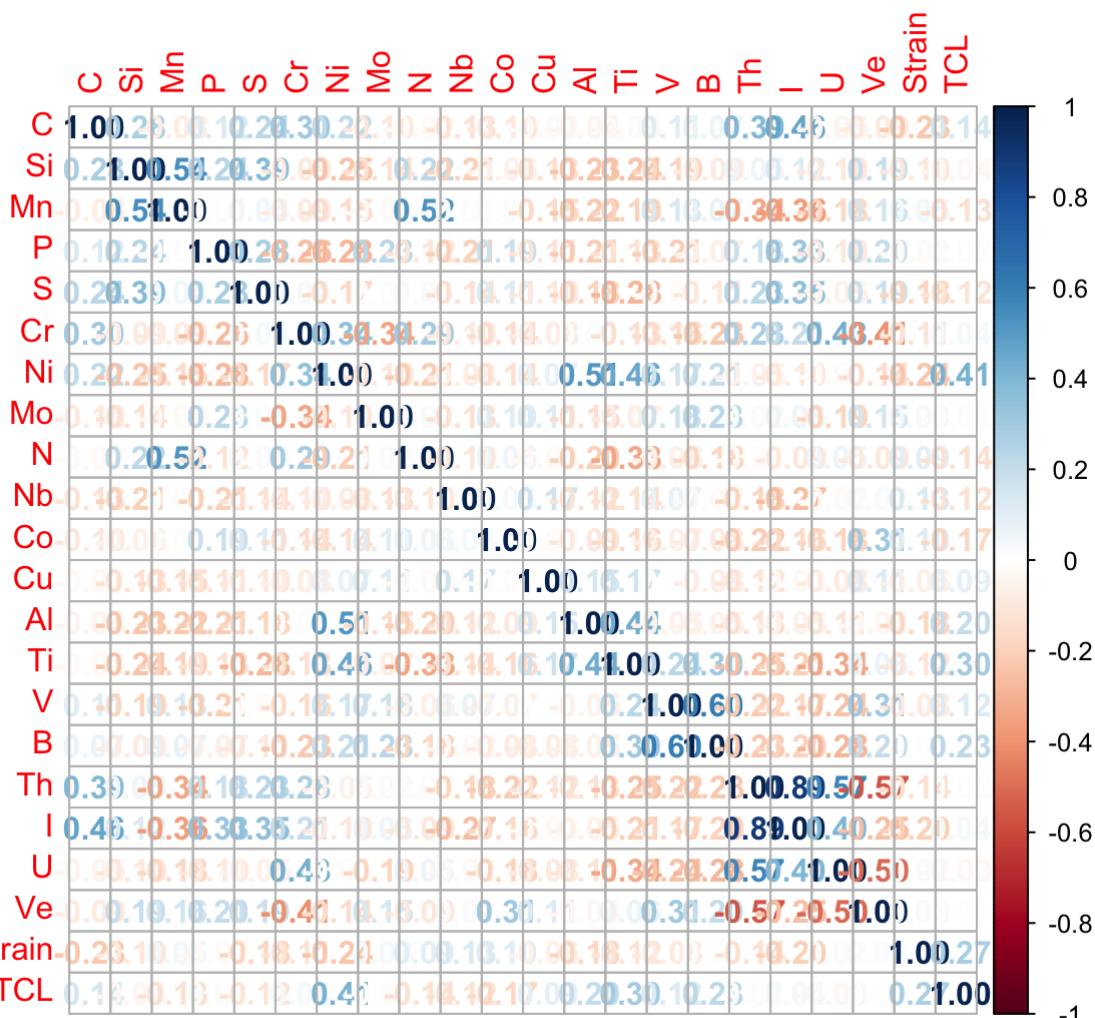
```
#removing blank rows and name/number coulombs  
toDelete <- seq(1, 974, 2)  
stl <- stl_raw[-toDelete, ]  
al <- al_raw[-toDelete, ]  
stl <- stl[c(-1)]  
al <- al[c(-1)]
```

```
#Separating Data into Elemental Compositons, Strain, and Crack Length Data
```

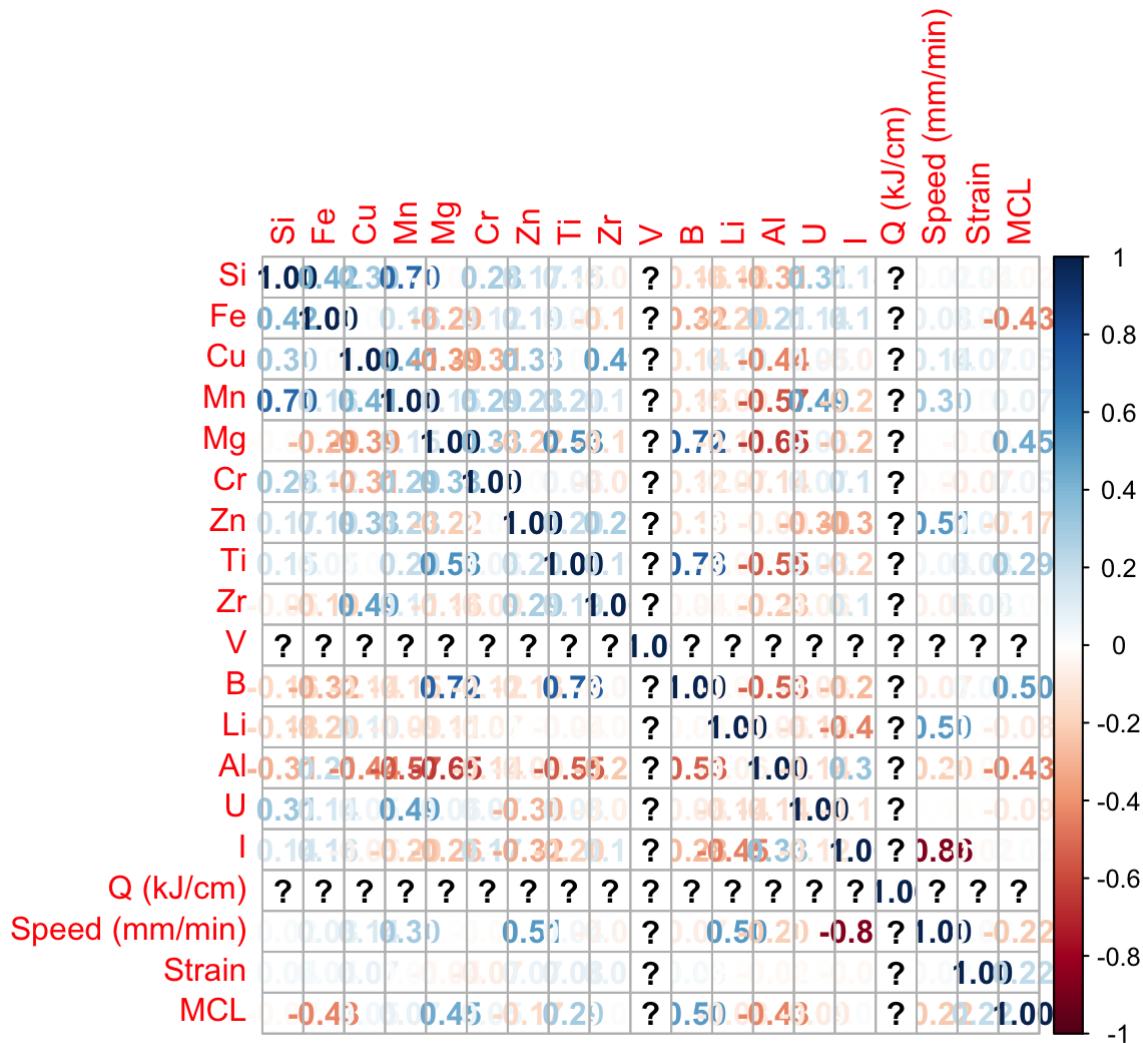
```
stEl <- stl[c(1:16)]
alEl <- al[c(1:13)]
stSTN <- stl[c(21)]
stTCL <- stl[c(22)]
alSTN <- al[c(18)]
alMCL <- al[c(19)]
```

```
#First Visual check of possible correlations between variables
```

```
cors <- cor(stl)
cora <- cor(al)
corrplot(cors, method='number')
```



```
corrplot(cora, method='number')
```



cors

##	C	Si	Mn	P	S	Cr
## C	1.00000000	0.27845666	-0.07746384	0.12421010	0.23829076	0.30252074
## Si	0.27845666	1.00000000	0.54288940	0.24403814	0.38630488	-0.08870044
## Mn	-0.07746384	0.54288940	1.00000000	-0.01076547	0.04041910	-0.08914258
## P	0.12421010	0.24403814	-0.01076547	1.00000000	0.28347200	-0.26479798
## S	0.23829076	0.38630488	0.04041910	0.28347200	1.00000000	0.03644738
## Cr	0.30252074	-0.08870044	-0.08914258	-0.26479798	0.03644738	1.00000000
## Ni	0.21548779	-0.24832062	-0.14784543	-0.28337310	-0.16648652	0.34011443
## Mo	-0.09549286	-0.13819168	-0.02845569	0.23045744	0.01192515	-0.34391819
## N	-0.02481124	0.22103907	0.52387534	-0.11579085	0.02805146	0.29194807
## Nb	-0.12579354	-0.20581200	0.01201195	-0.20526057	-0.13866867	-0.09773600
## Co	-0.09669215	-0.05572159	0.01495095	0.18950145	0.11394630	-0.14301930
## Cu	-0.02918592	-0.13300644	-0.15206014	-0.11422006	-0.09795553	-0.07875766
## Al	-0.04188047	-0.23260626	-0.22424775	-0.21209939	-0.18421137	0.01428158
## Ti	-0.02531531	-0.24281444	-0.19008803	-0.12976077	-0.28308274	-0.13446150
## V	0.10629471	-0.18519455	0.13304538	-0.21233624	-0.01149909	-0.15701641
## B	0.06647661	-0.08702766	0.07066607	-0.06543854	-0.10817779	-0.22581597
## Th	0.39039469	0.05270954	-0.33500554	0.18192163	0.23337489	0.27515468
## I	0.46477037	0.11991791	-0.36261756	0.32546704	0.35347395	0.20792318
## U	-0.04636266	-0.09558830	-0.17551355	-0.09585409	-0.05185645	0.42670047
## Ve	-0.08895166	0.19076415	0.15947449	0.20161899	0.18986257	-0.40924212
## Strain	-0.22771168	-0.10258401	0.04880527	-0.02243995	-0.18159471	-0.11390472
## TCL	0.14411528	-0.03647010	-0.13099629	0.01561729	-0.11962175	0.03810476
##	Ni	Mo	N	Nb	Co	Cu
## C	0.21548779	-0.09549286	-0.02481124	-0.12579354	-0.09669215	-0.02918592
## Si	-0.24832062	-0.13819168	0.22103907	-0.20581200	-0.05572159	-0.13300644
## Mn	-0.14784543	-0.02845569	0.52387534	0.01201195	0.01495095	-0.15206014
## P	-0.28337310	0.23045744	-0.11579085	-0.20526057	0.18950145	-0.11422006
## S	-0.16648652	0.01192515	0.02805146	-0.13866867	0.11394630	-0.09795553
## Cr	0.34011443	-0.34391819	0.29194807	-0.09773600	-0.14301930	-0.07875766
## Ni	1.00000000	-0.10408396	-0.21280034	-0.07952737	-0.13510072	0.07337490
## Mo	-0.10408396	1.00000000	0.02217061	-0.13280162	0.10478750	0.11093110
## N	-0.21280034	0.02217061	1.00000000	-0.10490213	0.05665789	-0.02548007
## Nb	-0.07952737	-0.13280162	-0.10490213	1.00000000	0.03159128	0.16736976
## Co	-0.13510072	0.10478750	0.05665789	0.03159128	1.00000000	-0.01809023
## Cu	0.07337490	0.11093111	-0.02548007	0.16736977	-0.01809023	1.00000000
## Al	0.51472603	-0.14673019	-0.20282819	-0.12249223	-0.09141180	0.14678152
## Ti	0.45899169	-0.07394976	-0.33377034	-0.13915043	-0.15716960	0.16767656
## V	0.16933219	0.17510812	-0.06142697	0.07451321	-0.06661316	0.00261103
## B	0.20741000	0.23308727	-0.17500646	-0.02469812	-0.08367700	-0.07896361
## Th	-0.05261927	0.02448561	-0.01815577	-0.18477329	-0.21509654	-0.11765185
## I	-0.09658742	0.05869663	-0.09046160	-0.26644977	-0.15674378	-0.02997228
## U	-0.01035120	-0.18796257	0.05176411	-0.02213752	-0.18049103	-0.08173343
## Ve	-0.14203631	0.14721887	-0.09227065	0.02429409	0.30537700	0.11340314
## Strain	-0.23926109	0.03004494	0.09036873	0.12658403	0.10194218	-0.06497995
## TCL	0.41290930	-0.01386736	-0.13861387	-0.12135101	-0.17329686	0.08959043
##	Al	Ti	V	B	Th	
## C	-0.04188047	-0.02531531	0.106294712	0.0664766134	0.39039469	
## Si	-0.23260626	-0.24281444	-0.18519454	-0.0870276593	0.05270954	
## Mn	-0.22424775	-0.19008803	0.133045383	0.0706660690	-0.33500554	
## P	-0.21209939	-0.12976077	-0.212336238	-0.0654385438	0.18192163	
## S	-0.18421137	-0.28308274	-0.011499094	-0.1081777913	0.23337489	

	I	U	Ve	Strain	TCL
## Cr	0.01428158	-0.13446150	-0.157016408	-0.2258159695	0.27515468
## Ni	0.51472603	0.45899169	0.169332190	0.2074099994	-0.05261927
## Mo	-0.14673019	-0.07394976	0.175108125	0.2330872675	0.02448561
## N	-0.20282819	-0.33377034	-0.061426969	-0.1750064628	-0.01815577
## Nb	-0.12249223	-0.13915043	0.074513210	-0.0246981243	-0.18477329
## Co	-0.09141180	-0.15716960	-0.066613158	-0.0836769967	-0.21509654
## Cu	0.14678153	0.16767656	0.002611036	-0.0789636148	-0.11765186
## Al	1.00000000	0.43984444	-0.053377976	-0.0535430587	-0.12812092
## Ti	0.43984444	1.00000000	0.239107468	0.3038814520	-0.25461669
## V	-0.05337798	0.23910747	1.000000000	0.6001763018	-0.21830014
## B	-0.05354306	0.30388145	0.600176302	1.000000000	-0.22515558
## Th	-0.12812092	-0.25461669	-0.218300141	-0.2251555817	1.00000000
## I	-0.05461621	-0.20797505	-0.172939417	-0.1962436393	0.89051845
## U	-0.11431258	-0.33850179	-0.242033956	-0.2811999863	0.57241764
## Ve	-0.03926660	0.06426317	0.309047426	0.2020865864	-0.56769956
## Strain	-0.17956990	-0.11546403	-0.080976829	-0.0007026004	-0.13595857
## TCL	0.20227094	0.29650669	0.115898303	0.2297429217	0.01977808
##					
## C	0.46477037	-0.04636266	-0.088951656	-0.2277116838	0.144115282
## Si	0.11991791	-0.09558830	0.190764151	-0.1025840072	-0.036470102
## Mn	-0.36261756	-0.17551355	0.159474494	0.0488052736	-0.130996289
## P	0.32546704	-0.09585409	0.201618986	-0.0224399461	0.015617293
## S	0.35347395	-0.05185645	0.189862565	-0.1815947086	-0.119621747
## Cr	0.20792318	0.42670047	-0.409242123	-0.1139047217	0.038104765
## Ni	-0.09658742	-0.01035120	-0.142036314	-0.2392610946	0.412909297
## Mo	0.05869663	-0.18796257	0.147218868	0.0300449419	-0.013867362
## N	-0.09046160	0.05176411	-0.092270649	0.0903687260	-0.138613869
## Nb	-0.26644977	-0.02213752	0.024294094	0.1265840319	-0.121351014
## Co	-0.15674378	-0.18049103	0.305376999	0.1019421843	-0.173296857
## Cu	-0.02997228	-0.08173343	0.113403144	-0.0649799572	0.089590436
## Al	-0.05461621	-0.11431258	-0.039266597	-0.1795698951	0.202270939
## Ti	-0.20797505	-0.33850179	0.064263167	-0.1154640284	0.296506692
## V	-0.17293942	-0.24203396	0.309047426	-0.0809768292	0.115898303
## B	-0.19624364	-0.28119999	0.202086586	-0.0007026004	0.229742922
## Th	0.89051845	0.57241764	-0.567699564	-0.1359585695	0.019778081
## I	1.00000000	0.40114754	-0.250248230	-0.2020144197	0.036293631
## U	0.40114754	1.00000000	-0.502137485	0.0242652899	-0.033752736
## Ve	-0.25024823	-0.50213749	1.000000000	0.0274623077	-0.004297735
## Strain	-0.20201442	0.02426529	0.027462308	1.000000000	0.270733180
## TCL	0.03629363	-0.03375274	-0.004297735	0.2707331797	1.000000000

cora

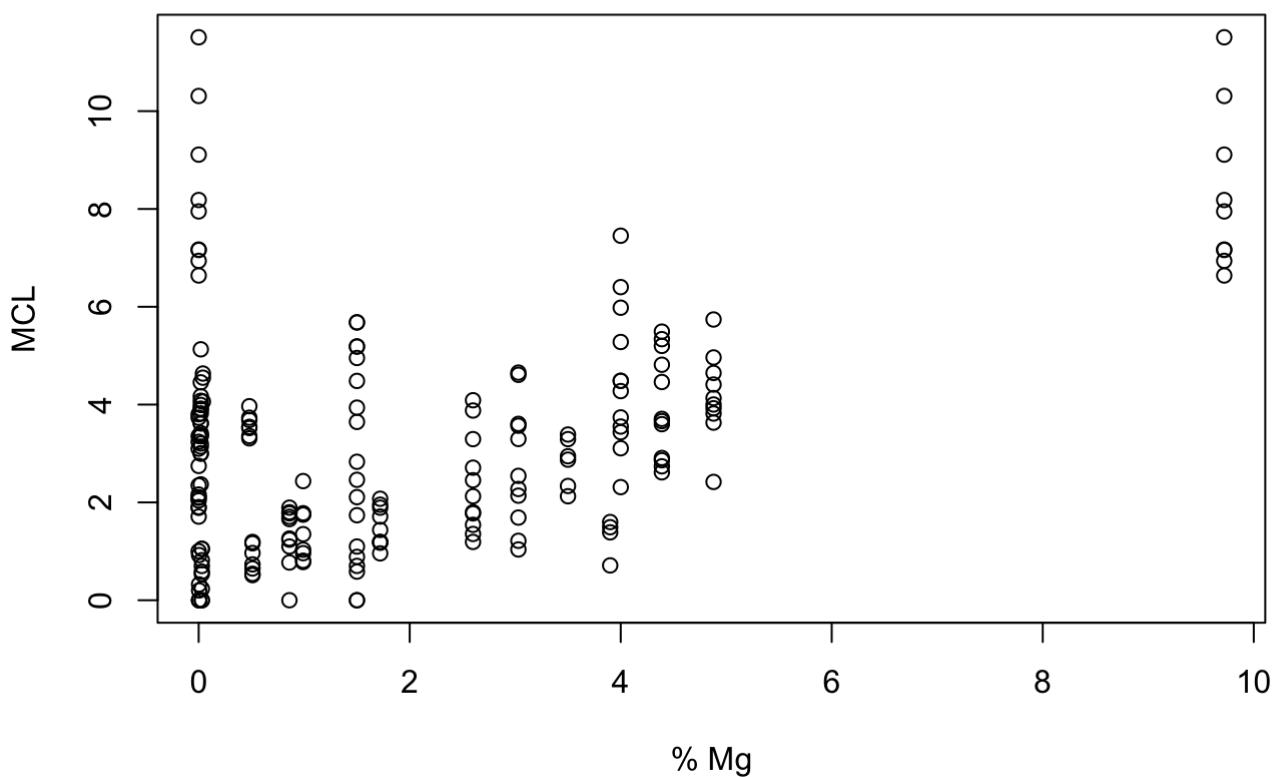
##	Si	Fe	Cu	Mn	Mg	
## Si	1.00000000	0.41946565	0.30371090	0.69985686	-0.01679518	
## Fe	0.41946565	1.00000000	0.01367361	0.16147267	-0.28617836	
## Cu	0.30371090	0.01367361	1.00000000	0.41177930	-0.38803728	
## Mn	0.69985686	0.16147267	0.41177930	1.00000000	0.14739982	
## Mg	-0.01679518	-0.28617836	-0.38803728	0.14739982	1.00000000	
## Cr	0.28442177	0.11760990	-0.30875105	0.28998191	0.33209597	
## Zn	0.16886697	0.18712629	0.32661512	0.23479904	-0.21757192	
## Ti	0.14720438	0.04773796	0.01882995	0.20480612	0.53313617	
## Zr	-0.05186117	-0.18685554	0.48864377	0.11018048	-0.15693241	
## V	NA	NA	NA	NA	NA	
## B	-0.16233589	-0.31560459	-0.13845711	-0.15064796	0.71706925	
## Li	-0.17604585	-0.20347178	0.10138874	-0.08560685	-0.11009762	
## Al	-0.30722007	0.21015028	-0.43847668	-0.56725161	-0.64583767	
## U	0.31146101	0.14376163	0.05113820	0.48994437	0.06454867	
## I	0.13730297	0.16278287	-0.05140019	-0.20438381	-0.25639121	
## Q (kJ/cm)	NA	NA	NA	NA	NA	
## Speed (mm/min)	0.02486073	0.07786063	0.13653407	0.29828949	0.00980461	
## Strain	0.03654705	0.03700105	0.07475819	0.01044924	-0.03762740	
## MCL	-0.02481564	-0.43339428	0.04703515	0.07191182	0.44590688	
##	Cr	Zn	Ti	Zr	V	B
## Si	0.28442177	0.16886697	0.14720438	-0.05186117	NA	-0.16233589
## Fe	0.11760990	0.18712629	0.04773796	-0.18685554	NA	-0.31560459
## Cu	-0.30875105	0.32661512	0.01882995	0.48864377	NA	-0.13845711
## Mn	0.28998191	0.23479904	0.20480612	0.11018048	NA	-0.15064796
## Mg	0.33209597	-0.21757192	0.53313617	-0.15693241	NA	0.71706925
## Cr	1.00000000	0.01412643	0.05551047	-0.09409050	NA	-0.11616247
## Zn	0.01412643	1.00000000	0.19682255	0.28933882	NA	-0.13367152
## Ti	0.05551047	0.19682255	1.00000000	0.18956819	NA	0.73226155
## Zr	-0.09409050	0.28933882	0.18956819	1.00000000	NA	-0.03919309
## V	NA	NA	NA	NA	1	NA
## B	-0.11616247	-0.13367152	0.73226155	-0.03919309	NA	1.00000000
## Li	-0.06601020	-0.00622270	-0.03687814	-0.02227177	NA	-0.02749633
## Al	-0.14139048	-0.08588036	-0.54585787	-0.22784879	NA	-0.53310298
## U	0.07318558	-0.29584598	0.07620284	-0.06182329	NA	-0.07632594
## I	0.16831816	-0.32416979	-0.19510682	0.13477017	NA	-0.27719980
## Q (kJ/cm)	NA	NA	NA	NA	NA	NA
## Speed (mm/min)	-0.01094396	0.50716711	0.04443688	-0.05772749	NA	-0.07126934
## Strain	-0.07356959	0.06596877	0.08489359	0.07694974	NA	0.03121366
## MCL	0.05095885	-0.16555223	0.29152053	0.01311892	NA	0.49887358
##	Li	Al	U	I	Q (kJ/cm)	
## Si	-0.176045854	-0.30722007	3.114610e-01	0.13730297	NA	
## Fe	-0.203471785	0.21015028	1.437616e-01	0.16278287	NA	
## Cu	0.101388738	-0.43847668	5.113820e-02	-0.05140019	NA	
## Mn	-0.085606846	-0.56725161	4.899444e-01	-0.20438381	NA	
## Mg	-0.110097619	-0.64583767	6.454867e-02	-0.25639121	NA	
## Cr	-0.066010205	-0.14139048	7.318558e-02	0.16831816	NA	
## Zn	-0.006222700	-0.08588036	-2.958460e-01	-0.32416979	NA	
## Ti	-0.036878139	-0.54585787	7.620284e-02	-0.19510682	NA	
## Zr	-0.022271770	-0.22784879	-6.182329e-02	0.13477017	NA	
## V	NA	NA	NA	NA	NA	
## B	-0.027496334	-0.53310298	-7.632594e-02	-0.27719980	NA	

```

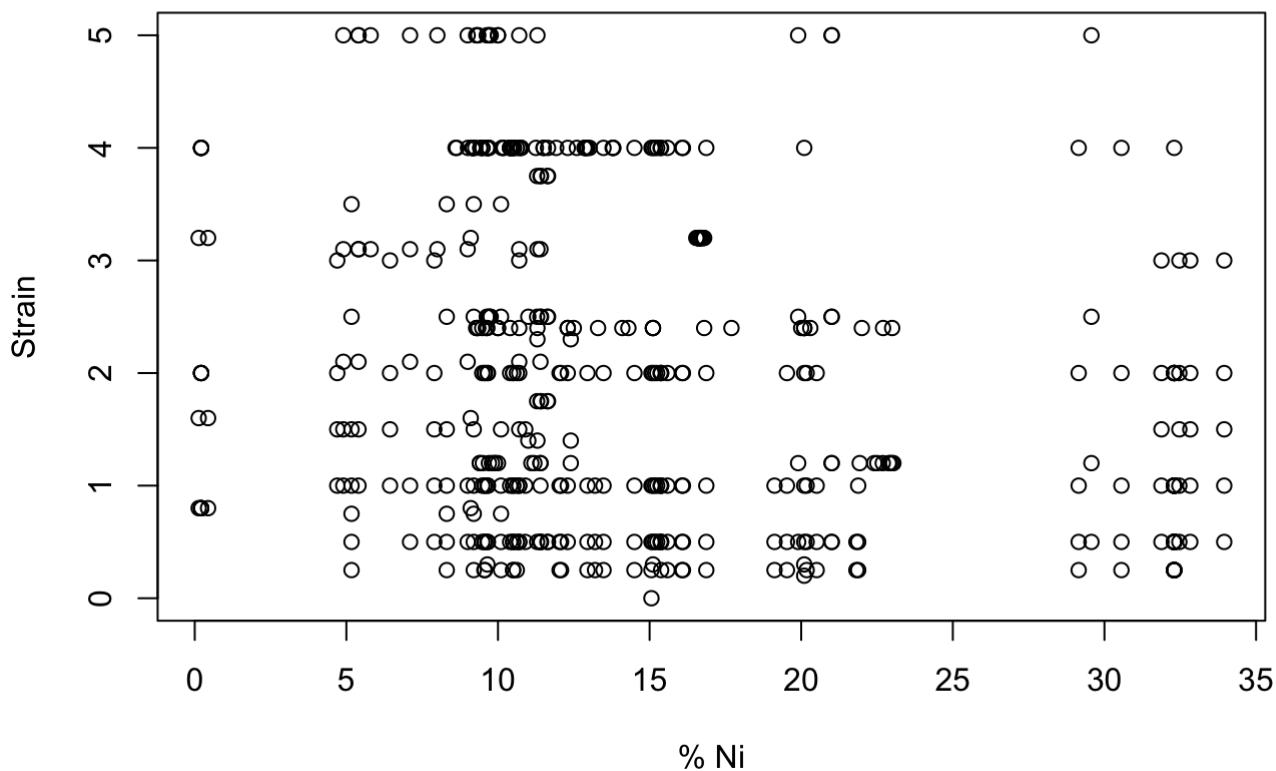
## Li           1.000000000 -0.05764260 -1.434639e-01 -0.44560128     NA
## Al          -0.057642603  1.000000000 -1.436014e-01  0.33418752     NA
## U           -0.143463859 -0.14360139  1.000000e+00 -0.11904365     NA
## I           -0.445601278  0.33418752 -1.190437e-01  1.000000000    NA
## Q (kJ/cm)      NA          NA          NA          NA          1
## Speed (mm/min) 0.500221599 -0.20118505 -3.739558e-03 -0.86067536     NA
## Strain       -0.007889434 -0.02267967 -9.099931e-05 -0.02044928     NA
## MCL          -0.082222575 -0.43379658 -8.733686e-02  0.01904222     NA
##               Speed (mm/min)   Strain        MCL
## Si            0.024860725  3.654705e-02 -0.02481564
## Fe            0.077860633  3.700105e-02 -0.43339428
## Cu            0.136534070  7.475819e-02  0.04703515
## Mn            0.298289490  1.044924e-02  0.07191182
## Mg            0.009804610 -3.762740e-02  0.44590688
## Cr            -0.010943957 -7.356959e-02  0.05095885
## Zn            0.507167108  6.596877e-02 -0.16555223
## Ti            0.044436878  8.489359e-02  0.29152053
## Zr            -0.057727490  7.694974e-02  0.01311892
## V             NA          NA          NA
## B             -0.071269339  3.121366e-02  0.49887358
## Li            0.500221599 -7.889434e-03 -0.08222257
## Al            -0.201185049 -2.267967e-02 -0.43379658
## U             -0.003739558 -9.099931e-05 -0.08733686
## I             -0.860675362 -2.044928e-02  0.01904222
## Q (kJ/cm)      NA          NA          NA
## Speed (mm/min) 1.000000000  2.002035e-02 -0.22277685
## Strain         0.020020351  1.000000e+00  0.21957716
## MCL           -0.222776847  2.195772e-01  1.000000000

```

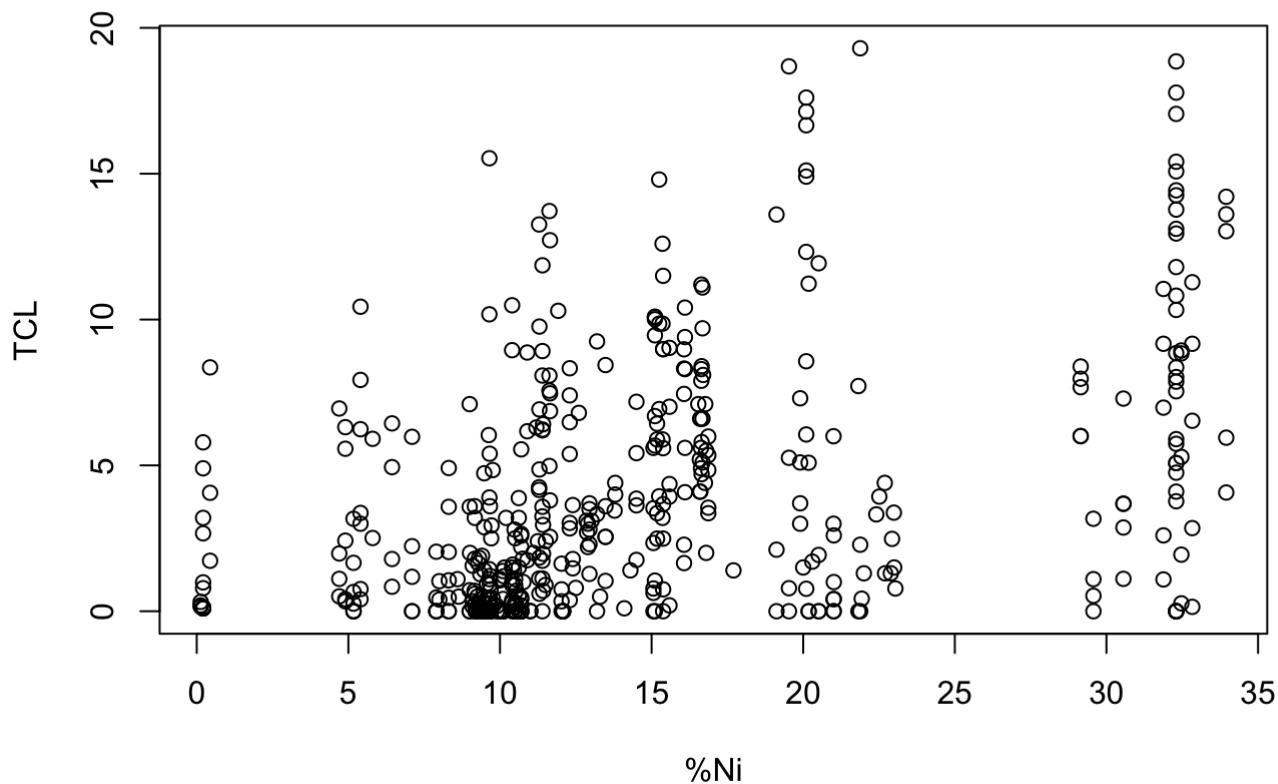
```
plot(al$Mg, al$MCL, xlab = "% Mg", ylab = "MCL")
```



```
plot(stl$Ni, stl$Strain , xlab = "% Ni", ylab = "Strain")
```



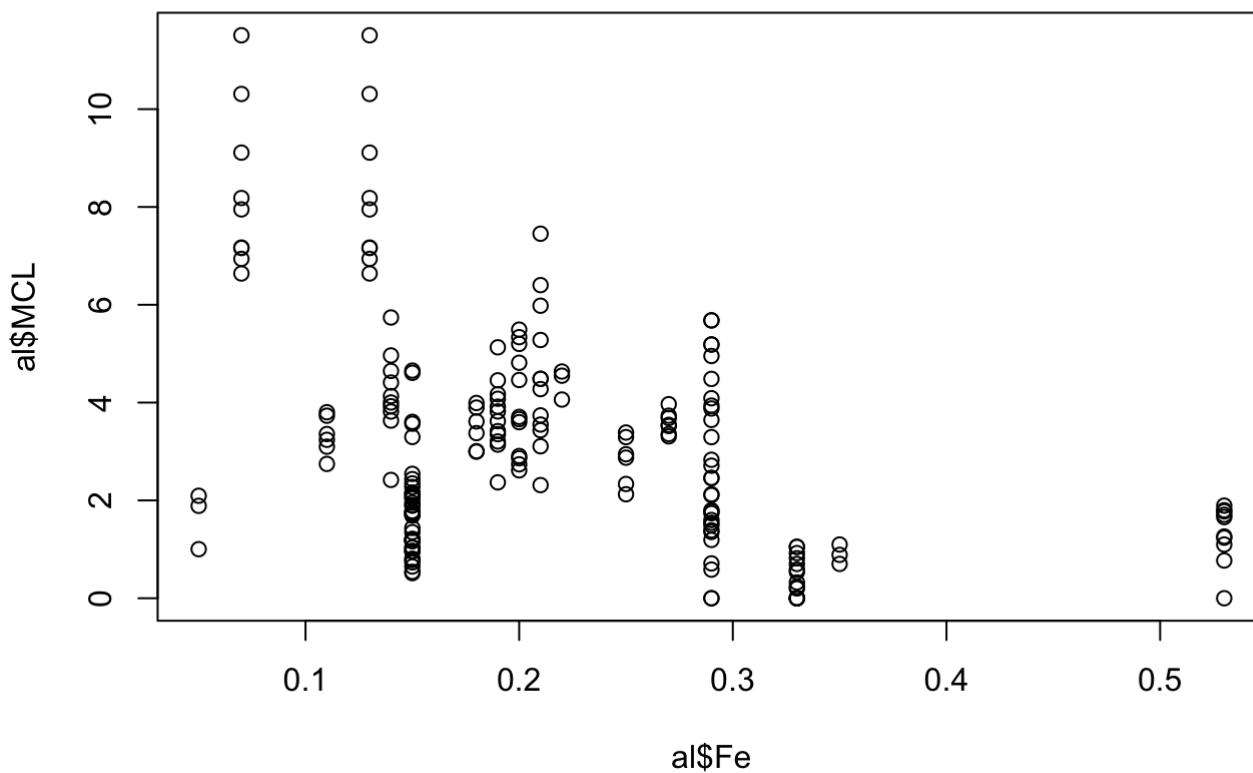
```
plot(stl$Ni, stl$TCL, xlab = "%Ni", ylab = "TCL")
```



```
xlab("%Ni")
```

```
## $x
## [1] "%Ni"
##
## attr(,"class")
## [1] "labels"
```

```
plot(al$Fe, al$MCL)
```



```
#Multiple Linear Regression Observing Strain for Steel Dataset
LMstl = lm(Strain ~ C + Si + Mn + P + S + Cr + Ni + Mo + N + Nb + Co + Cu + Al + Ti + V
+ B, data = stl)
summary(LMstl)
```

```

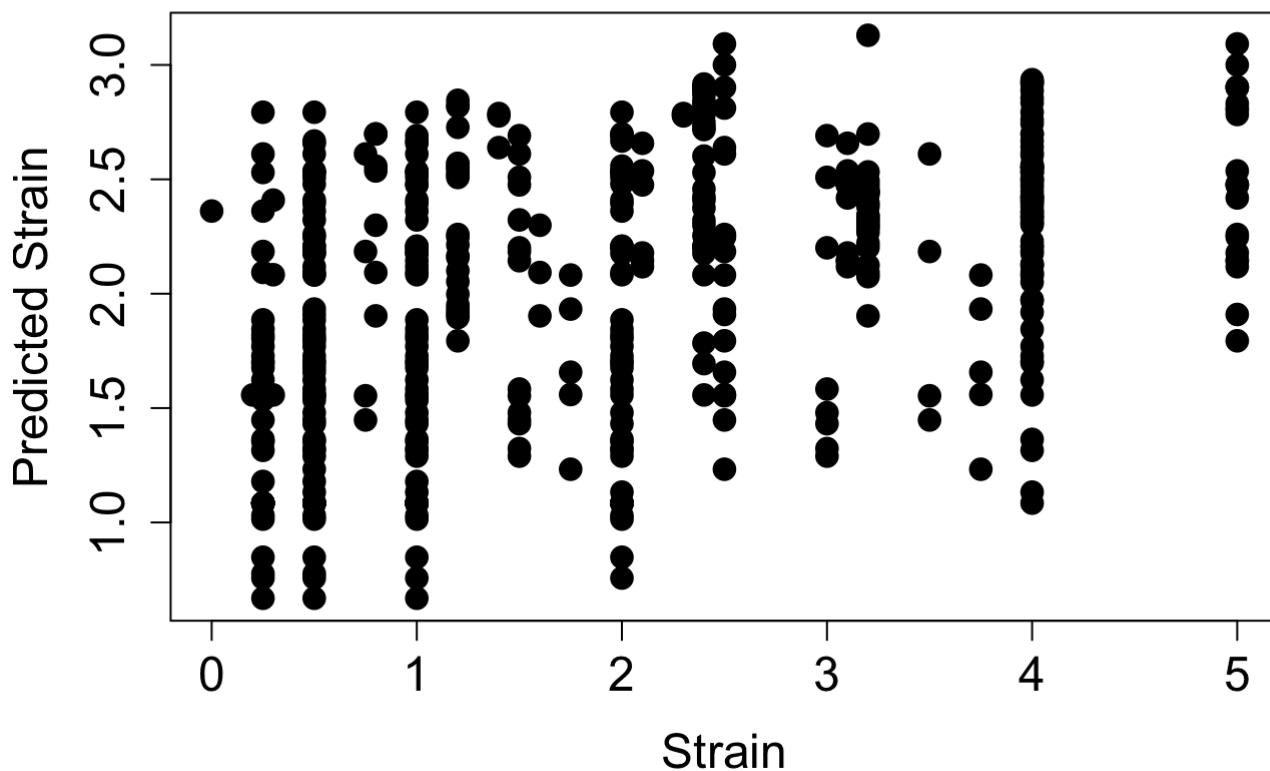
## 
## Call:
## lm(formula = Strain ~ C + Si + Mn + P + S + Cr + Ni + Mo + N +
##     Nb + Co + Cu + Al + Ti + V + B, data = stl)
##
## Residuals:
##    Min      1Q  Median      3Q     Max 
## -2.5433 -1.0243 -0.2570  0.9557  3.2059 
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept)  3.87275   0.60176   6.436 3.04e-10 ***
## C          -1.19395   0.91073  -1.311   0.1905    
## Si         -0.32091   0.15590  -2.058   0.0401 *  
## Mn          0.03211   0.08126   0.395   0.6929    
## P           -7.40409   7.51020  -0.986   0.3247    
## S          -31.55089  10.96724  -2.877   0.0042 ** 
## Cr          -0.04012   0.02877  -1.394   0.1638    
## Ni          -0.02603   0.01140  -2.283   0.0229 *  
## Mo          -0.05497   0.06677  -0.823   0.4108    
## N           1.63127   1.43070   1.140   0.2548    
## Nb          0.29512   0.39216   0.753   0.4521    
## Co          1.92190   1.47413   1.304   0.1930    
## Cu          -0.17698   0.12385  -1.429   0.1537    
## Al          -1.15149   0.49299  -2.336   0.0199 *  
## Ti          -0.20741   0.42841  -0.484   0.6285    
## V            -1.23342   0.50128  -2.461   0.0142 *  
## B           80.31391  49.08260   1.636   0.1024    
## ---      
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.305 on 470 degrees of freedom
## Multiple R-squared:  0.1675, Adjusted R-squared:  0.1391 
## F-statistic:  5.91 on 16 and 470 DF,  p-value: 7.467e-12

```

```

preLMmstl=predict(LMstl)
plot(stl$Strain, predict(LMstl), xlab = "Strain",
      ylab = "Predicted Strain", cex.axis = 1.5,
      cex.lab = 1.5, cex = 1.5, pch = 19)

```



```
#MLR Analysis for Strain using the Aluminum Dataset
```

```
LMal = lm(Strain ~ Si + Mn + Fe + Mg + Cr + Mn + Zn + Cu + Al + Ti + V + Zr + B + Li, data = al)
summary(LMal)
```

```

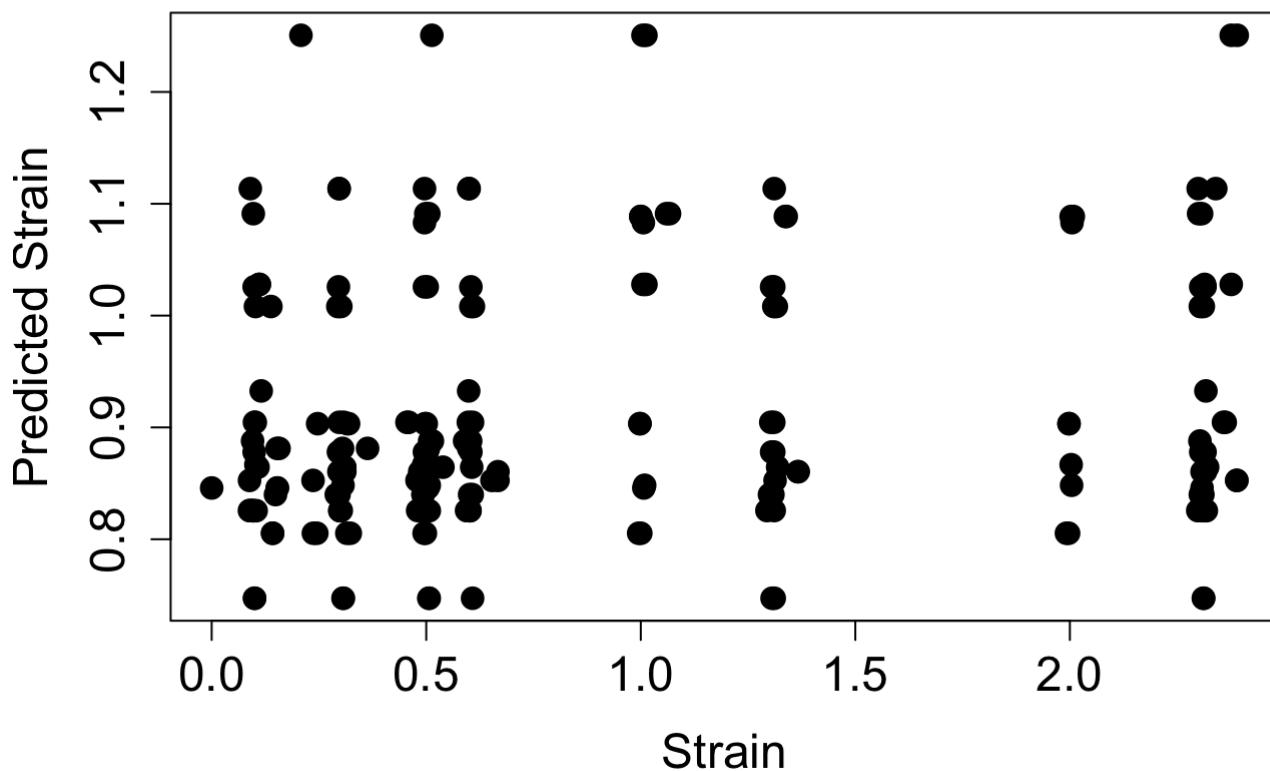
## 
## Call:
## lm(formula = Strain ~ Si + Mn + Fe + Mg + Cr + Mn + Zn + Cu +
##     Al + Ti + V + Zr + B + Li, data = al)
##
## Residuals:
##      Min      1Q Median      3Q      Max
## -1.0427 -0.5723 -0.3104  0.4602  1.5648
##
## Coefficients: (2 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.517734  22.830605 -0.023   0.982
## Si          1.093791  2.072920  0.528   0.598
## Mn         -0.153049  0.511311 -0.299   0.765
## Fe          0.066153  0.767541  0.086   0.931
## Mg          0.004737  0.233416  0.020   0.984
## Cr         -0.715096  1.072397 -0.667   0.506
## Zn          0.791344  3.725710  0.212   0.832
## Cu          0.018793  0.246490  0.076   0.939
## Al          0.013137  0.228959  0.057   0.954
## Ti          4.164781  6.154100  0.677   0.499
## V            NA        NA        NA        NA
## Zr          1.329972  3.366489  0.395   0.693
## B           -32.679515 215.865829 -0.151   0.880
## Li          NA        NA        NA        NA
##
## Residual standard error: 0.7917 on 183 degrees of freedom
## Multiple R-squared:  0.02249,    Adjusted R-squared:  -0.03627
## F-statistic: 0.3827 on 11 and 183 DF,  p-value: 0.9616

```

```

preLMal=predict(LMal)
plot(al$Strain, predict(LMal), xlab = "Strain",
      ylab = "Predicted Strain", cex.axis = 1.5,
      cex.lab = 1.5, cex = 1.5, pch = 19)

```



```
#Using MLR to predict Total Crack Length in the Steel Dataset
```

```
LMstl_C = lm(TCL ~ C + Si + Mn + P + S + Cr + Ni + Mo + N + Nb + Co + Cu + Al + Ti + V +  
B, data = stl)  
summary(LMstl_C)
```

```

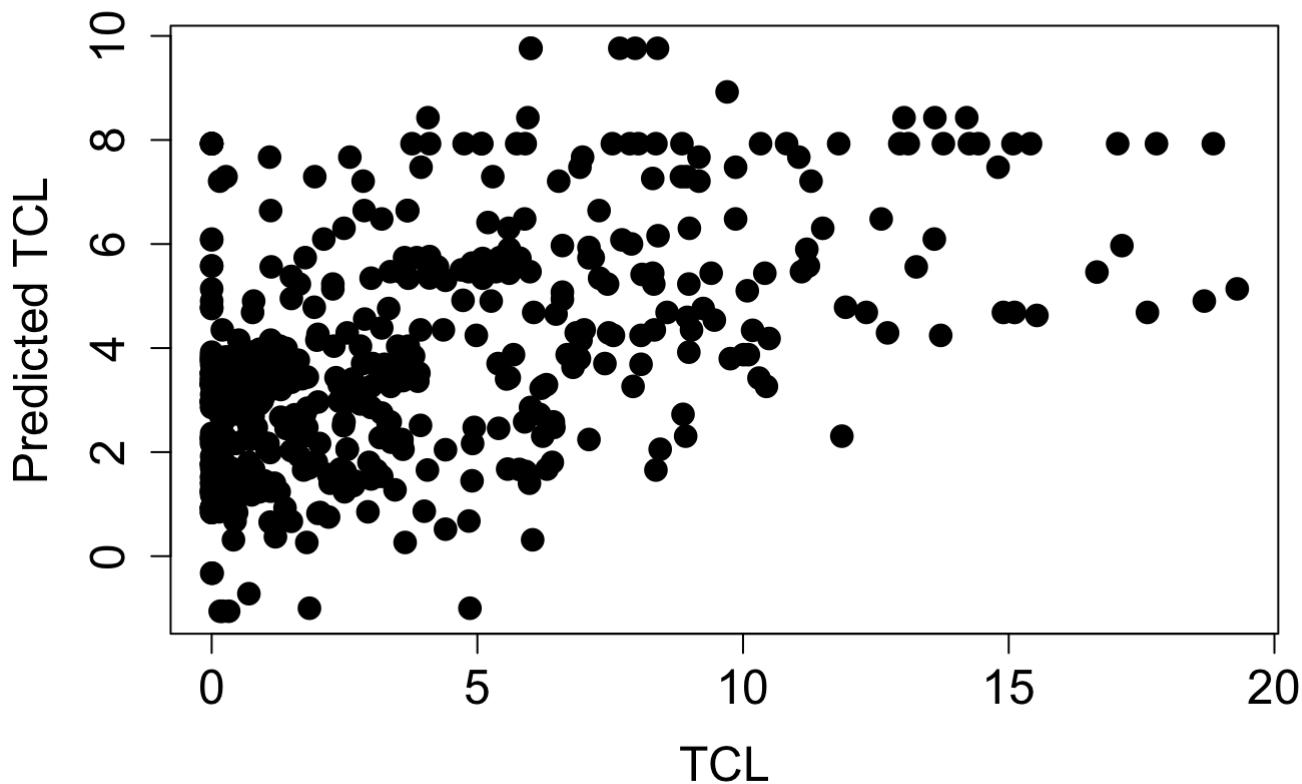
## 
## Call:
## lm(formula = TCL ~ C + Si + Mn + P + S + Cr + Ni + Mo + N + Nb +
##     Co + Cu + Al + Ti + V + B, data = stl)
##
## Residuals:
##    Min      1Q  Median      3Q     Max 
## -7.9291 -2.3798 -0.7128  1.6535 14.1621 
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept)  2.33857   1.66958   1.401  0.161965    
## C          -1.14612   2.52681  -0.454  0.650338    
## Si          1.17301   0.43255   2.712  0.006936 **  
## Mn         -0.84531   0.22544  -3.750  0.000199 *** 
## P           74.07559  20.83690   3.555  0.000416 *** 
## S          -75.77125  30.42840  -2.490  0.013113 *   
## Cr          -0.13295   0.07982  -1.666  0.096449 .  
## Ni          0.23756   0.03163   7.512  2.97e-13 *** 
## Mo          -0.28271   0.18525  -1.526  0.127662    
## N           9.23596   3.96945   2.327  0.020402 *  
## Nb          -1.00382   1.08804  -0.923  0.356692    
## Co         -11.31785  4.08994  -2.767  0.005876 **  
## Cu          0.57644   0.34362   1.678  0.094102 .  
## Al          -1.09992   1.36779  -0.804  0.421711    
## Ti          0.51171   1.18862   0.431  0.667023    
## V            1.27378   1.39080   0.916  0.360210    
## B           350.08885 136.17870   2.571  0.010453 *  
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.621 on 470 degrees of freedom
## Multiple R-squared:  0.2772, Adjusted R-squared:  0.2526 
## F-statistic: 11.26 on 16 and 470 DF,  p-value: < 2.2e-16

```

```

preLMstl_C=predict(LMstl_C)
plot(stl$TCL, predict(LMstl_C), xlab = "TCL",
      ylab = "Predicted TCL", cex.axis = 1.5,
      cex.lab = 1.5, cex = 1.5, pch = 19)

```



```
#Using MLR to predict Maximum crack Length in Aluminum Dataset
LMal_C = lm(MCL ~ Si + Mn + Fe + Mg + Cr + Mn + Zn + Cu + Al + Ti + V + Zr + B + Li, da
ta = al)
summary(LMal_C)
```

```

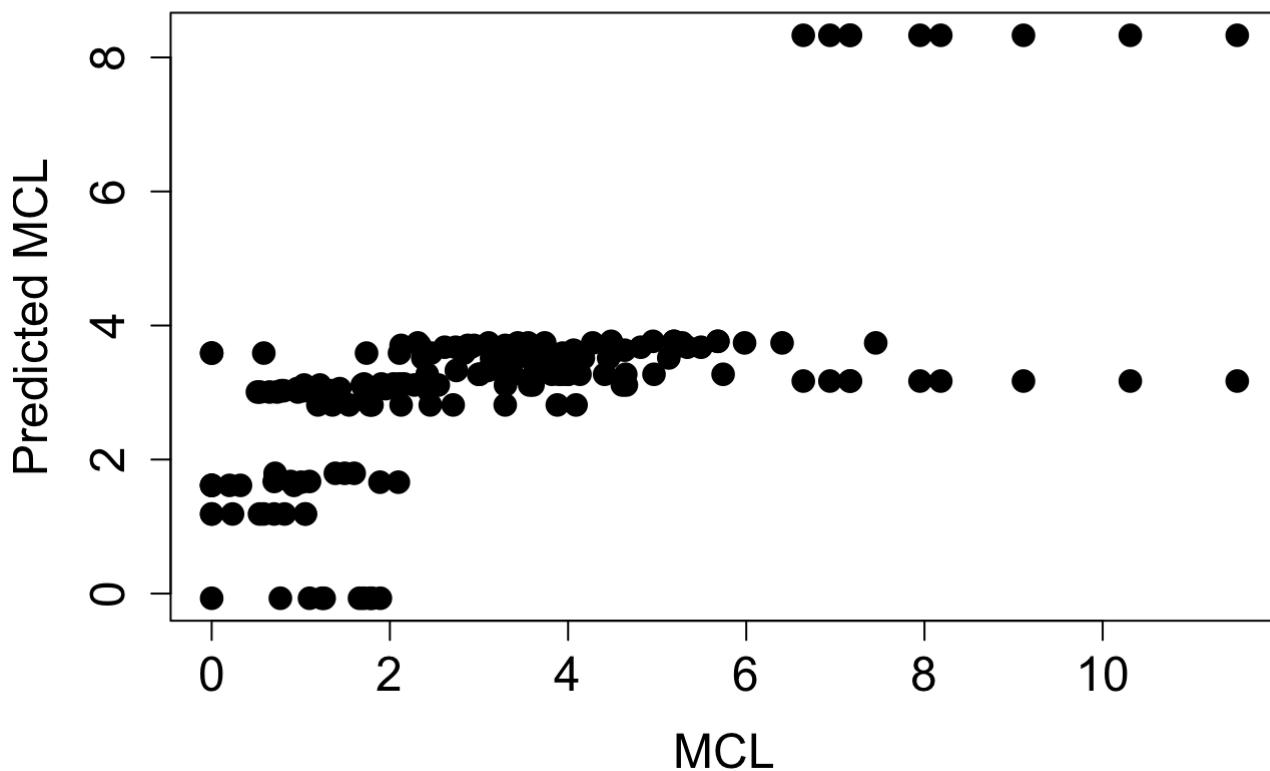
## 
## Call:
## lm(formula = MCL ~ Si + Mn + Fe + Mg + Cr + Mn + Zn + Cu + Al +
##     Ti + V + Zr + B + Li, data = al)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5886 -1.1537 -0.2303  0.7453  8.3386
##
## Coefficients: (2 not defined because of singularities)
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -130.3028    51.3230 -2.539  0.01195 *
## Si           2.8753     4.6599  0.617  0.53799
## Mn           1.4725     1.1494  1.281  0.20178
## Fe          -7.2942     1.7254 -4.227 3.73e-05 ***
## Mg           1.3854     0.5247  2.640  0.00900 **
## Cr           6.4416     2.4107  2.672  0.00822 **
## Zn          -12.4408     8.3754 -1.485  0.13916
## Cu           1.6765     0.5541  3.026  0.00284 **
## Al           1.3449     0.5147  2.613  0.00972 **
## Ti           7.1976    13.8344  0.520  0.60351
## V            NA         NA       NA       NA
## Zr          -11.9012     7.5678 -1.573  0.11754
## B            934.4192   485.2644  1.926  0.05571 .
## Li           NA         NA       NA       NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.78 on 183 degrees of freedom
## Multiple R-squared:  0.4277, Adjusted R-squared:  0.3933
## F-statistic: 12.43 on 11 and 183 DF,  p-value: < 2.2e-16

```

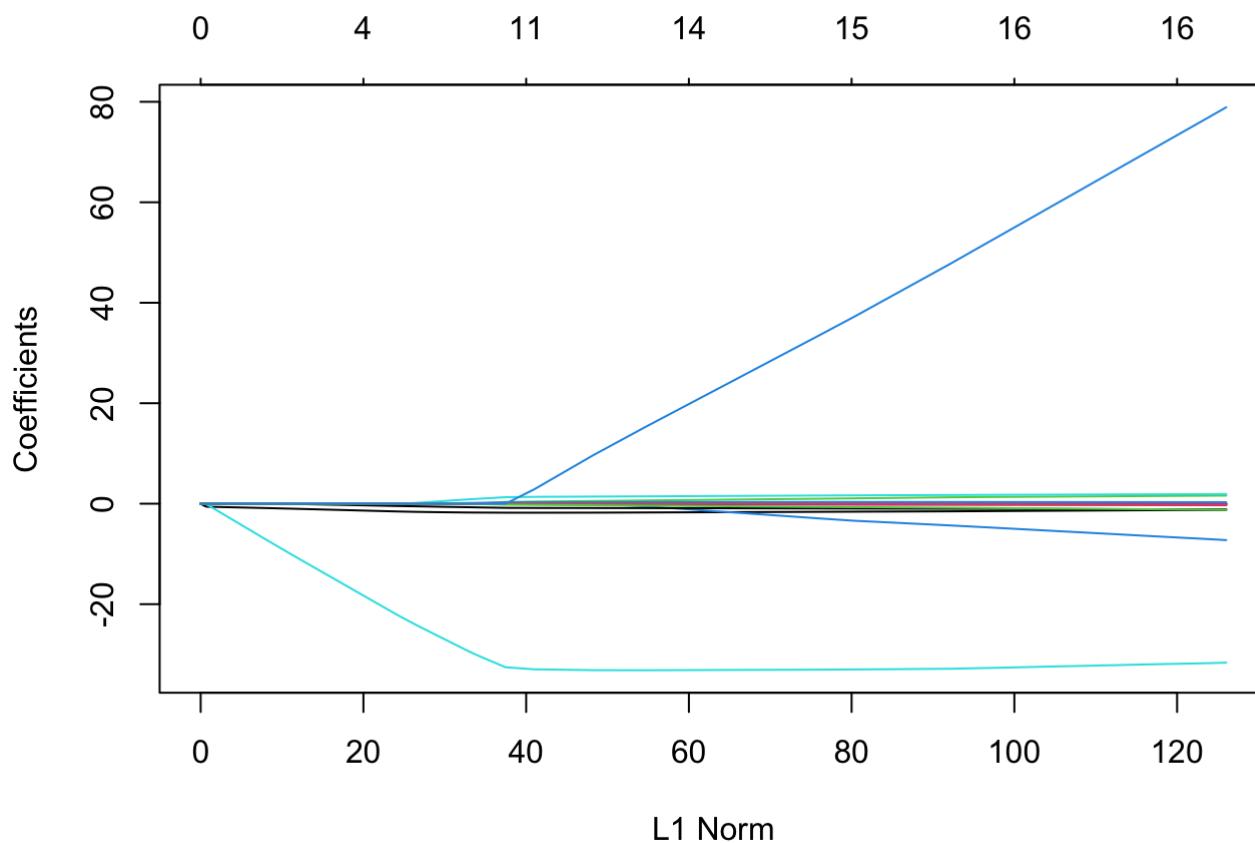
```

preLMal_C=predict(LMal_C)
plot(al$MCL, predict(LMal_C), xlab = "MCL",
      ylab = "Predicted MCL", cex.axis = 1.5,
      cex.lab = 1.5, cex = 1.5, pch = 19)

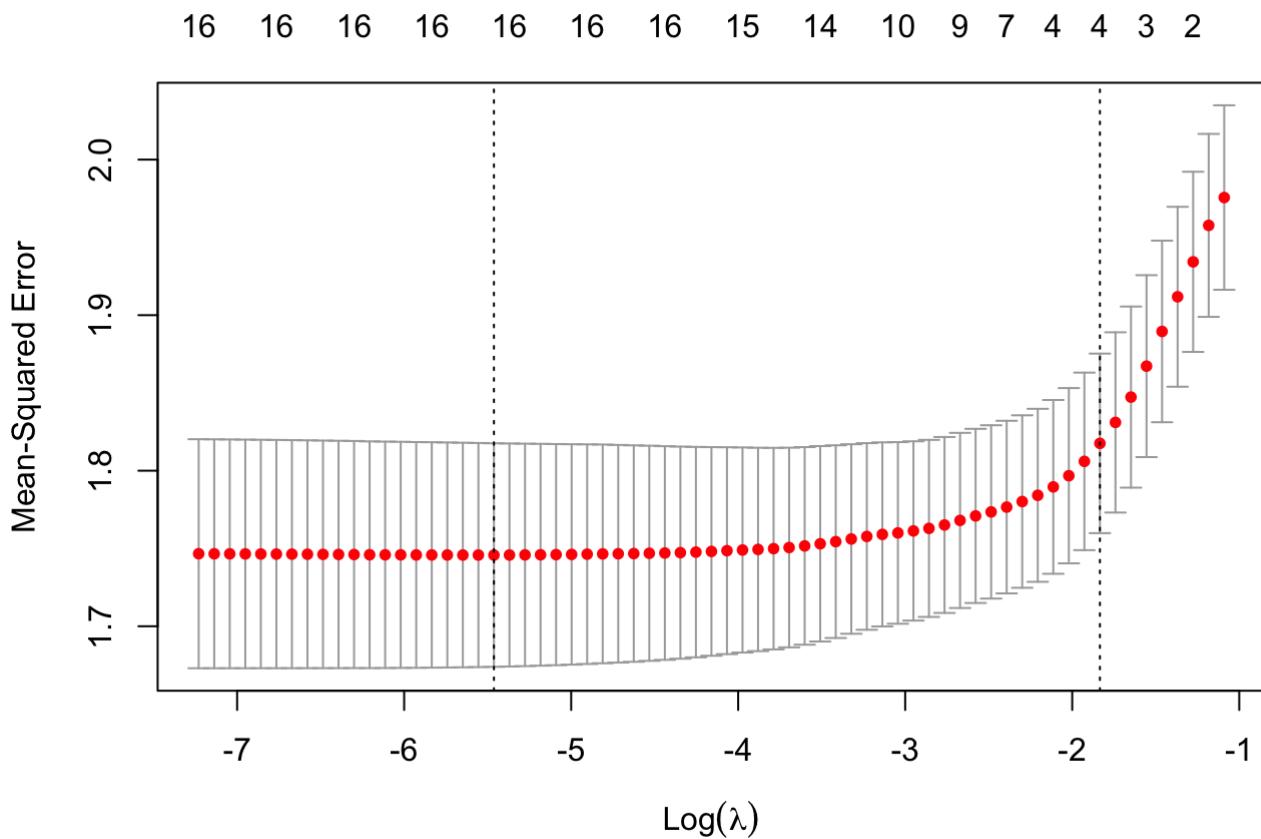
```



```
#Feature Selection and PCA to Predict Strain Values  
#Steel Strain Model  
stEl <- as.matrix(stEl)  
stSTN <- as.matrix(stSTN)  
COMstl <- glmnet(stEl ,stSTN)  
plot(COMstl)
```



```
COMfit_stSTN = cv.glmnet(stEl, stSTN, alpha = 1)
plot(COMfit_stSTN)
```



```
COMfit_stSTN$lambda.min
```

```
## [1] 0.004242854
```

```
COMfit_stSTN$lambda.1se
```

```
## [1] 0.1597407
```

```
library(plotmo)
```

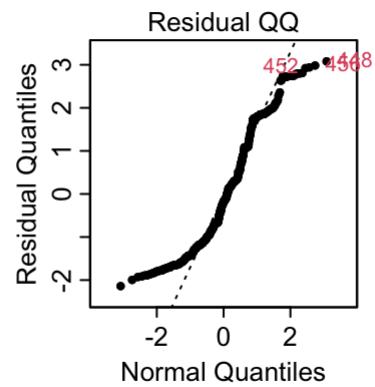
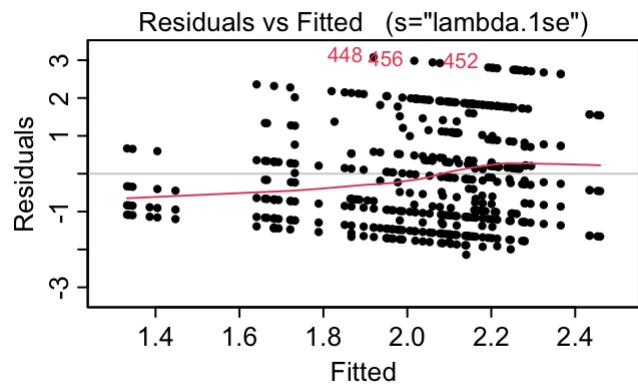
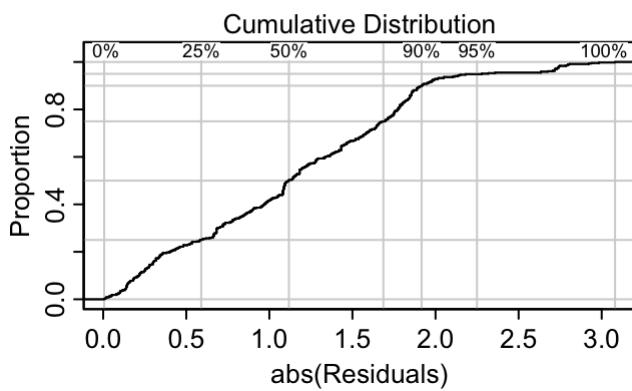
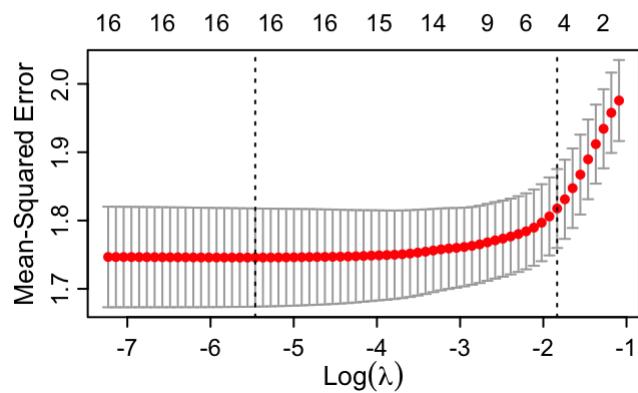
```
## Loading required package: Formula
```

```
## Loading required package: plotrix
```

```
## Loading required package: TeachingDemos
```

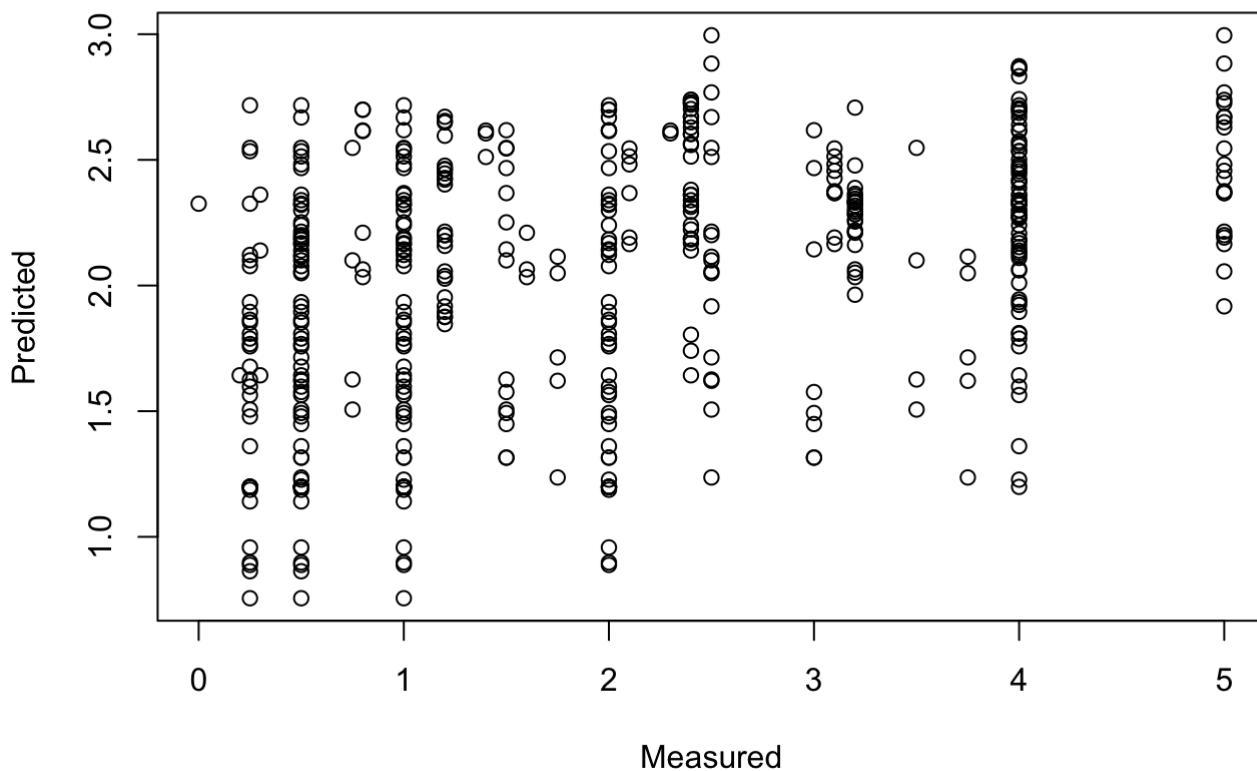
```
plotres(COMfit_stSTN)
```

s="lambda.1se" cv.glmnet(x...

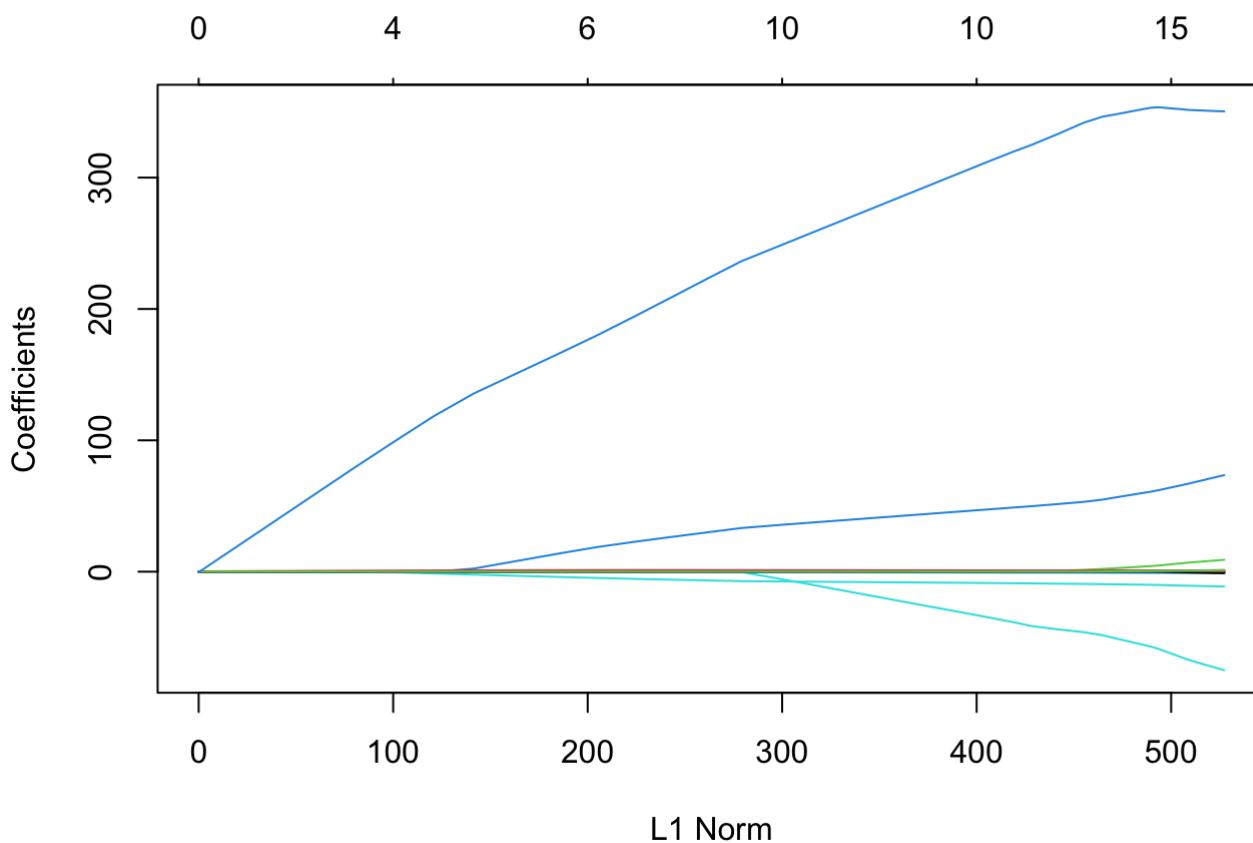


```
pred_stl <- predict(COMstl, stEl, s = .02)
```

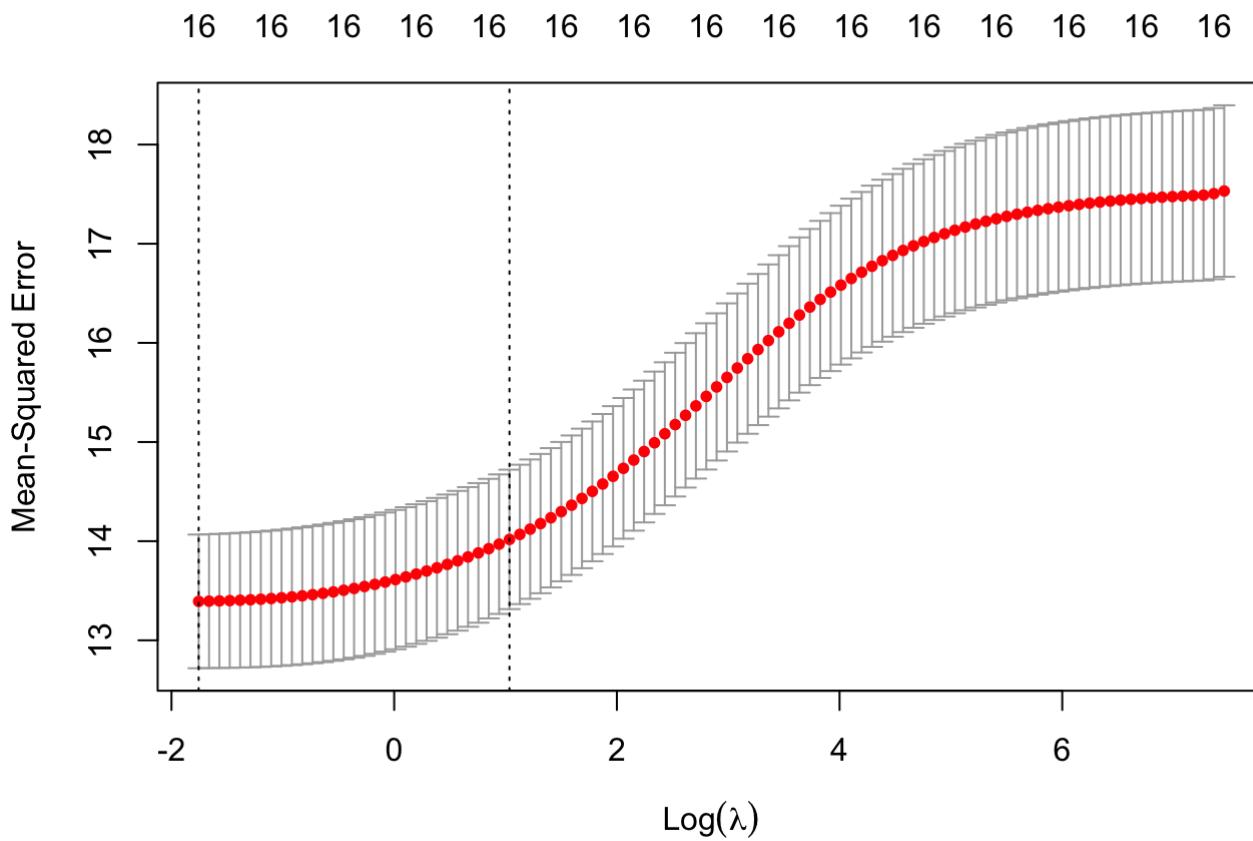
```
stl_t <- pred_stl[,c("st1")]
stl_s <- as.numeric(stSTN)
plot(stl_s,stl_t, xlab="Measured", ylab = "Predicted")
```



```
#Steel TCL Model
stEl <- as.matrix(stEl)
stTCL <- as.matrix(stTCL)
COMstl2 <- glmnet(stEl ,stTCL)
plot(COMstl2)
```



```
COMfit_stTCL = cv.glmnet(stEl, stTCL, alpha = 0)
plot(COMfit_stTCL)
```



```
COMfit_stTCL$lambda.min
```

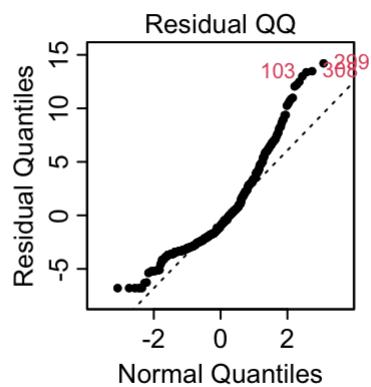
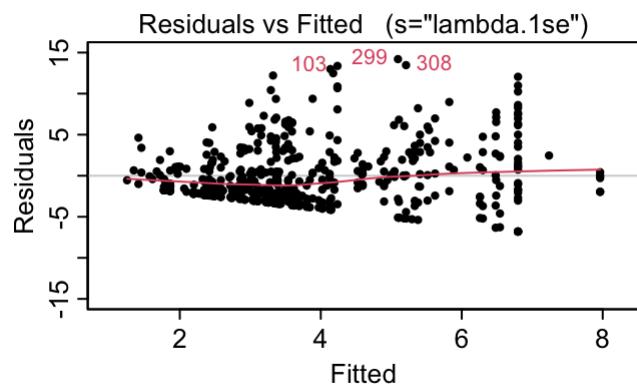
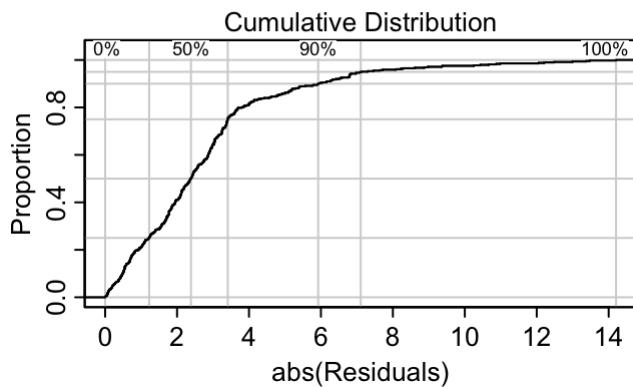
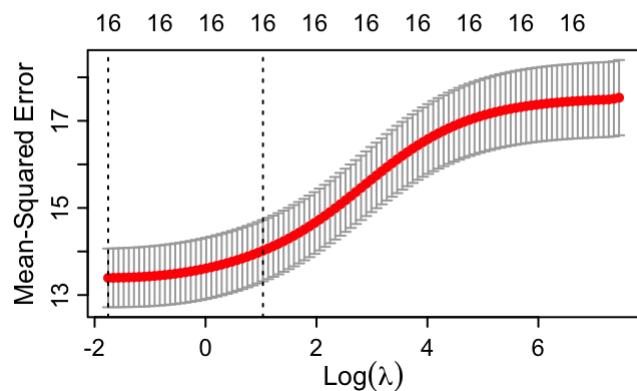
```
## [1] 0.1727779
```

```
COMfit_stTCL$lambda.1se
```

```
## [1] 2.815849
```

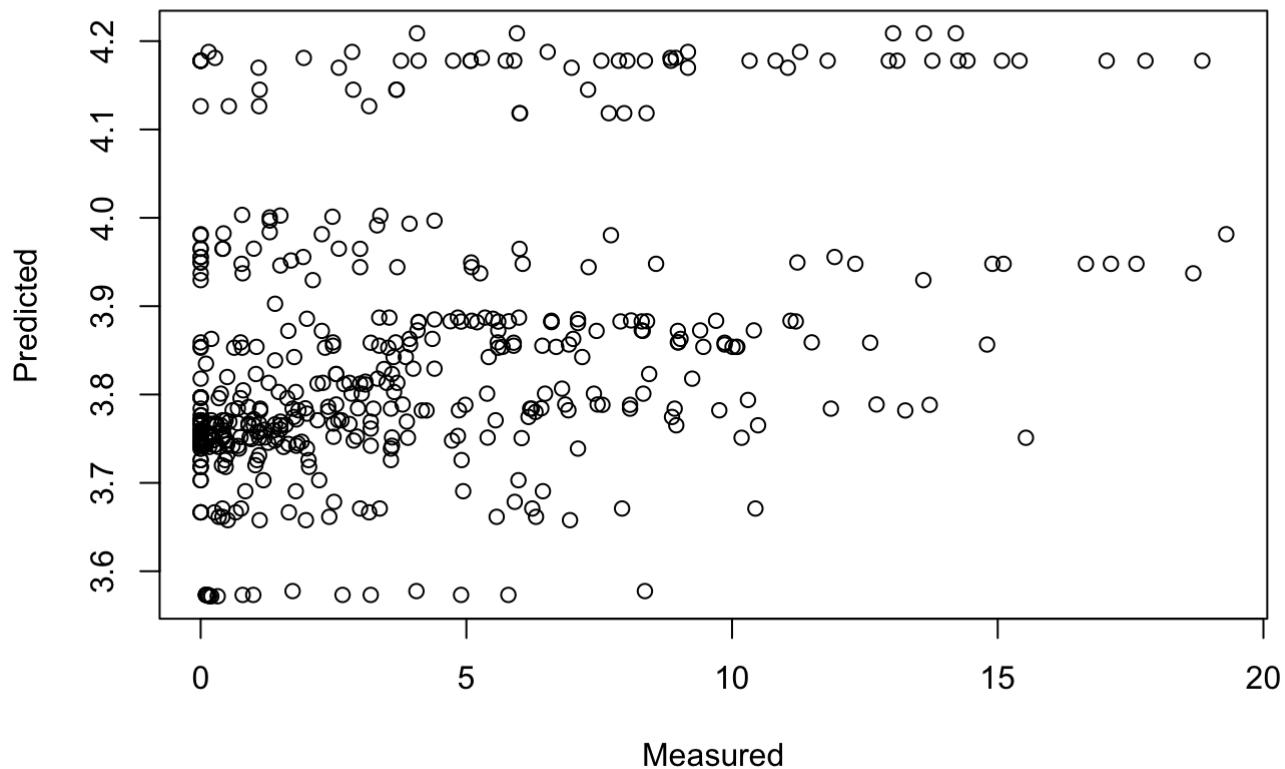
```
plotres(COMfit_stTCL)
```

s="lambda.1se" cv.glmnet(x...

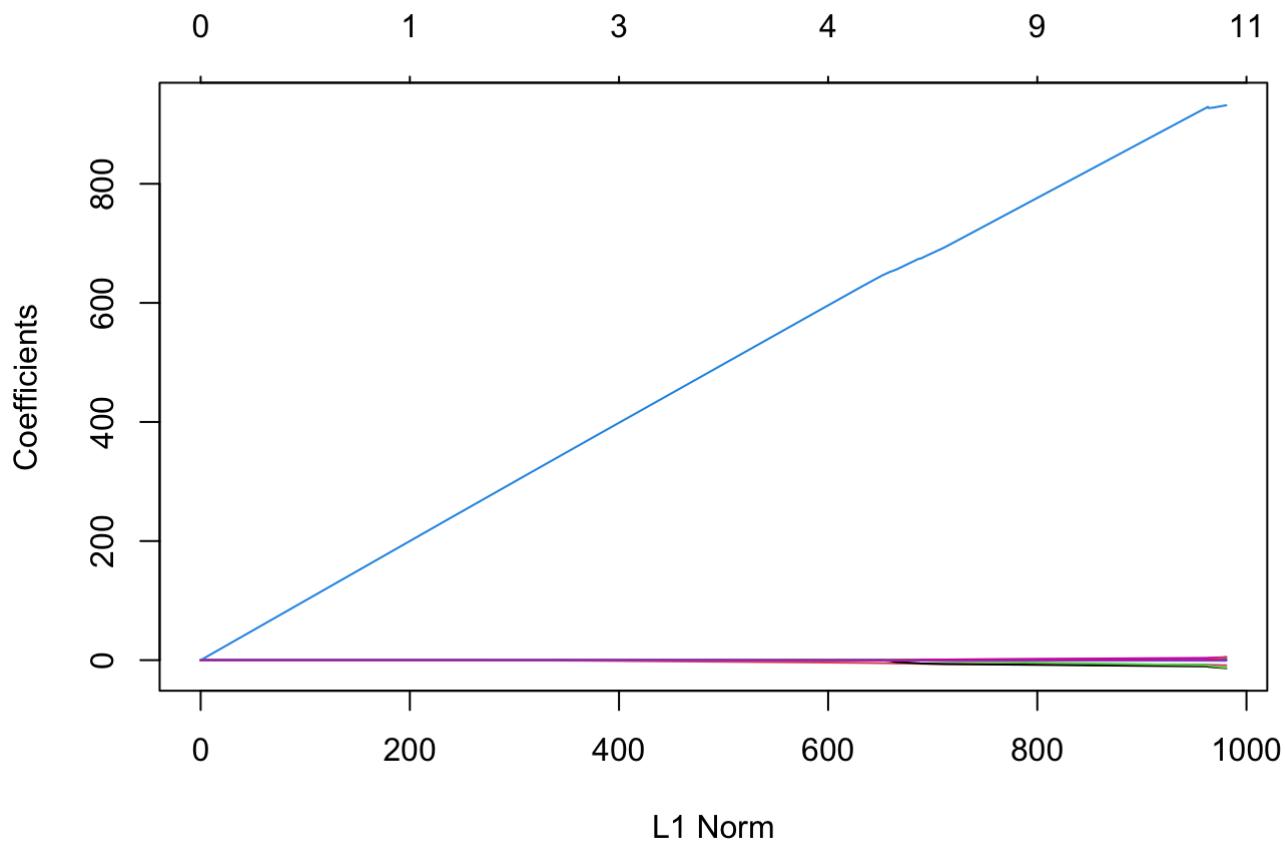


```
pred_stl2 <- predict(COMstl2, stEl)
```

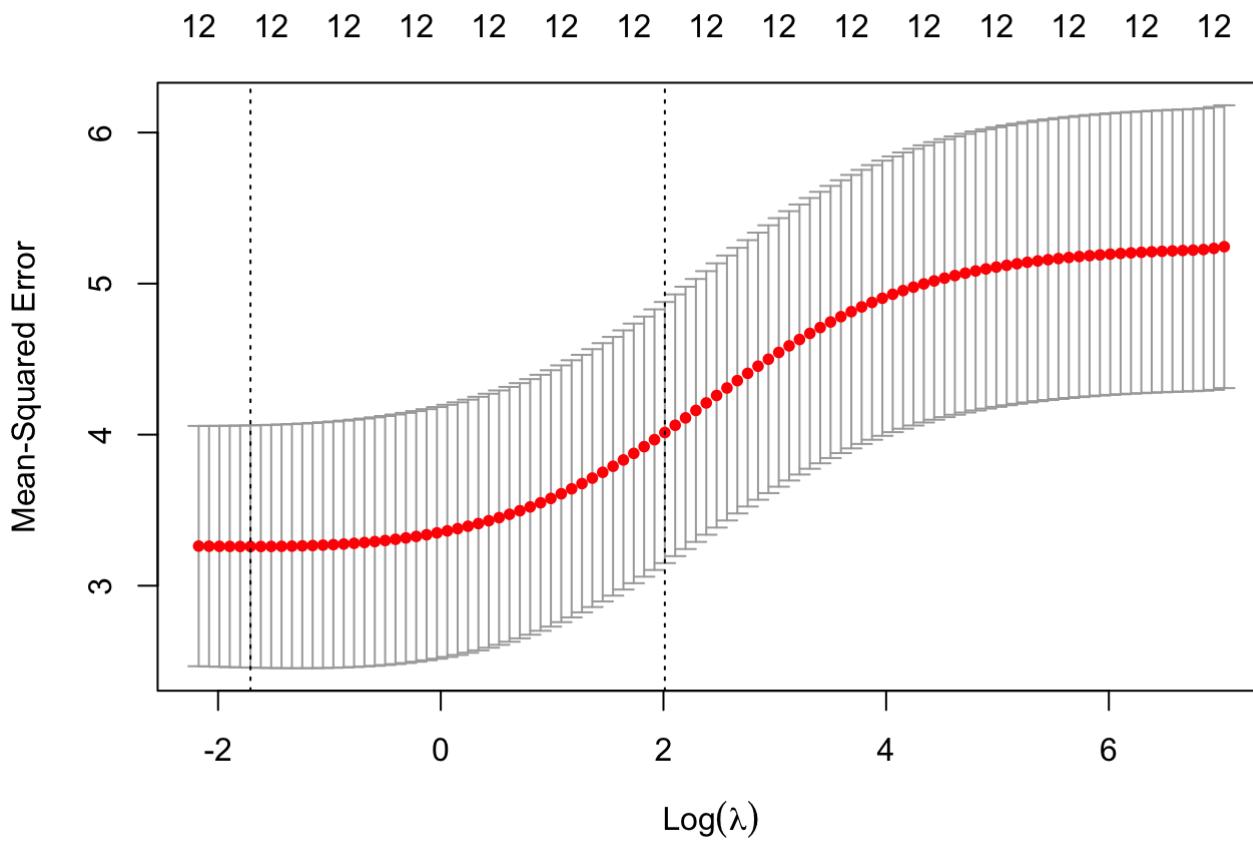
```
stl_t2 <- pred_stl2[,c("s1")]
stl_s2 <- as.numeric(stTCL)
plot(stl_s2,stl_t2, xlab="Measured", ylab = "Predicted")
```



```
aleL <- as.matrix(aleL)
alMCL <- as.matrix(alMCL)
COMal <- glmnet(aleL ,alMCL)
plot(COMal)
```



```
COMfit_al = cv.glmnet(alEl, almCL, alpha = 0)
plot(COMfit_al)
```



```
COMfit_al$lambda.min
```

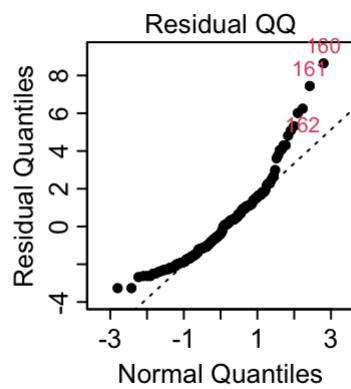
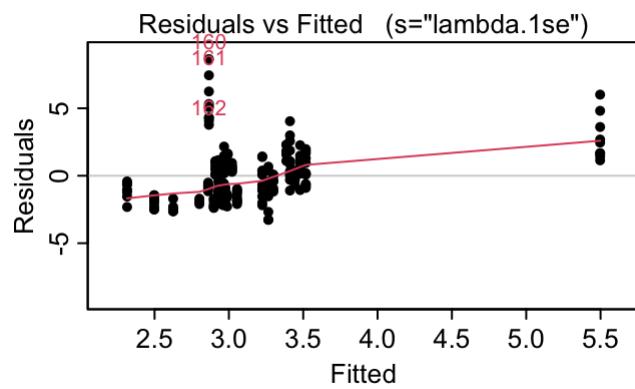
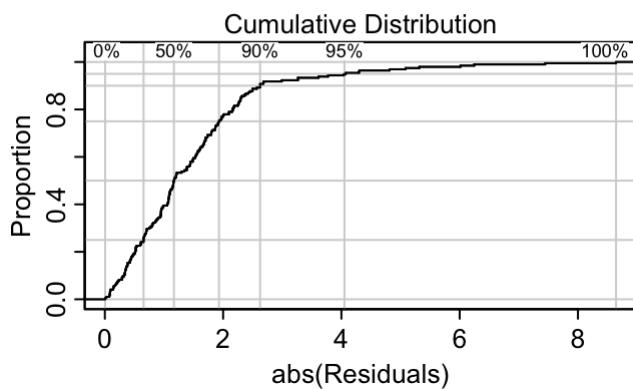
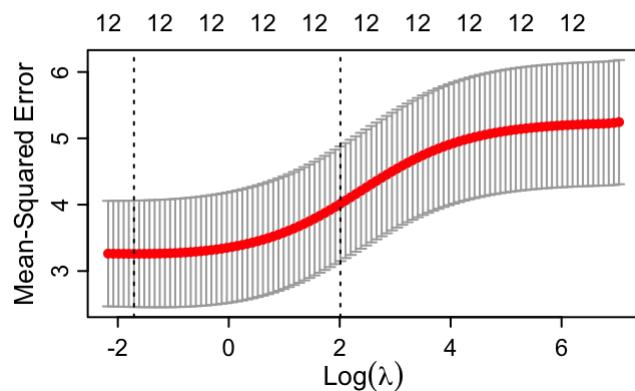
```
## [1] 0.1810444
```

```
COMfit_al$lambda.1se
```

```
## [1] 7.480779
```

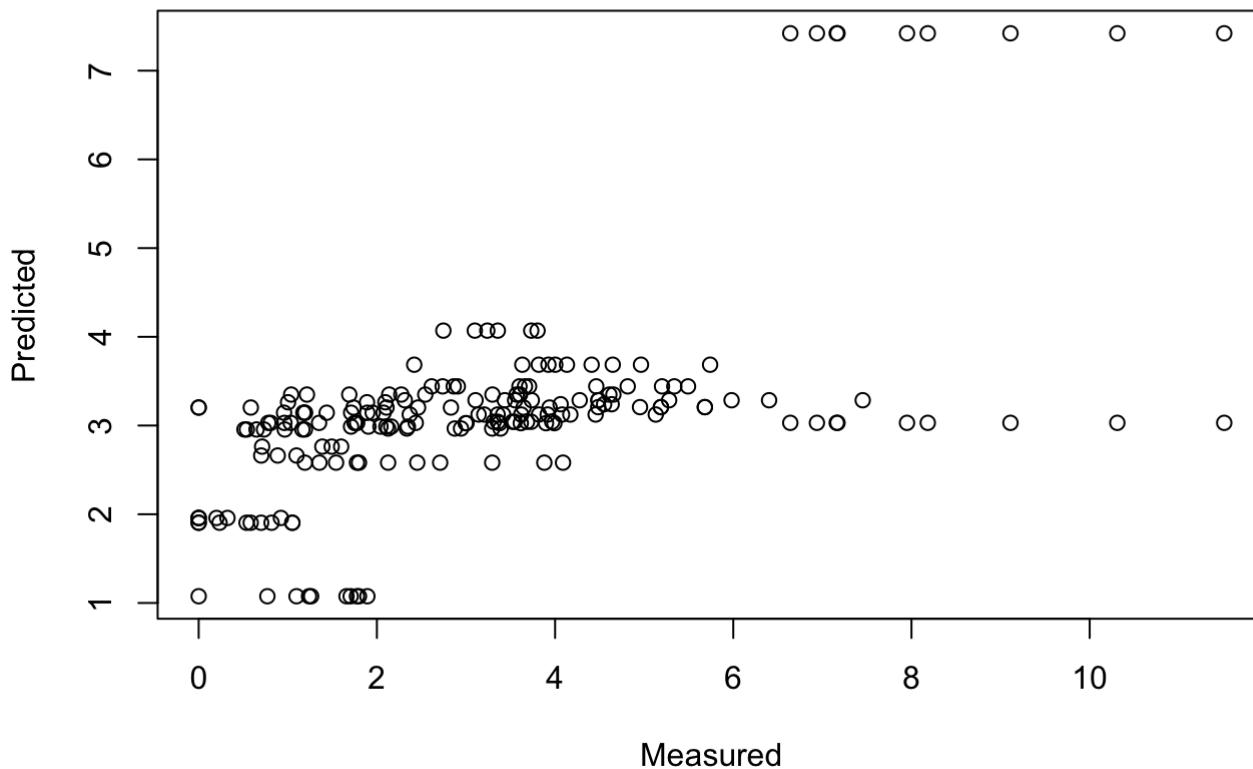
```
plotres(COMfit_al)
```

s="lambda.1se" cv.glmnet(x...



```
pred_al <- predict(COMal, alEl, s = .2)
```

```
al_t2 <- pred_al[,c("s1")]
al_s2 <- as.numeric(alMCL)
plot(al_s2,al_t2, xlab="Measured", ylab = "Predicted")
```



```
#defining dataframes that can be used for PCA
alrf_MCL <- al[c(1:13,19)]
alrf_STN <- al[c(1:13,18)]
stlrf_TCL <- stl[c(1:16,22)]
stlrf_STN <- stl[c(1:16,21)]
```

```
#Using PCA to identify sources of variance and hopefully insights into model simplifications
Si <- as.numeric(alrf_MCL$Si)
Fe <- as.numeric(alrf_MCL$Fe)
Cu <- as.numeric(alrf_MCL$Cu)
Mn <- as.numeric(alrf_MCL$Mn)
Mg <- as.numeric(alrf_MCL$Mg)
Cr <- as.numeric(alrf_MCL$Cr)
Zn <- as.numeric(alrf_MCL$Zn)
Ti <- as.numeric(alrf_MCL$Ti)
Zr <- as.numeric(alrf_MCL$Zr)
V <- as.numeric(alrf_MCL$V)
B <- as.numeric(alrf_MCL$B)
Li <- as.numeric(alrf_MCL$Li)
Al <- as.numeric(alrf_MCL$Al)
MCL <- as.numeric(alrf_MCL$MCL)
alSTN <- as.numeric(alrf_STN$Strain)
```

```
#PCA for MCL in Al
pralT <- cbind(Si,Fe,Cu,Mn,Mg,Cr,Zn,Ti,Zr,V,B,Li,Al,MCL)
```

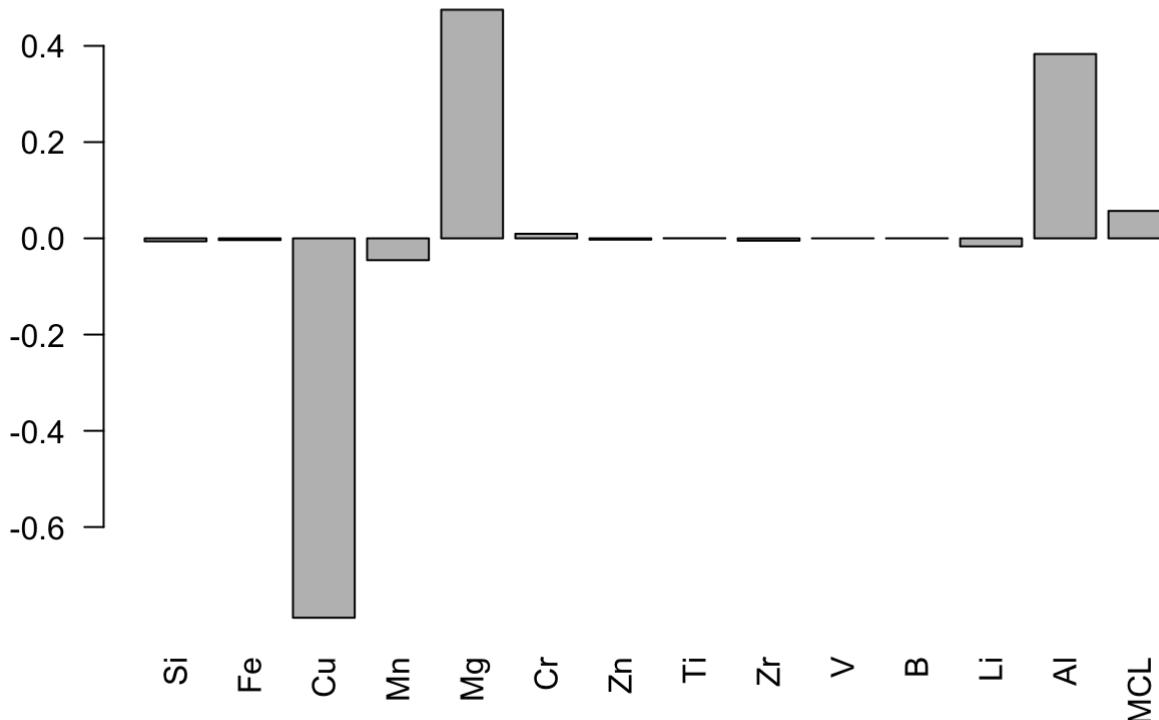
```
pralPCA = prcomp(pralT, scale=F)
```

```
summary(pralPCA)
```

```
## Importance of components:
##                               PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation     3.4231  2.3611  1.8349  0.30080  0.25441  0.09686  0.06397
## Proportion of Variance 0.5625  0.2676  0.1616  0.00434  0.00311  0.00045  0.00020
## Cumulative Proportion  0.5625  0.8302  0.9918  0.99615  0.99926  0.99971  0.99991
##                               PC8      PC9      PC10     PC11     PC12     PC13
## Standard deviation     0.02839 0.02624 0.01585 0.01254 0.0002719 1.85e-15
## Proportion of Variance 0.00004 0.00003 0.00001 0.00001 0.0000000 0.00e+00
## Cumulative Proportion  0.99995 0.99998 0.99999 1.00000 1.0000000 1.00e+00
##                               PC14
## Standard deviation     5.653e-19
## Proportion of Variance 0.000e+00
## Cumulative Proportion 1.000e+00
```

```
barplot(pralPCA$rotation[,2], main="PC 1 Loadings Plot", las=2)
```

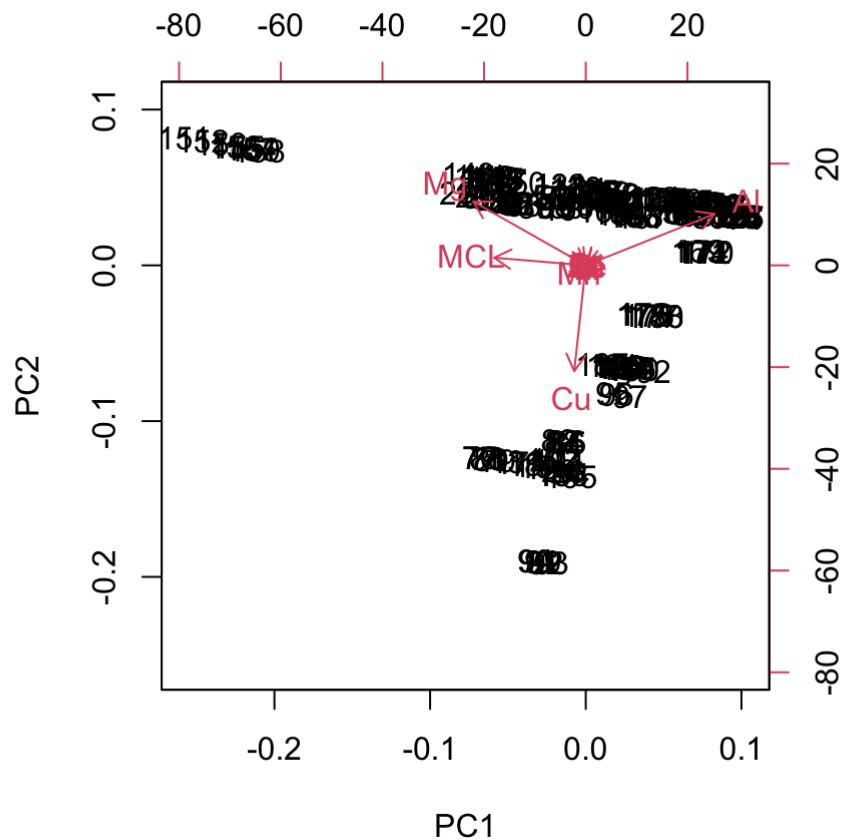
PC 1 Loadings Plot



```
biplot(pralPCA)
```

```
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

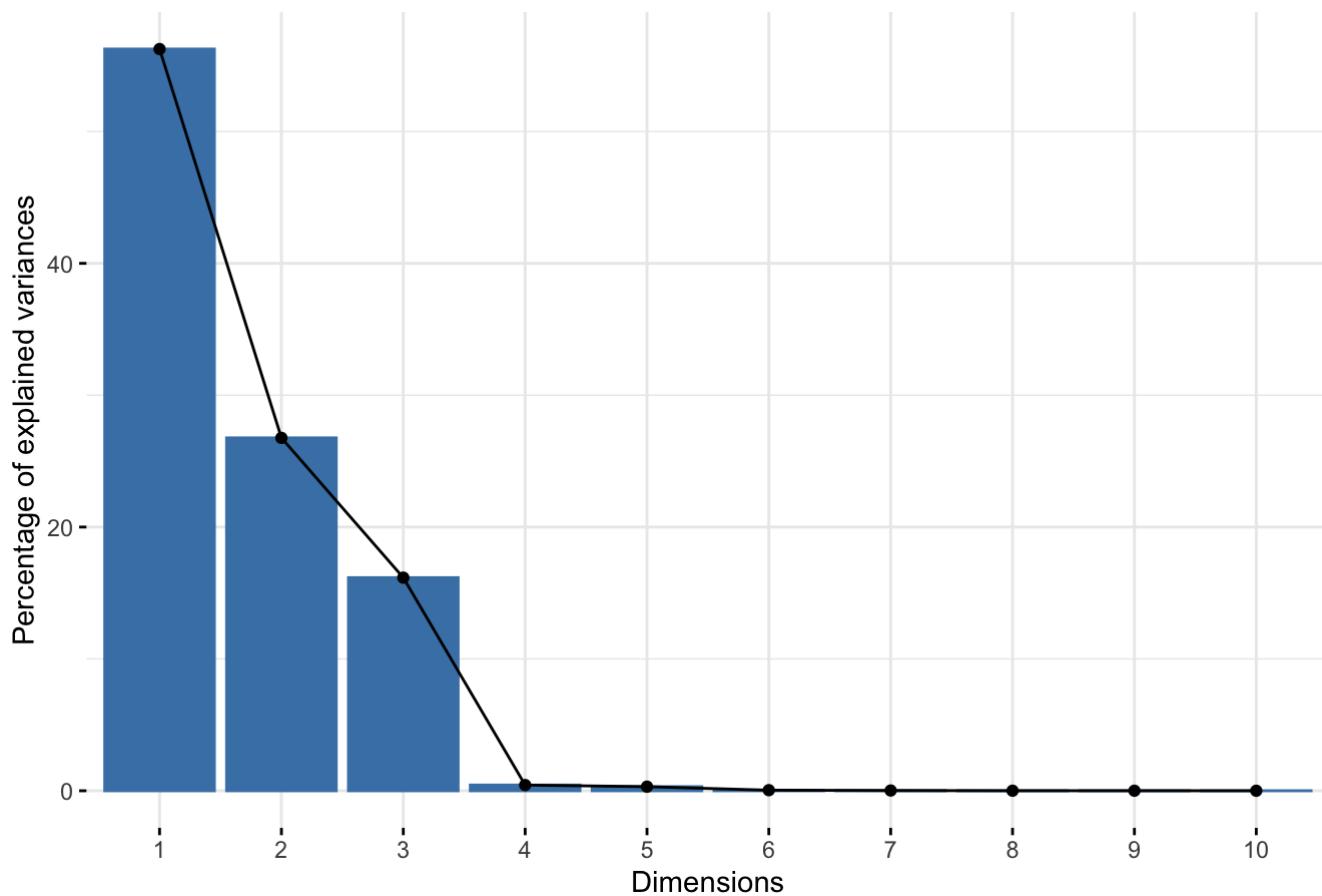
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped
```



```
require(showtext)
library(FactoMineR)
require(FactoMineR)
require(factoextra)

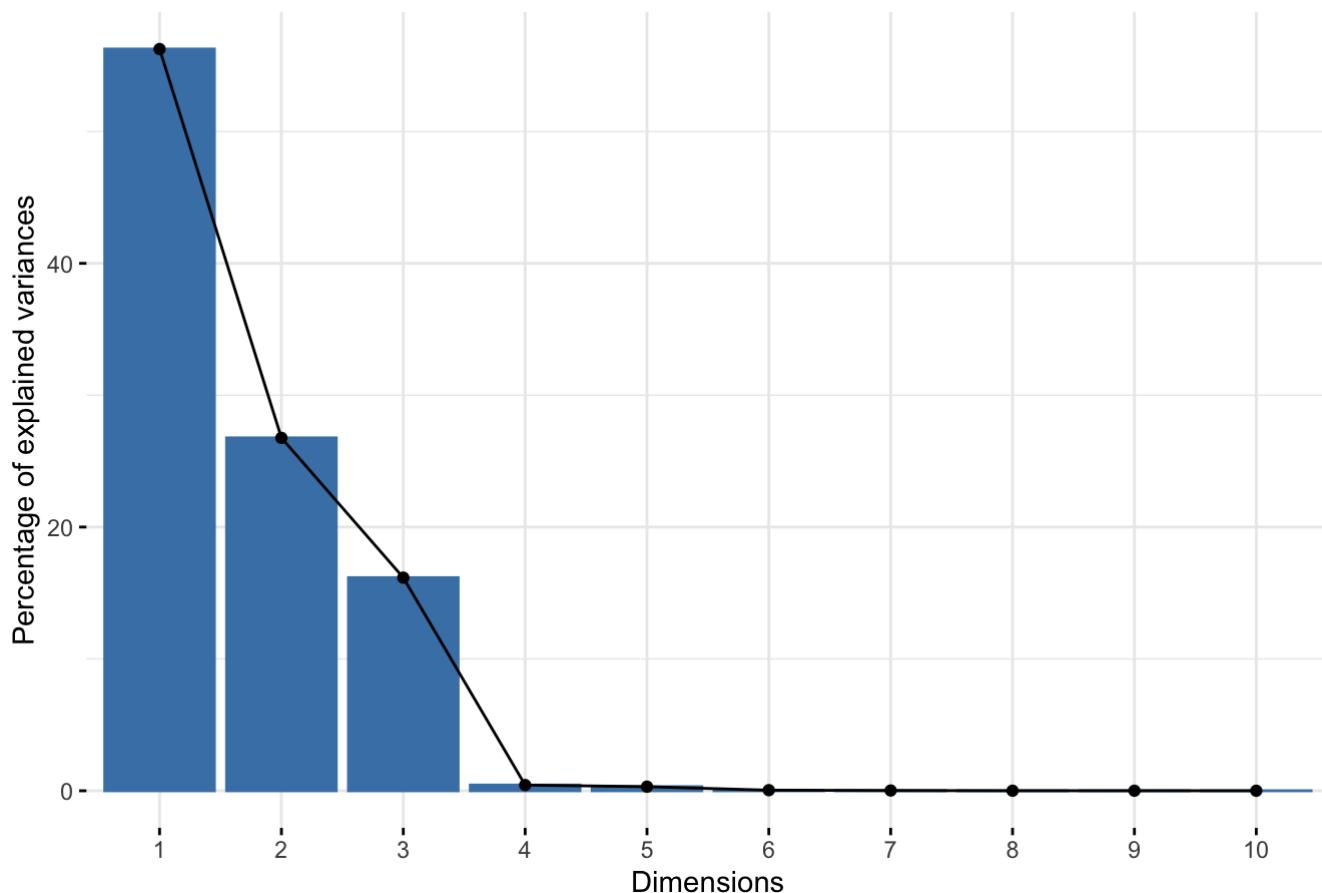
fviz_eig(pralPCA)
```

Scree plot

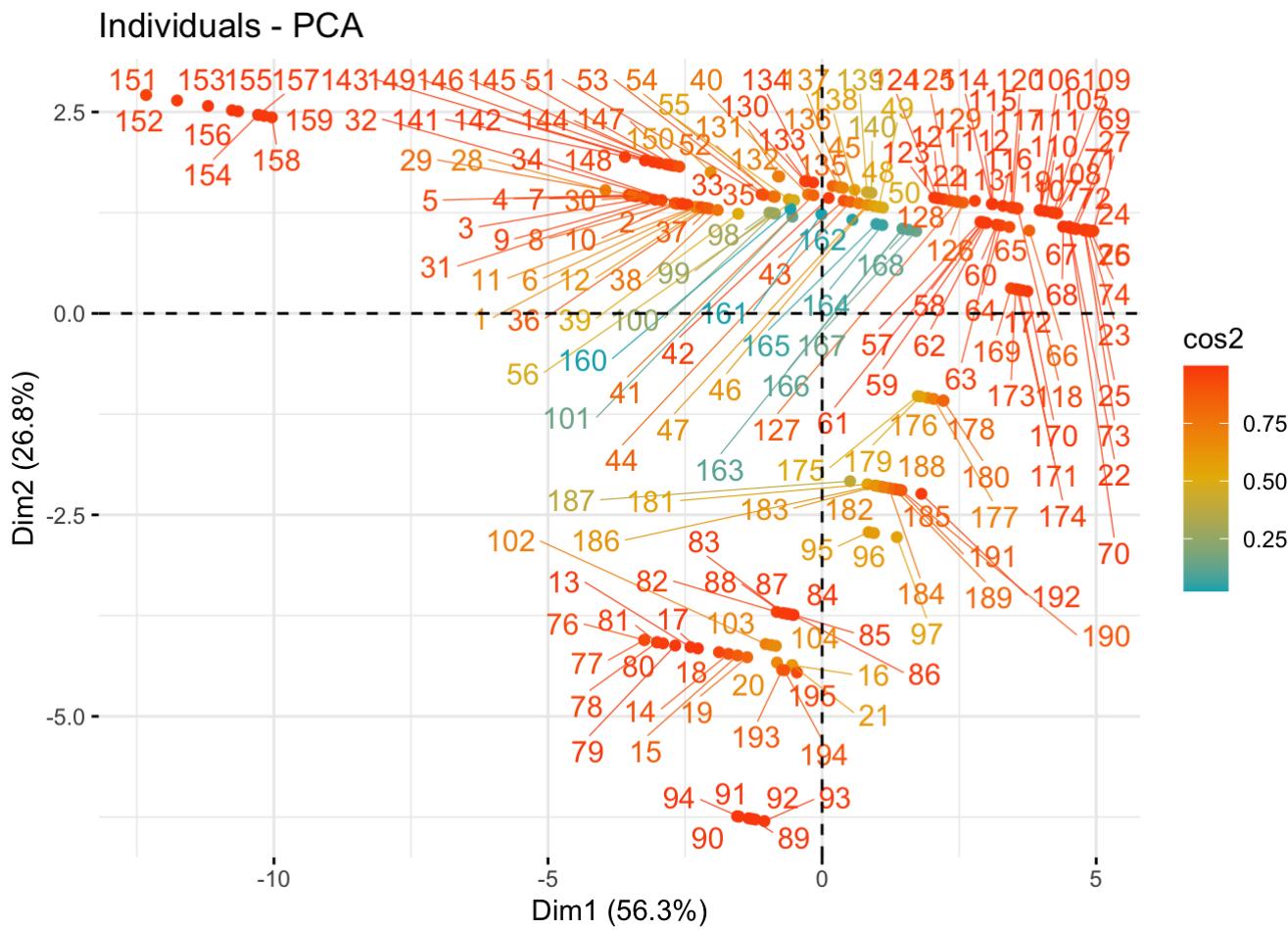


```
library(factoextra)  
fviz_eig(pralPCA)
```

Scree plot

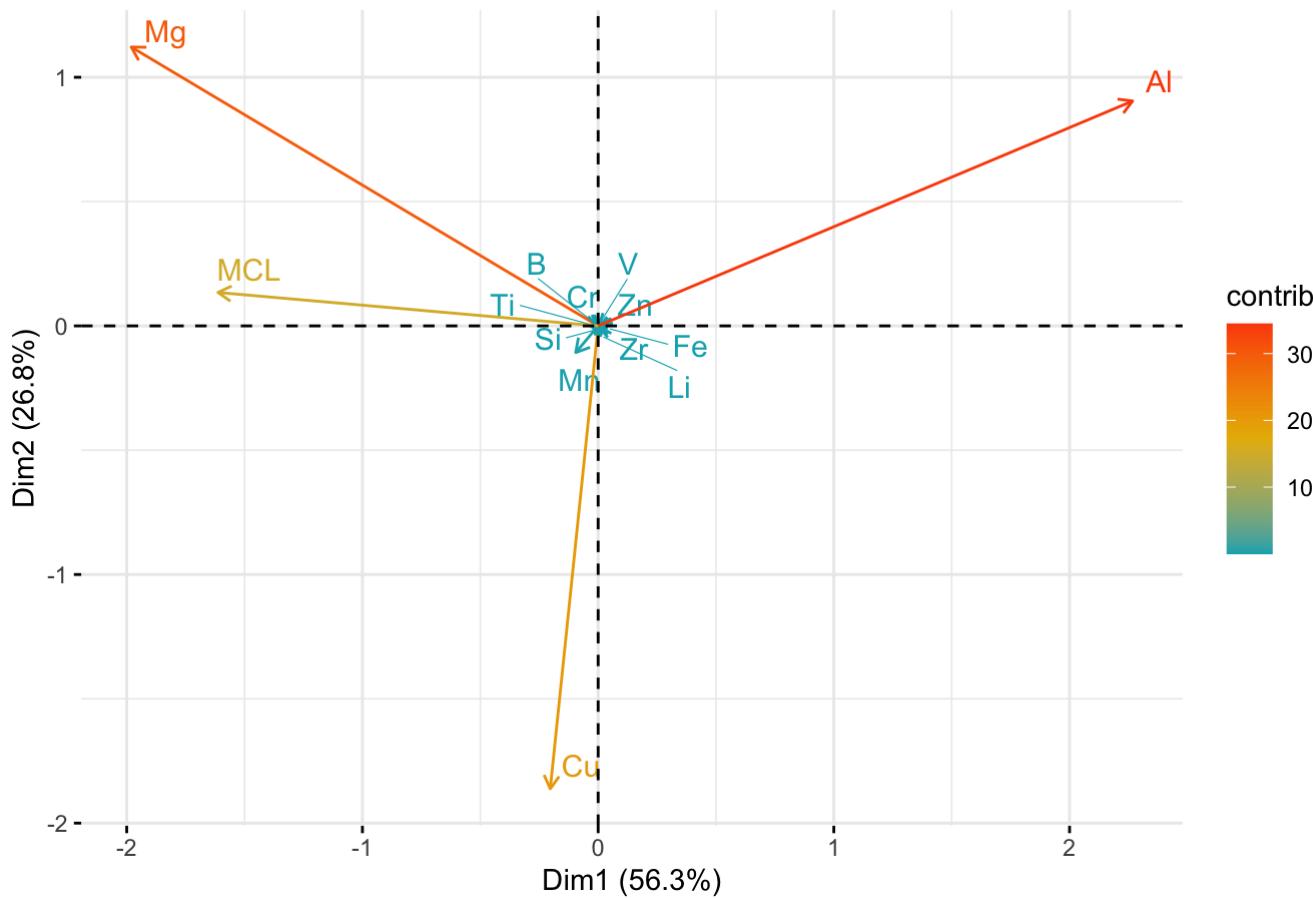


```
fviz_pca_ind(pralPCA,
  col.ind = "cos2", # Color by the quality of representation
  gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
  repel = TRUE      # Avoid text overlapping
)
```



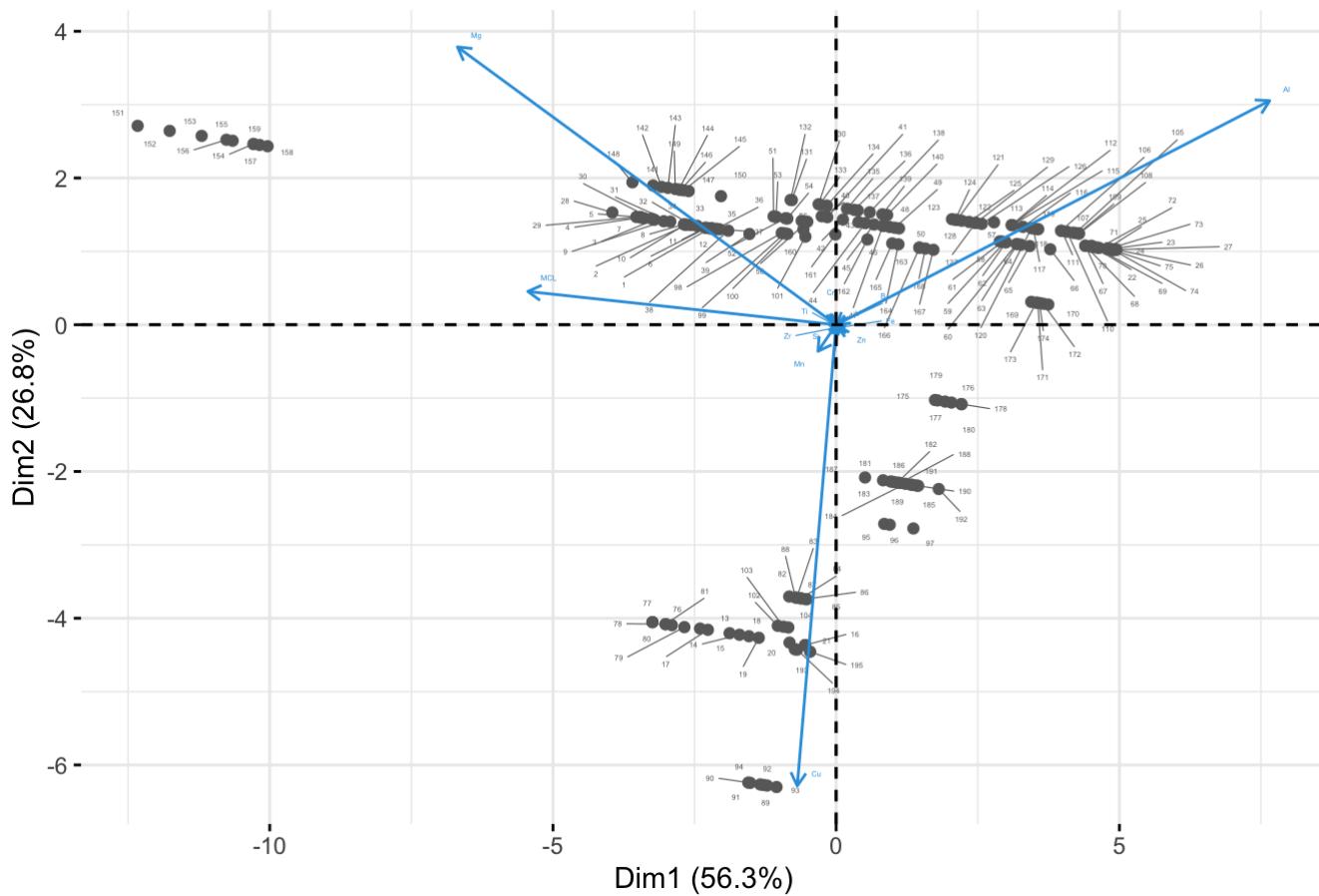
```
fviz_pca_var(pralPCA,
             col.var = "contrib", # Color by contributions to the PC
             gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
             repel = TRUE        # Avoid text overlapping
)
```

Variables - PCA



```
fviz_pca_biplot(pralPCA, repel = TRUE, labels = TRUE,
                 col.var = "#2E9FDF", # Variables color
                 col.ind = "#696969" # Individuals color
               )
```

PCA - Biplot

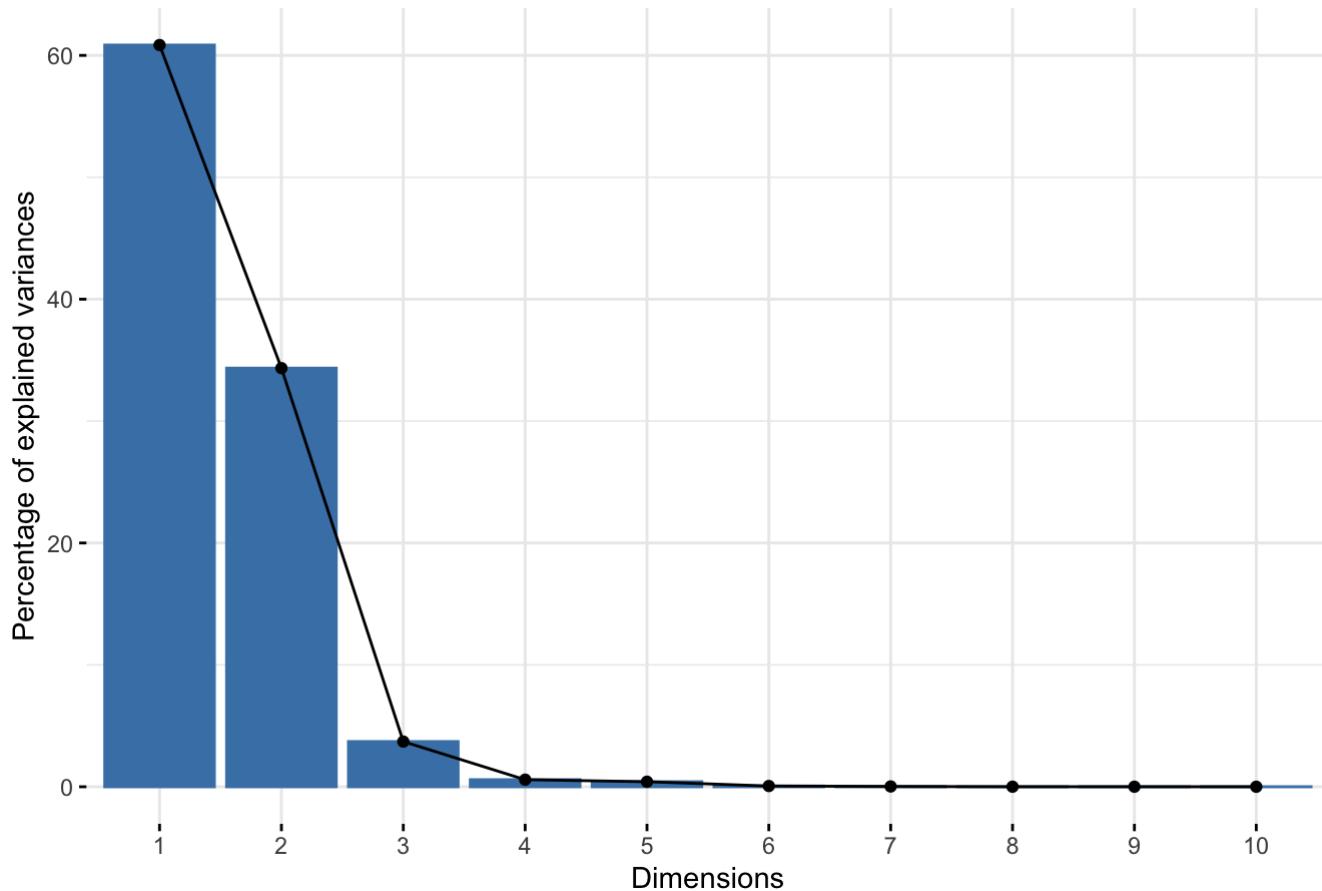


```
#MLR using only Al, Mg, Cu while looking at MCL
LMal2 = lm(MCL ~ Mg + Cu + Al, data = al)
summary(LMal2)
```

```
##
## Call:
## lm(formula = MCL ~ Mg + Cu + Al, data = al)
##
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -3.7049 -1.1715 -0.3048  0.7032  9.3522 
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) -121.6902    35.2737  -3.450 0.000690 ***
## Mg          1.8080     0.3707   4.877 2.26e-06 ***
## Cu          1.6661     0.3944   4.224 3.71e-05 ***
## Al          1.2390     0.3542   3.498 0.000583 ***
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 1.926 on 191 degrees of freedom
## Multiple R-squared:  0.3006, Adjusted R-squared:  0.2897 
## F-statistic: 27.37 on 3 and 191 DF,  p-value: 9.054e-15
```

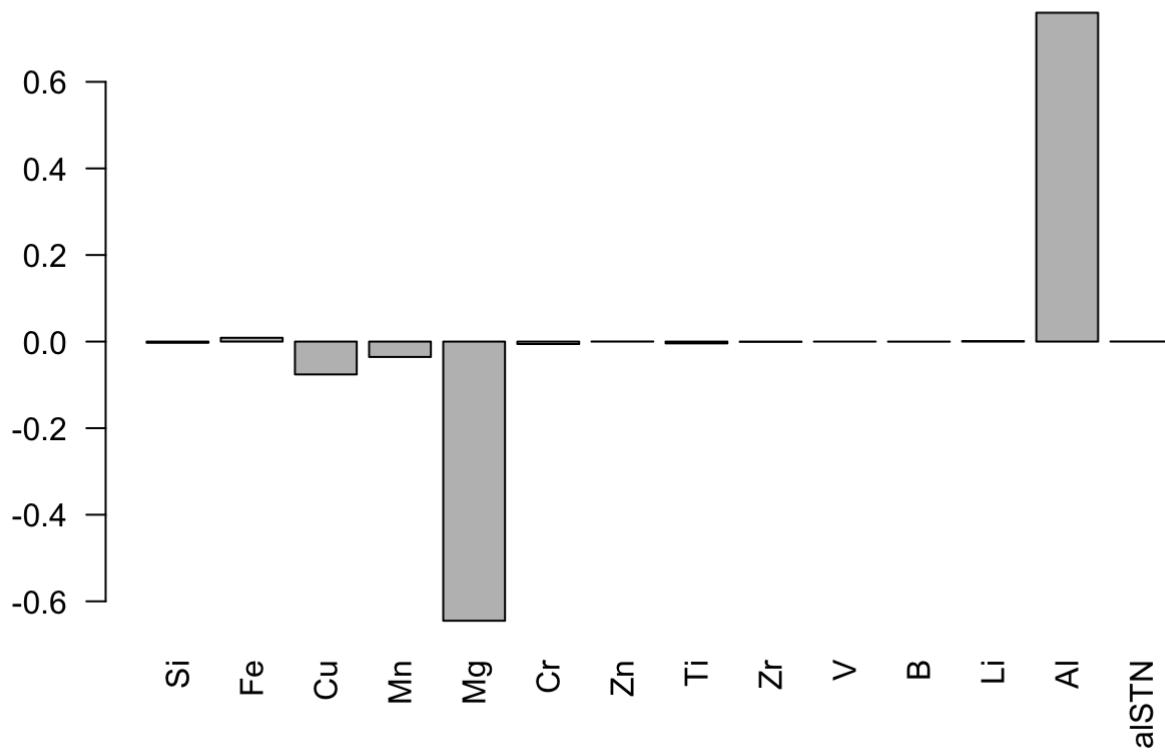
```
pralstn <- cbind(Si,Fe,Cu,Mn,Mg,Cr,Zn,Ti,Zr,V,B,Li,Al,alSTN)
pcaalstn = prcomp(pralstn, scale=F)
fviz_eig(pcaalstn)
```

Scree plot



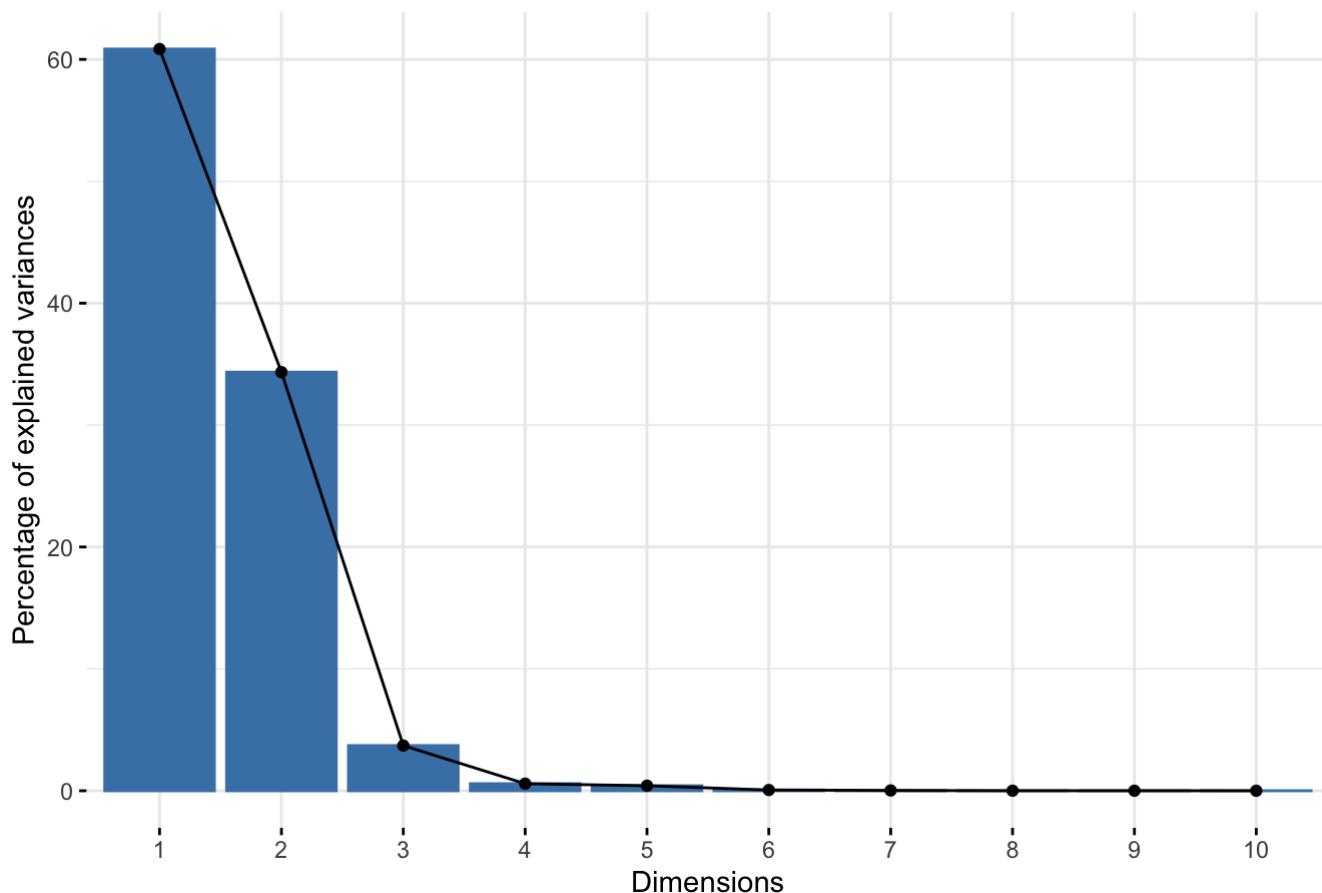
```
barplot(pcaalstn$rotation[,1], main="PC 1 Loadings Plot", las=2)
```

PC 1 Loadings Plot

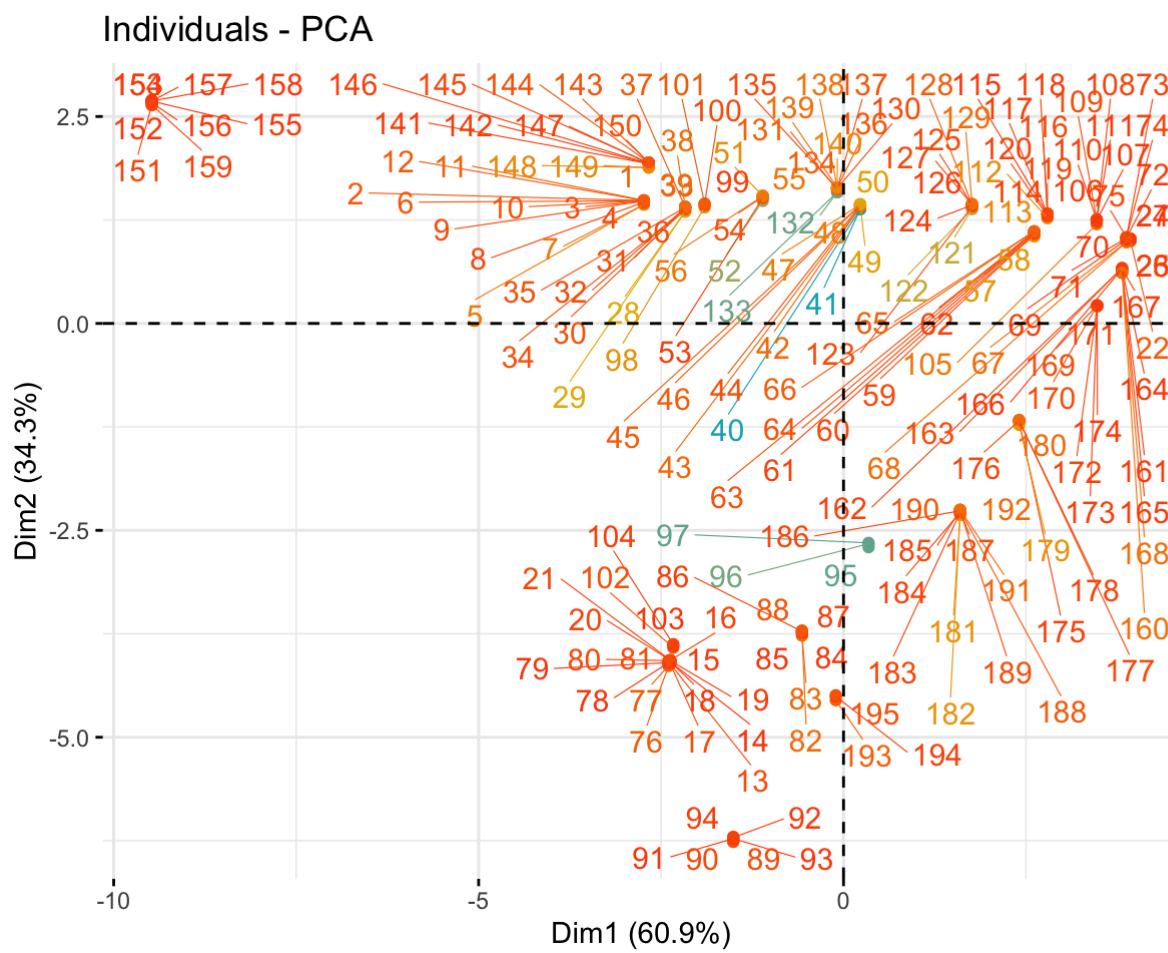


```
fviz_eig(pcaalstn)
```

Scree plot

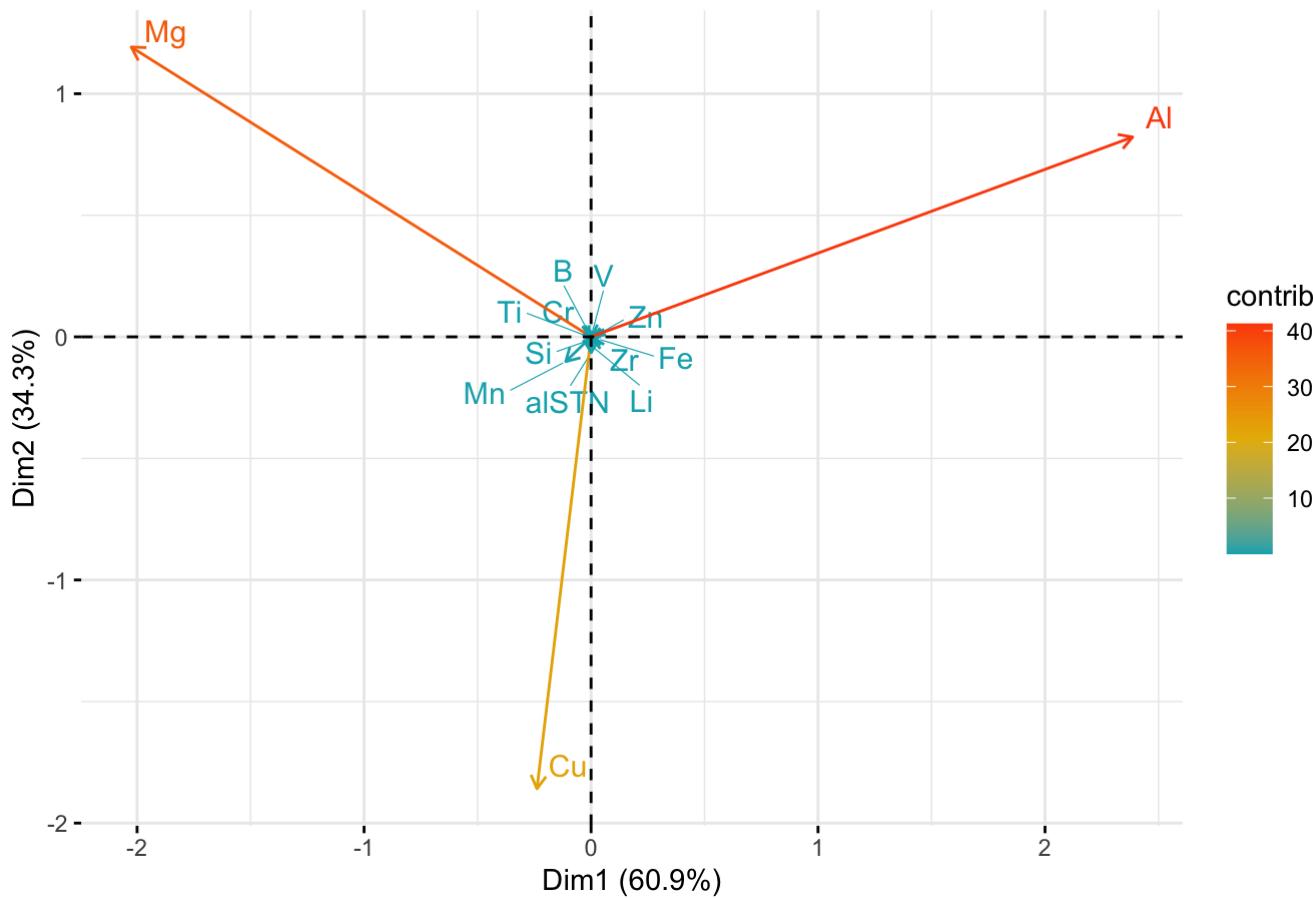


```
fviz_pca_ind(pcaalstn,
              col.ind = "cos2", # Color by the quality of representation
              gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
              repel = TRUE      # Avoid text overlapping
            )
```



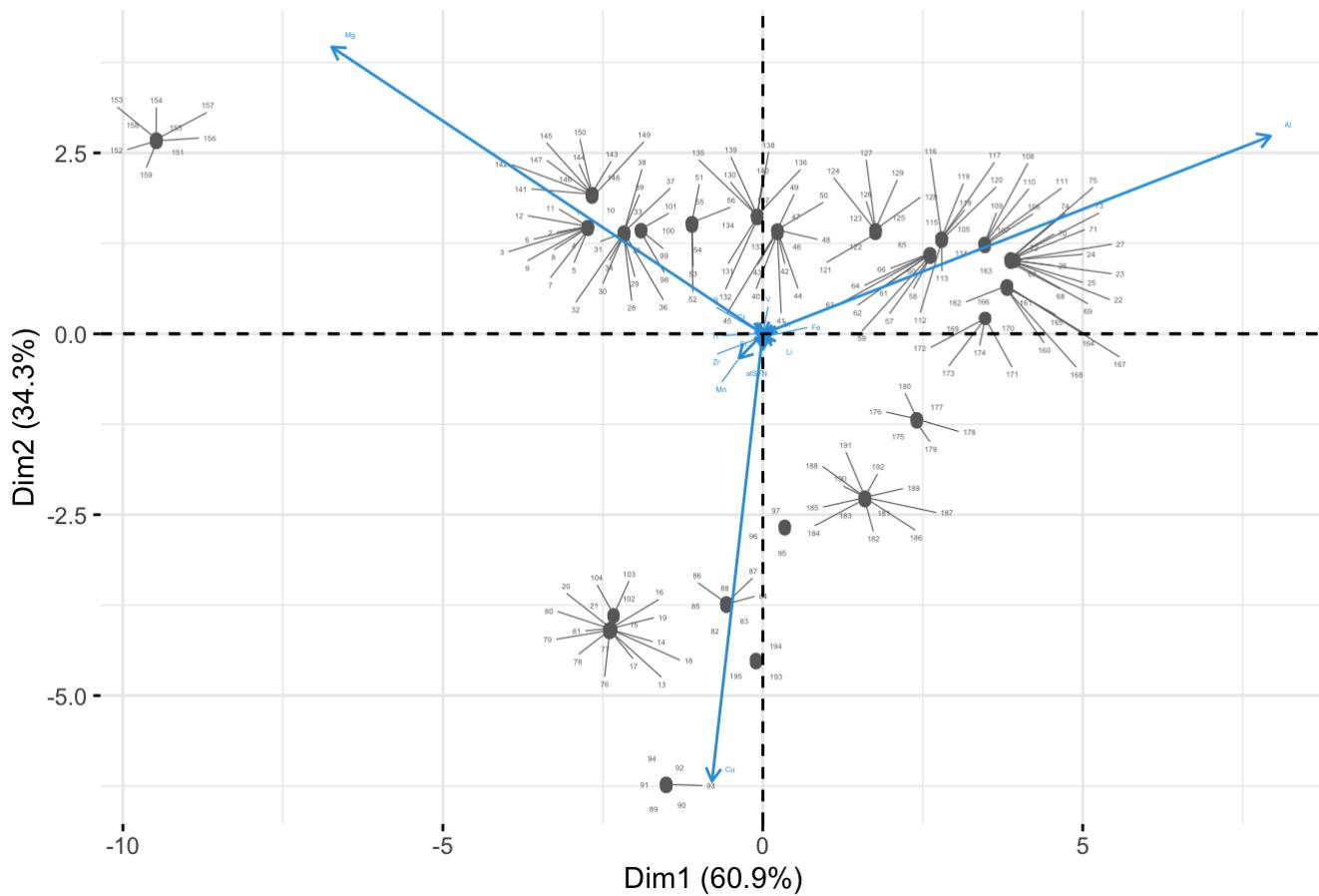
```
fviz_pca_var(pcaalstn,
              col.var = "contrib", # Color by contributions to the PC
              gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
              repel = TRUE      # Avoid text overlapping
            )
```

Variables - PCA



```
fviz_pca_biplot(pcaalstn, repel = TRUE, labels = TRUE,  
                 col.var = "#2E9FDF", # Variables color  
                 col.ind = "#696969" # Individuals color  
)
```

PCA - Biplot



```

sSi <-as.numeric(stlrf_STN$Si)
sC <-as.numeric(stlrf_STN$C)
sCu <-as.numeric(stlrf_STN$Cu)
sMn <-as.numeric(stlrf_STN$Mn)
sMo <-as.numeric(stlrf_STN$Mo)
sCr <-as.numeric(stlrf_STN$Cr)
sN <-as.numeric(stlrf_STN$N)
sTi <-as.numeric(stlrf_STN$Ti)
sP <-as.numeric(stlrf_STN$P)
sV <-as.numeric(stlrf_STN$V)
sB <-as.numeric(stlrf_STN$B)
sNi <-as.numeric(stlrf_STN$Ni)
sAl <-as.numeric(stlrf_STN$Al)
sNb <-as.numeric(stlrf_STN$Nb)
sCo <-as.numeric(stlrf_STN$Co)
sS <-as.numeric(stlrf_STN$S)
STN <-as.numeric(stlrf_STN$Strain)
TCL <- as.numeric(stlrf TCL$TCL)

```

```
prasstl <- cbind(sSi,sC,sCu,sMn,sMo,sCr,sN,sTi,sP,sV,sB,sAl,sNb,sCo,ss,STN)
```

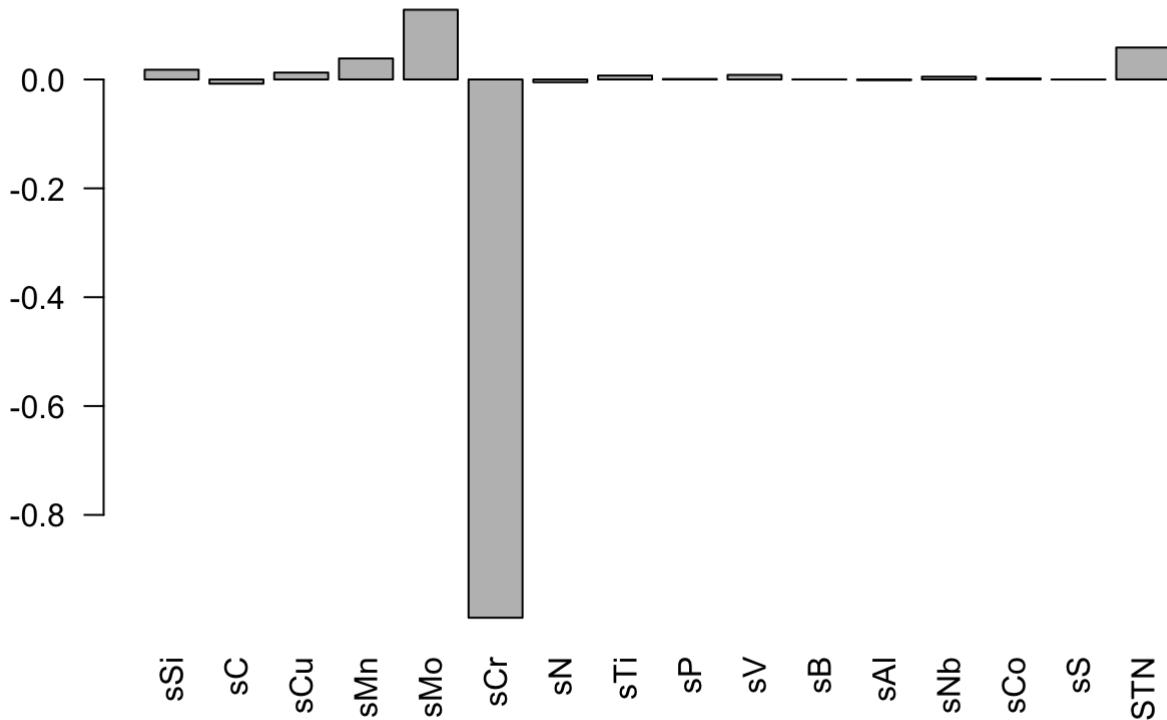
```
pcasstl = prcomp(prasstl, scale=F)

summary(pcasstl)
```

```
## Importance of components:
##                               PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation     3.3093  1.4020  1.3052  1.03458 0.53020 0.50246 0.21049
## Proportion of Variance 0.6703  0.1203  0.1043  0.06551 0.01721 0.01545 0.00271
## Cumulative Proportion  0.6703  0.7906  0.8948  0.96035 0.97756 0.99301 0.99572
##                               PC8      PC9      PC10     PC11     PC12     PC13     PC14
## Standard deviation     0.1669  0.14092 0.11861 0.06646 0.04400 0.04068 0.008156
## Proportion of Variance 0.0017  0.00122 0.00086 0.00027 0.00012 0.00010 0.000000
## Cumulative Proportion  0.9974  0.99864 0.99950 0.99977 0.99989 0.99999 1.000000
##                               PC15     PC16
## Standard deviation     0.005316 0.001222
## Proportion of Variance 0.000000 0.000000
## Cumulative Proportion  1.000000 1.000000
```

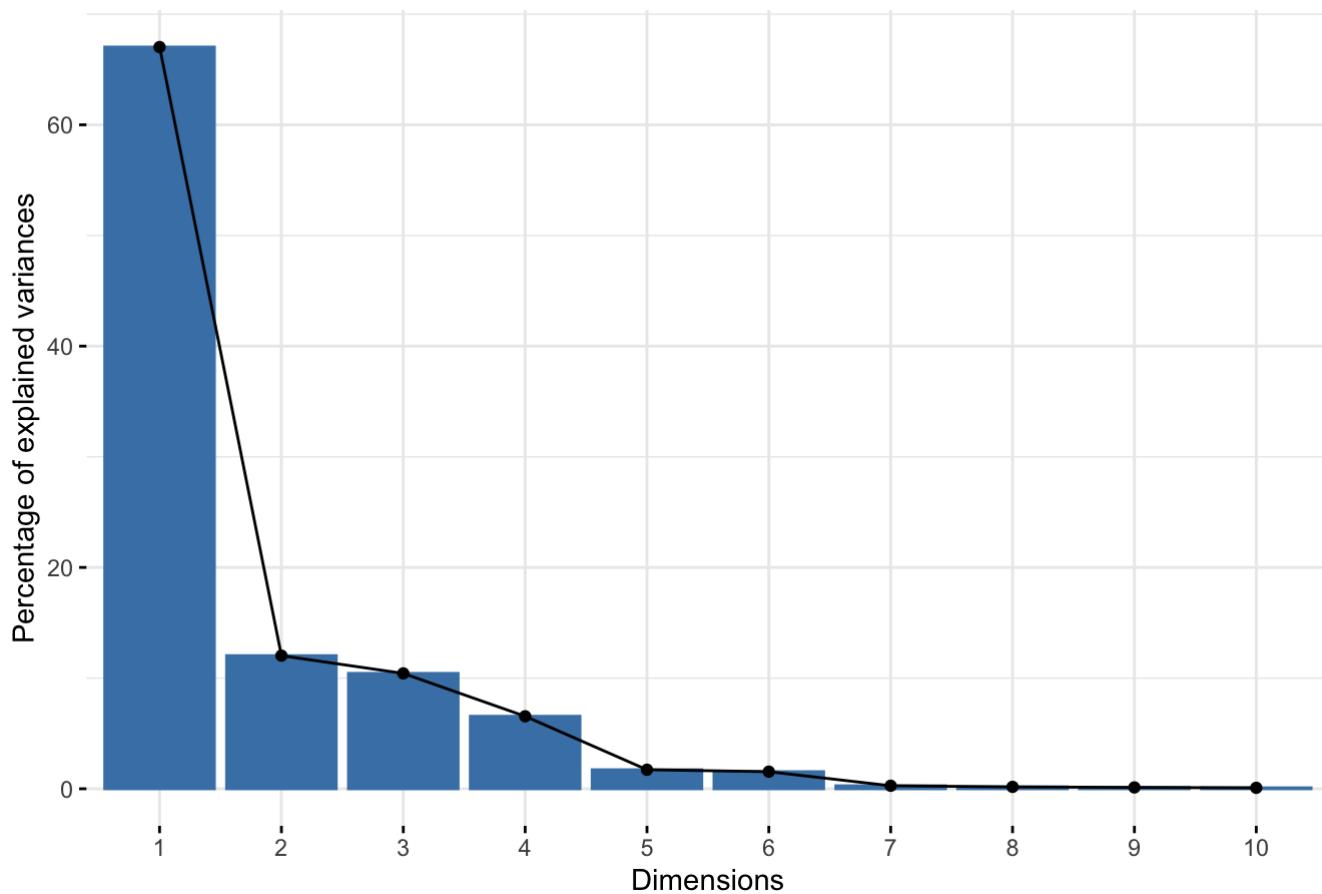
```
barplot(pcasstl$rotation[,1], main="PC 1 Loadings Plot", las=2)
```

PC 1 Loadings Plot



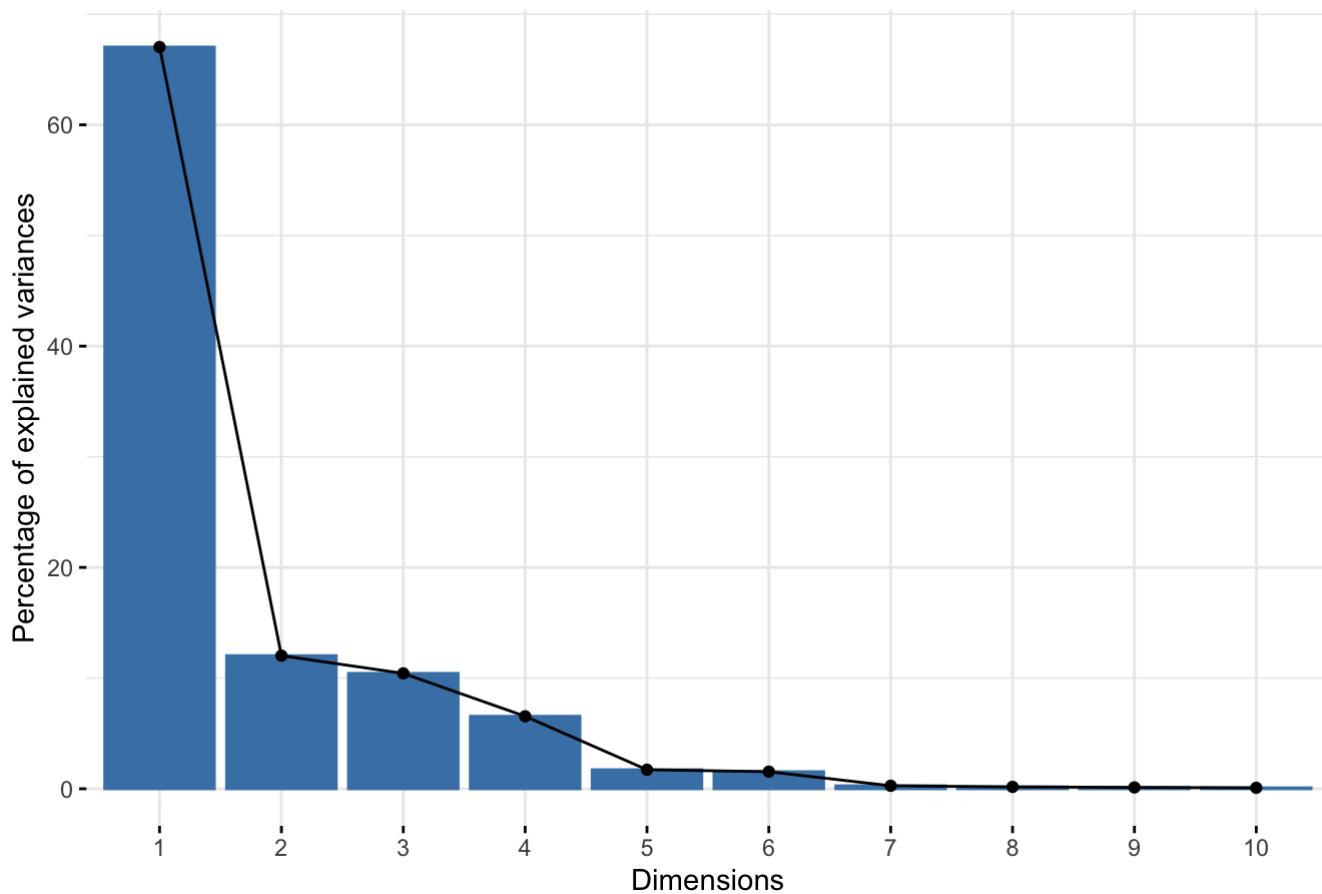
```
fviz_eig(pcasstl)
```

Scree plot



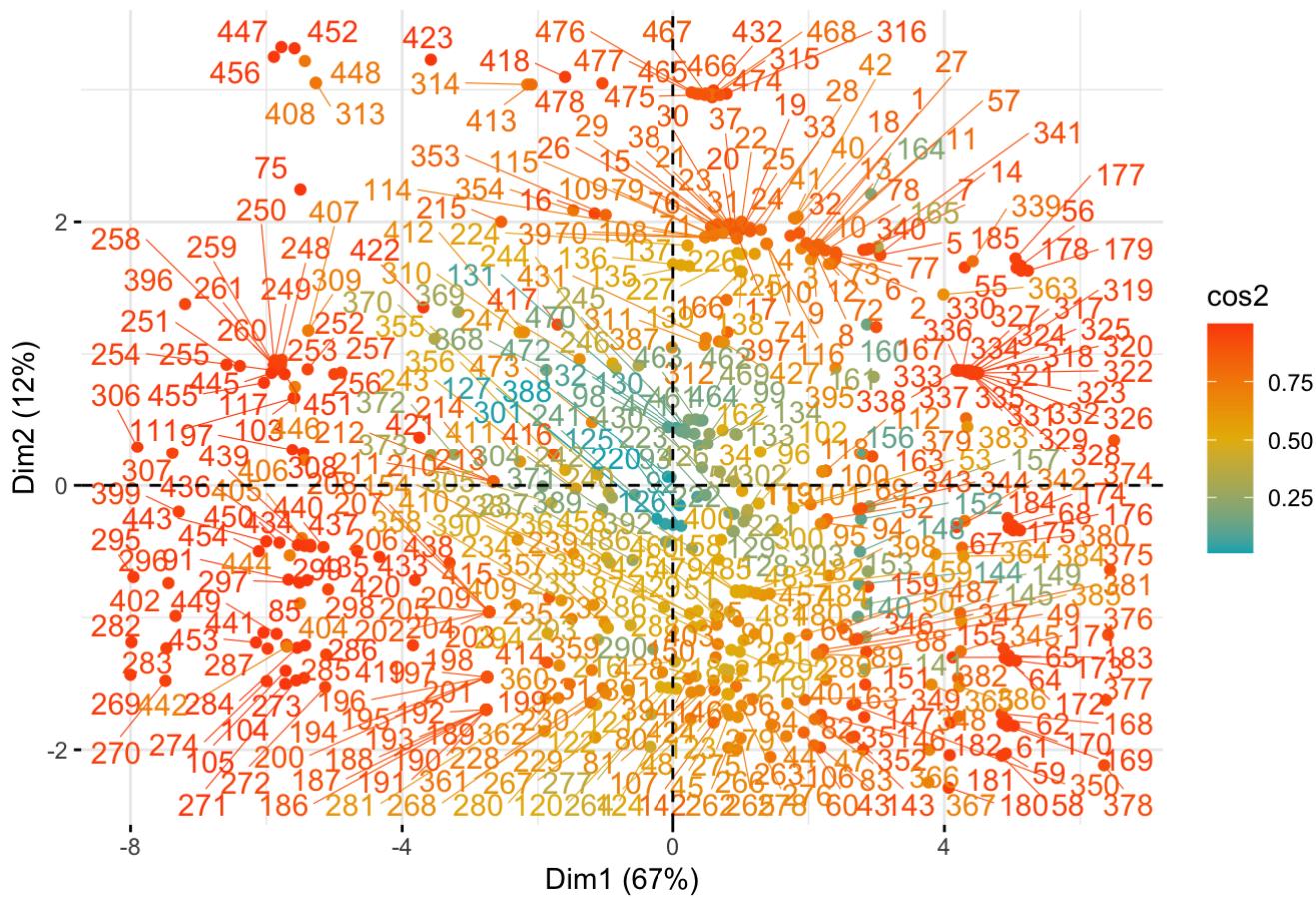
```
library(factoextra)
fviz_eig(pcass1)
```

Scree plot



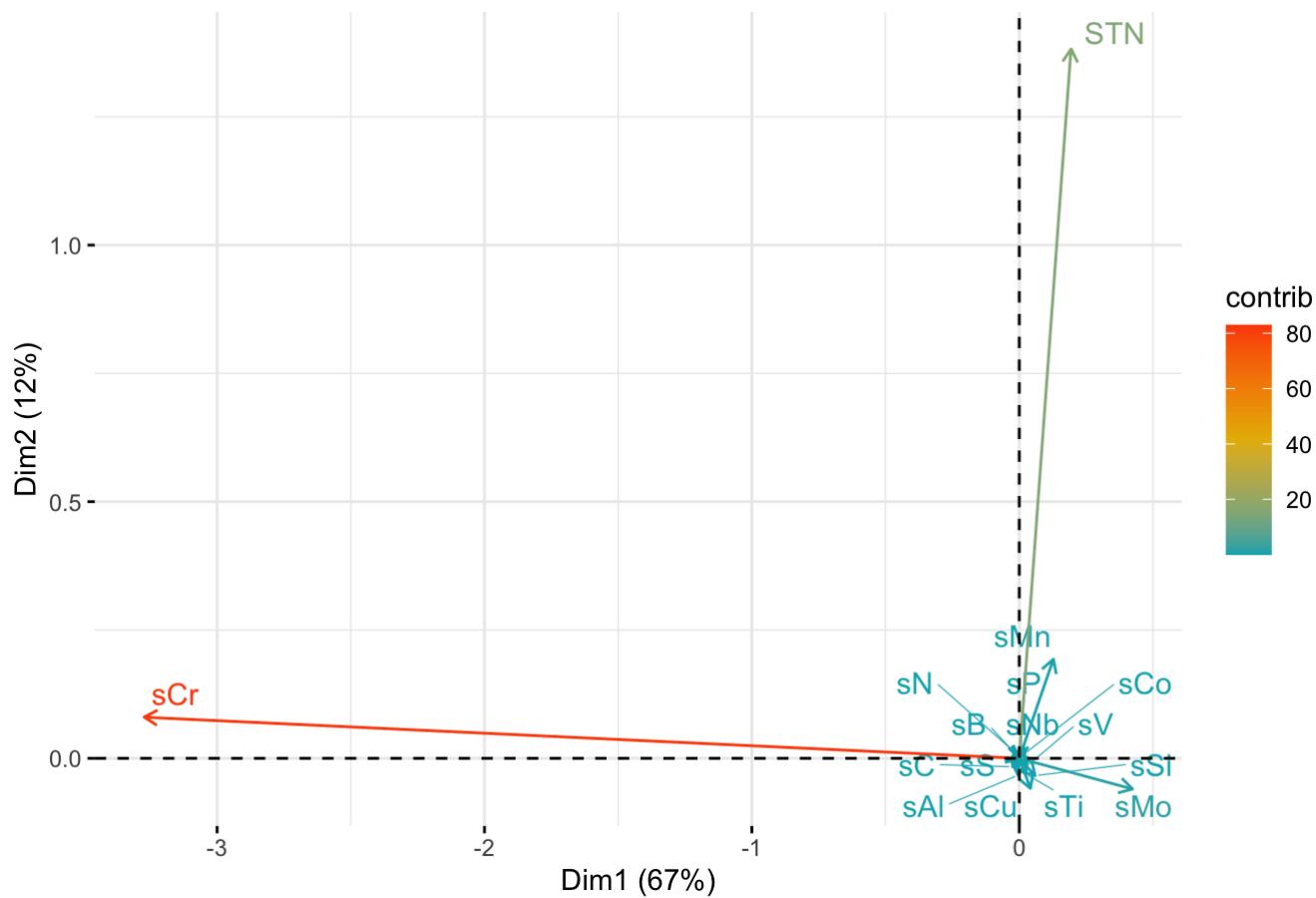
```
fviz_pca_ind(pcass1,
  col.ind = "cos2", # Color by the quality of representation
  gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
  repel = TRUE      # Avoid text overlapping
)
```

Individuals - PCA



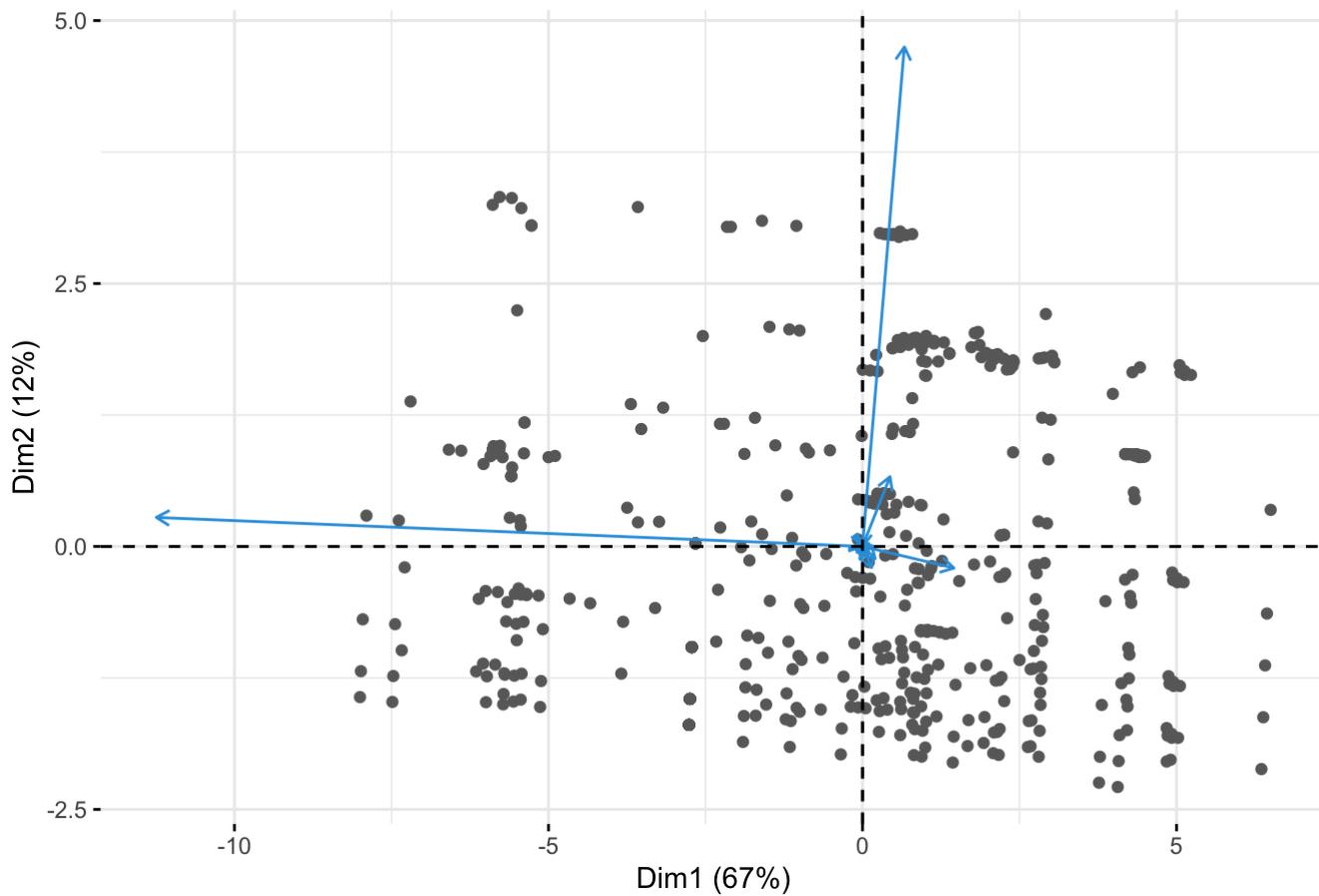
```
fviz_pca_var(pcassl,
  col.var = "contrib", # Color by contributions to the PC
  gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
  repel = TRUE      # Avoid text overlapping
)
```

Variables - PCA



```
fviz_pca_biplot(pcasstl, repel = TRUE, label = TRUE,  
                 col.var = "#2E9FDF", # Variables color  
                 col.ind = "#696969" # Individuals color  
)
```

PCA - Biplot



```
prasstl2 <- cbind(sSi,sC,sCu,sMn,sMo,sCr,sN,sTi,sP,sV,sB,sAl,sNb,sCo,sS,TCL)
```

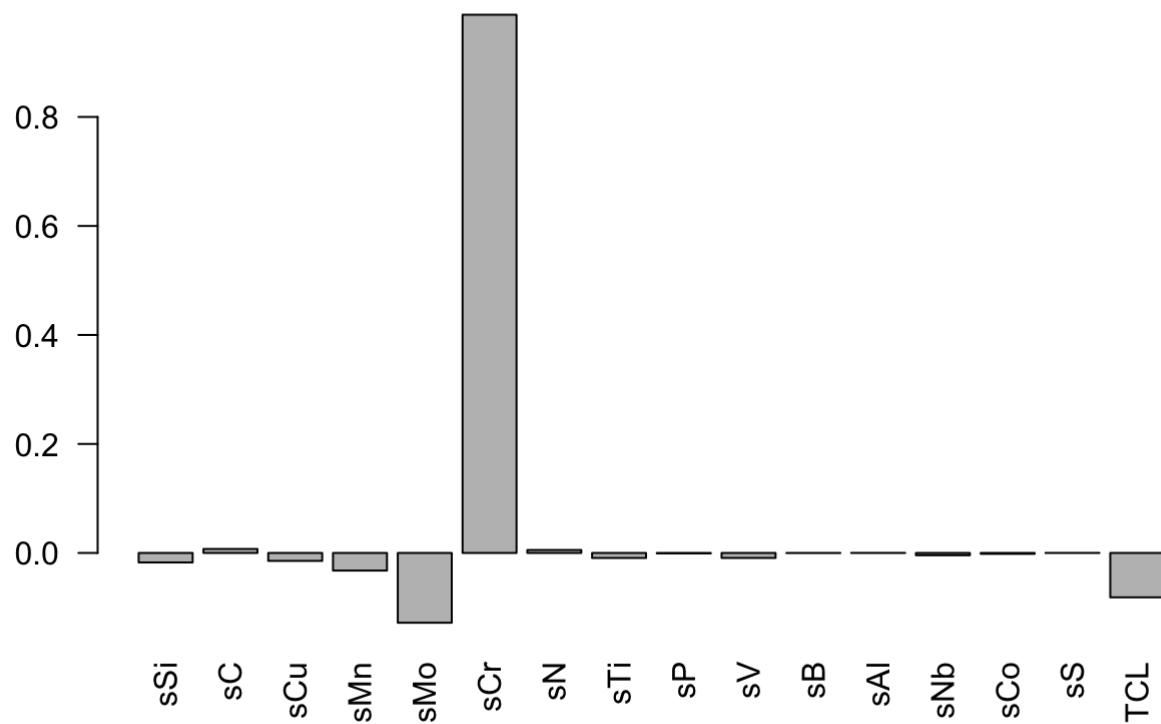
```
pcatstl = prcomp(prasstl2, scale=F)
```

```
summary(pcatstl)
```

```
## Importance of components:
##                 PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation 4.1983 3.2978 1.29642 1.03397 0.53302 0.50704 0.20541
## Proportion of Variance 0.5525 0.3409 0.05268 0.03351 0.00891 0.00806 0.00132
## Cumulative Proportion 0.5525 0.8933 0.94601 0.97952 0.98842 0.99648 0.99780
##                  PC8      PC9      PC10     PC11     PC12     PC13     PC14
## Standard deviation 0.16650 0.14091 0.12003 0.06655 0.04411 0.04050 0.008163
## Proportion of Variance 0.00087 0.00062 0.00045 0.00014 0.00006 0.00005 0.000000
## Cumulative Proportion 0.99867 0.99929 0.99975 0.99988 0.99995 1.00000 1.000000
##                  PC15     PC16
## Standard deviation 0.005308 0.001207
## Proportion of Variance 0.000000 0.000000
## Cumulative Proportion 1.000000 1.000000
```

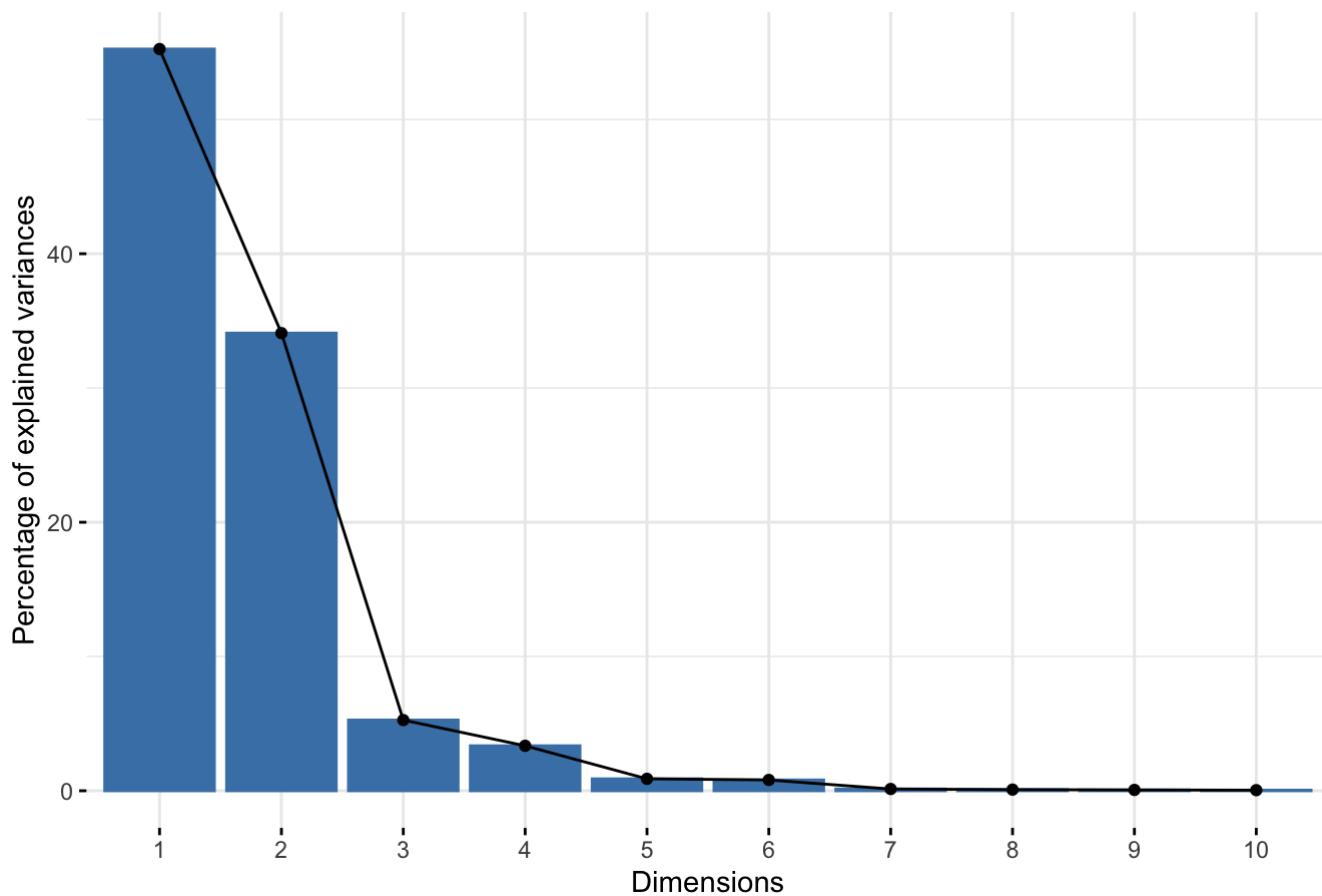
```
barplot(pcatstl$rotation[,2], main="PC 1 Loadings Plot", las=2)
```

PC 1 Loadings Plot



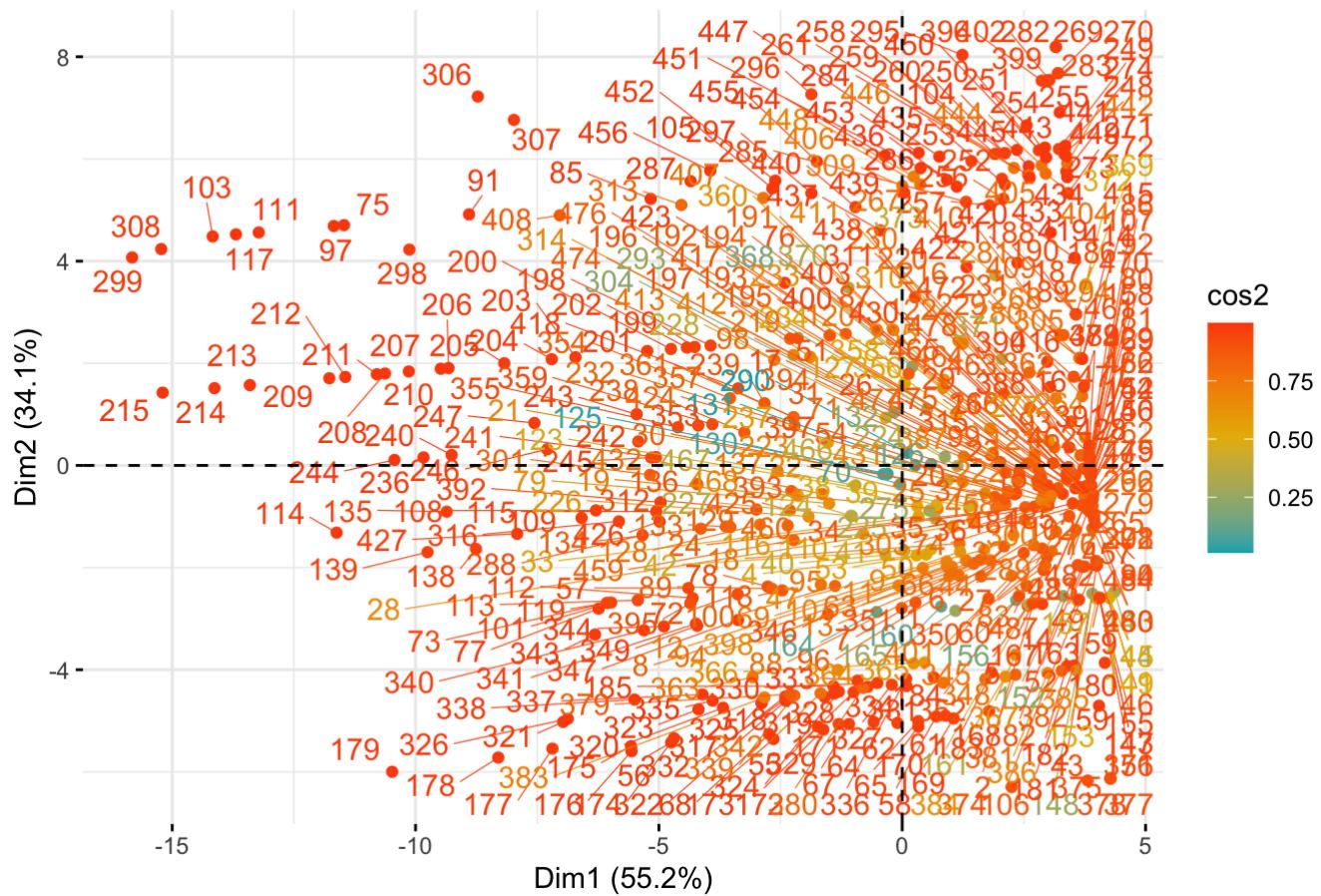
```
fviz_eig(pcatst1)
```

Scree plot



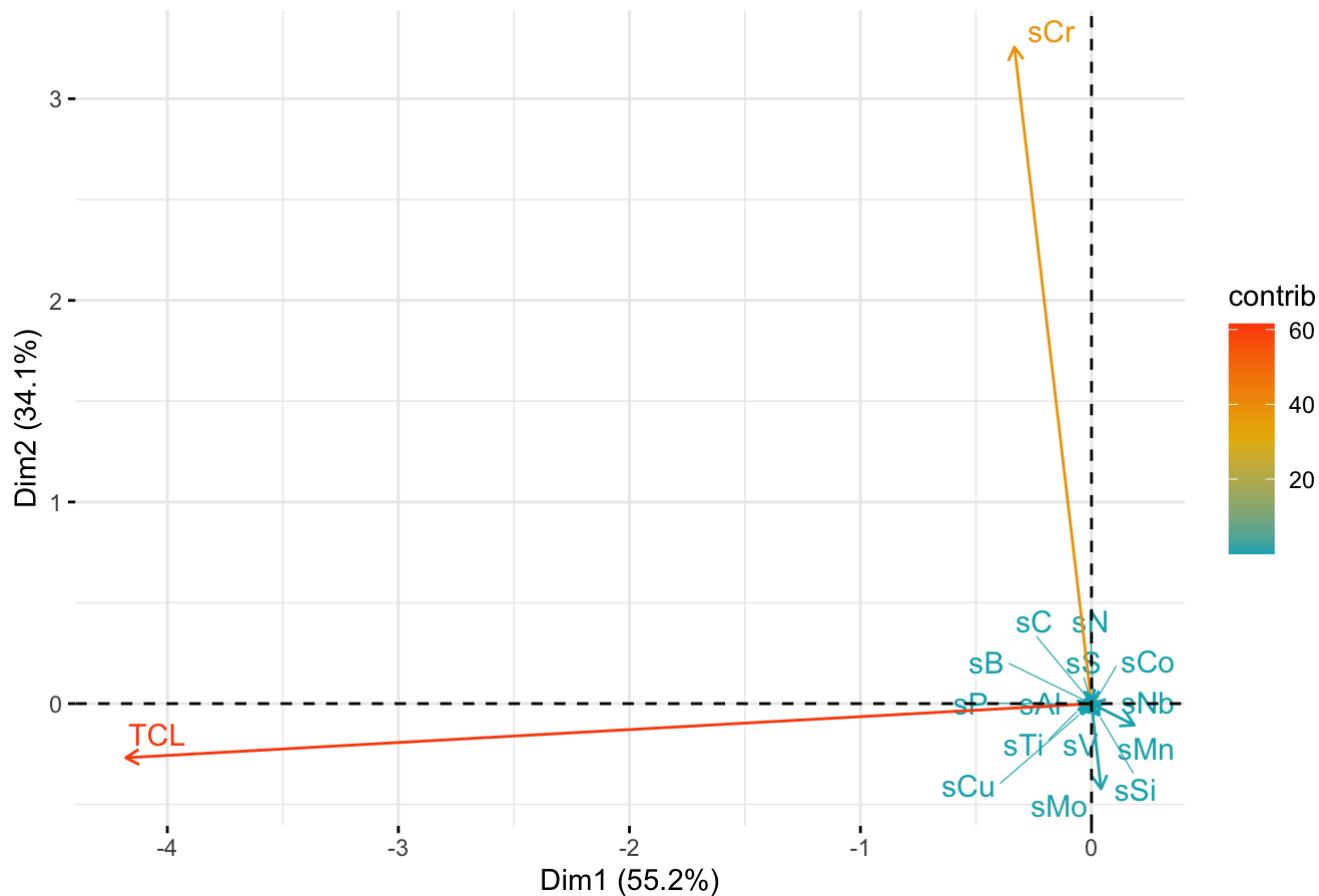
```
fviz_pca_ind(pcatstl,  
             col.ind = "cos2", # Color by the quality of representation  
             gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),  
             repel = TRUE      # Avoid text overlapping  
           )
```

Individuals - PCA



```
fviz_pca_var(pcatstl,
             col.var = "contrib", # Color by contributions to the PC
             gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
             repel = TRUE      # Avoid text overlapping
)
```

Variables - PCA



```
fviz_pca_biplot(pcatstl, repel = TRUE, label = TRUE,  
                 col.var = "#2E9FDF", # Variables color  
                 col.ind = "#696969" # Individuals color  
)
```

PCA - Biplot

