

# Demo

```
library(rnaGinesis)
library(gridExtra)

##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:Biobase':
##
##      combine

## The following object is masked from 'package:BiocGenerics':
##
##      combine

library(grid)
library(ggplot2)
library(lattice)
library(reshape)
library(GGally)
library(NMF)

## Loading required package: pkgmaker
## Loading required package: registry
##
## Attaching package: 'pkgmaker'
## The following object is masked from 'package:base':
##
##      isNamespaceLoaded

## Loading required package: rngtools
## Loading required package: cluster
## NMF - BioConductor layer [OK] | Shared memory capabilities [NO: bigmemory] | Cores 23/24
##   To enable shared memory capabilities, try: install.extras('
## NMF
## ')

library(nnls)
library(moments)
mu <- rnaGinesis::mu
A <- rnaGinesis::A
```

## Generate simulated data

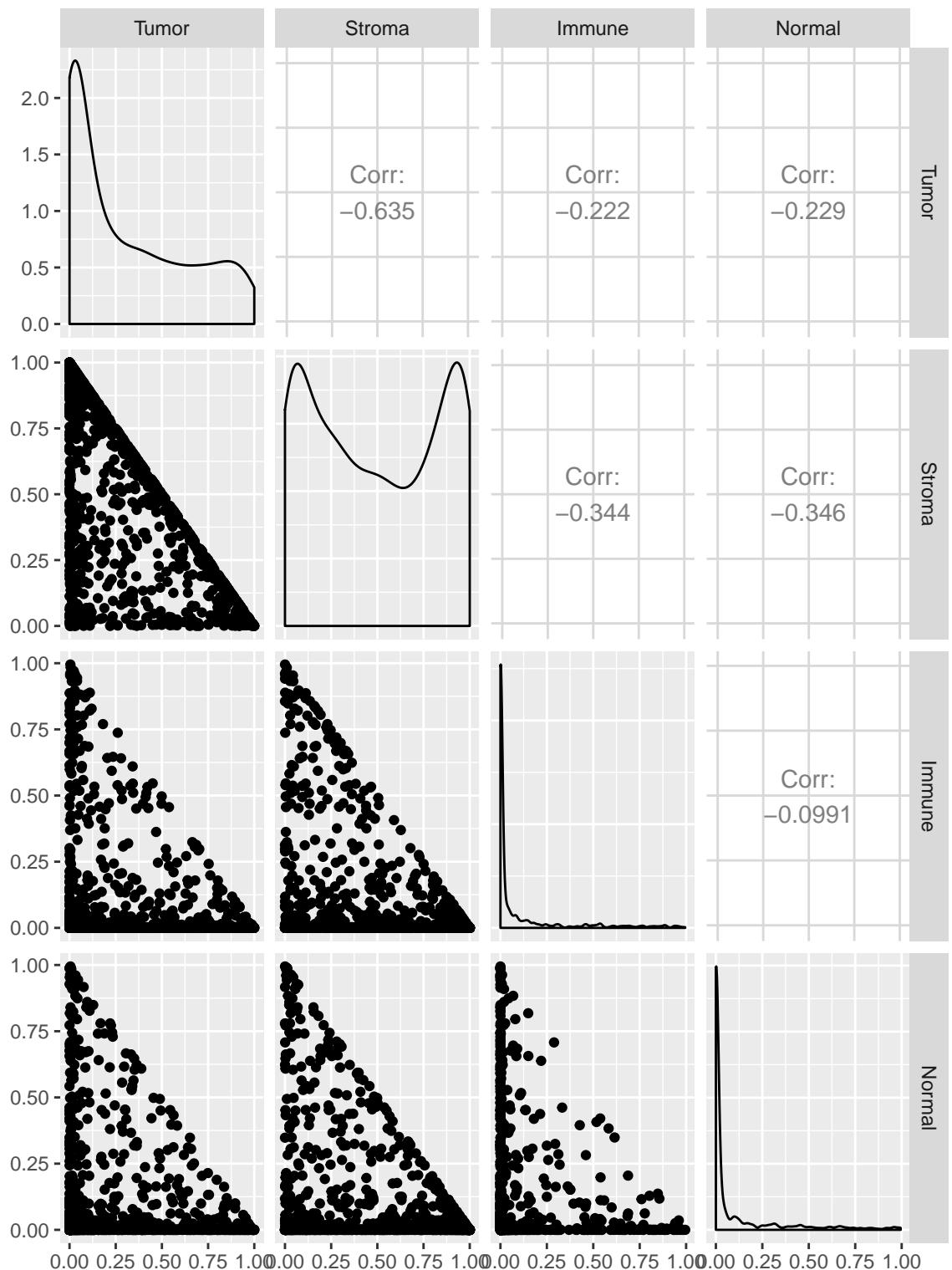
```
ngenes <- 500;
nsamples = 1000;
simresult <- Complete_simulation(A_tumor = A[1:ngenes,1:ngenes],
                                    mu_tumor = mu[1:ngenes],
                                    Samplesize = nsamples,
```

```
scaleFactor    = 100,
d.params      = c("Tumor"     = .3,
                  "Stroma"    = .5,
                  "Immune"    = .1,
                  "Normal"    = .1),
noise_setting = 1.5,
seed          = 1234)

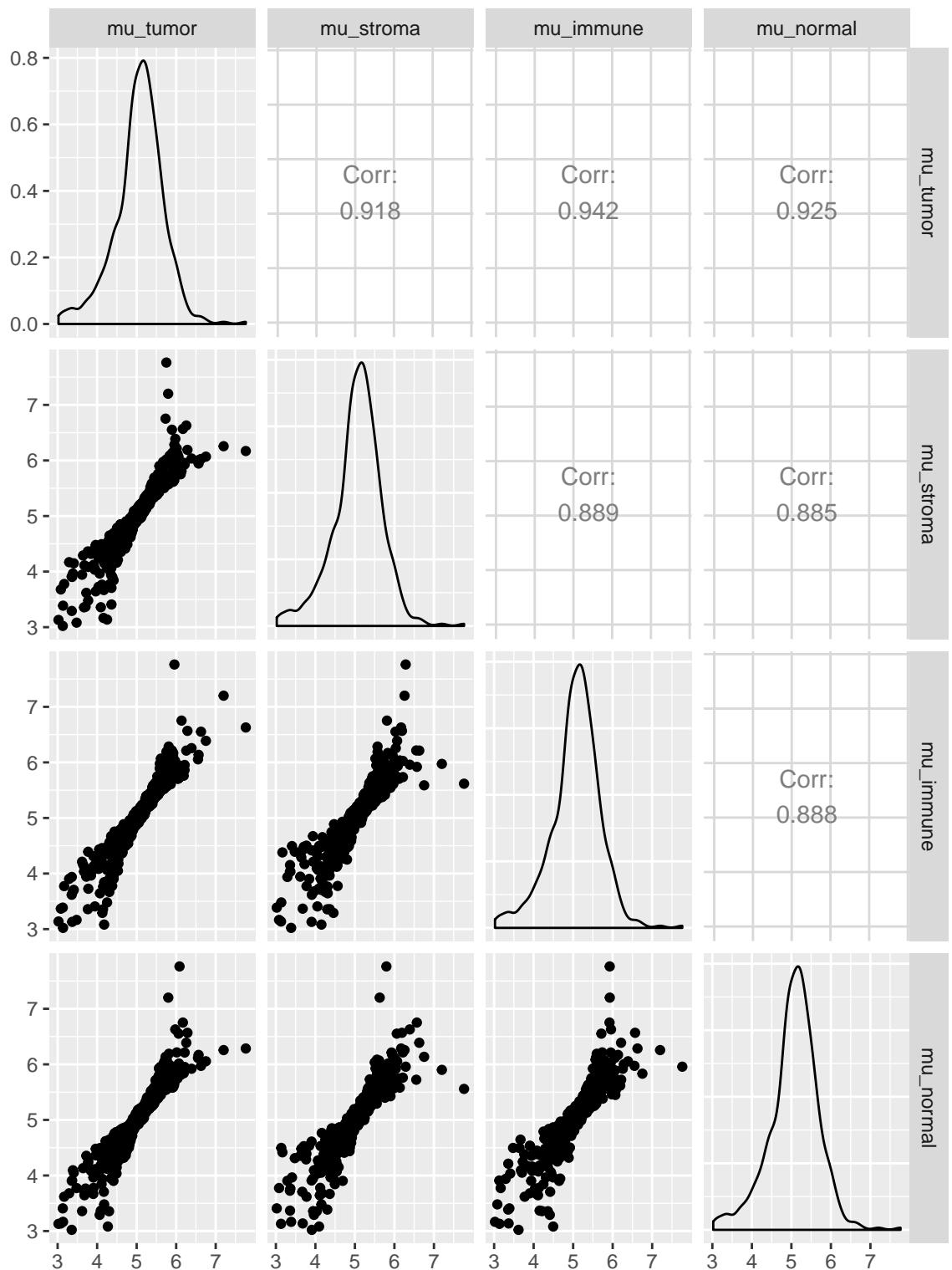
data   <- simresult[[1]]
W      <- simresult[[2]]
H      <- simresult[[3]]
rm("simresult")
rm("mu")
rm("A")
```

## Display data properties

```
ggpairs(data = data.frame(t(H)))
```



```
ggpairs(data = as.data.frame(lapply(data.frame(W),log10)))
```



Define functions

```

my_nnls <- function(W,X){
  # X = WH
  # solve for H
  pinv <- (solve(t(W) %*% W) %*% t(W))
  H <- pinv %*% X
  H[H<0] <- 0
  residual <- X - (W %*% H)

  ngenes = dim(W)[1]

  skew <- skewness(residual)
  kurt <- kurtosis(residual)

  residual.scaled <-
    sum((residual[residual>0])) /
    (ngenes * mean(mean(W)))

  return(list("H" = H,
             "residual" = residual.scaled,
             "skew" = skew,
             "kurt" = kurt))
}

fuzzy_nmf <- function(W,X){
  # X = WH
  # solve for H

  # start with nnls result, but don't let zeros happen
  nnls.result <- my_nnls(W,X)
  H <- nnls.result$H

  W.missing <- X - (W %*% H)
  W.missing[W.missing < 0] <- 0
  W.missing[W.missing < (mean(W.missing)*0.01)] <- (mean(W.missing)*0.01)

  H[H<mean(H)*0.01] <- mean(H)*0.01

  # add a dummy factor, populate it with garbage
  ngenes = dim(W)[1]
  nfactors = dim(W)[2]
  W <- cbind(W, W.missing)
  H <- rbind(H, nnls.result$residual)
  H.original <- H

  residual <- X - (W %*% H)
  exitcond <- TRUE
  # writeLines("-----")
  while(exitcond){
    # multiplicative NMF update, modified

    # stay close to known H
    H.new <- H * (t(W) %*% X) / (t(W) %*% W %*% H)
    # H[nfactors + 1] <- H.new[nfactors + 1]
}

```

```

# H[1:nfactors]    <- 0.5 * H.new[1:nfactors] + 0.5 * H.original[1:nfactors]
H <- 0.5 * H.new + 0.5 * H.original

# keep known W fixed
W.new <- W * ( X %*% t(H) ) / (W %*% H %*% t(H))
W[nfactors + 1] <- W.new[nfactors + 1]

new.residual <- X - (W %*% H)
change = sum(abs(residual)) /sum(abs(new.residual))
residual <- new.residual
exitcond <- change > 1.001

# print(H)
}
return(list("H" = H[1:nfactors],
           "residual" = H[nfactors+1]))
}

```

## Unmix single samples

perfect knowledge

```

known.set <- 1:4
nnls.H.perfect <- H[known.set,]
nnls.resids.perfect <- numeric(nsamples)
nnls.skew.perfect <- numeric(nsamples)
nnls.kurt.perfect <- numeric(nsamples)

nmf.H.perfect <- H[known.set,]
nmf.resids.perfect <- numeric(nsamples)

for(i in 1:nsamples){
  this_sample <- data[,i]

  nnls.result <- my_nnls(W[,known.set],this_sample)
  nmf.result <- fuzzy_nmf(W[,known.set],this_sample)

  nnls.H.perfect[,i] <- nnls.result$H
  nnls.resids.perfect[i] <- nnls.result$residual
  nnls.skew.perfect[i] <- nnls.result$skew
  nnls.kurt.perfect[i] <- nnls.result$kurt

  nmf.H.perfect[,i] <- nmf.result$H
  nmf.resids.perfect[i] <- nmf.result$residual
}

```

## imperfect knowledge

```
known.set <- 1:3
nnls.H.incomplete <- H[known.set,]
nnls.resids.incomplete <- numeric(nsamples)
nnls.skew.incomplete <- numeric(nsamples)
nnls.kurt.incomplete <- numeric(nsamples)

nmf.H.incomplete <- H[known.set,]
nmf.resids.incomplete <- numeric(nsamples)

for(i in 1:nsamples){
  this_sample <- data[,i]

  nnls.result <- my_nnls(W[,known.set],this_sample)
  nmf.result <- fuzzy_nmf(W[,known.set],this_sample)

  nnls.H.incomplete[,i] <- nnls.result$H
  nnls.resids.incomplete[i] <- nnls.result$residual
  nnls.skew.incomplete[i] <- nnls.result$skew
  nnls.kurt.incomplete[i] <- nnls.result$kurt

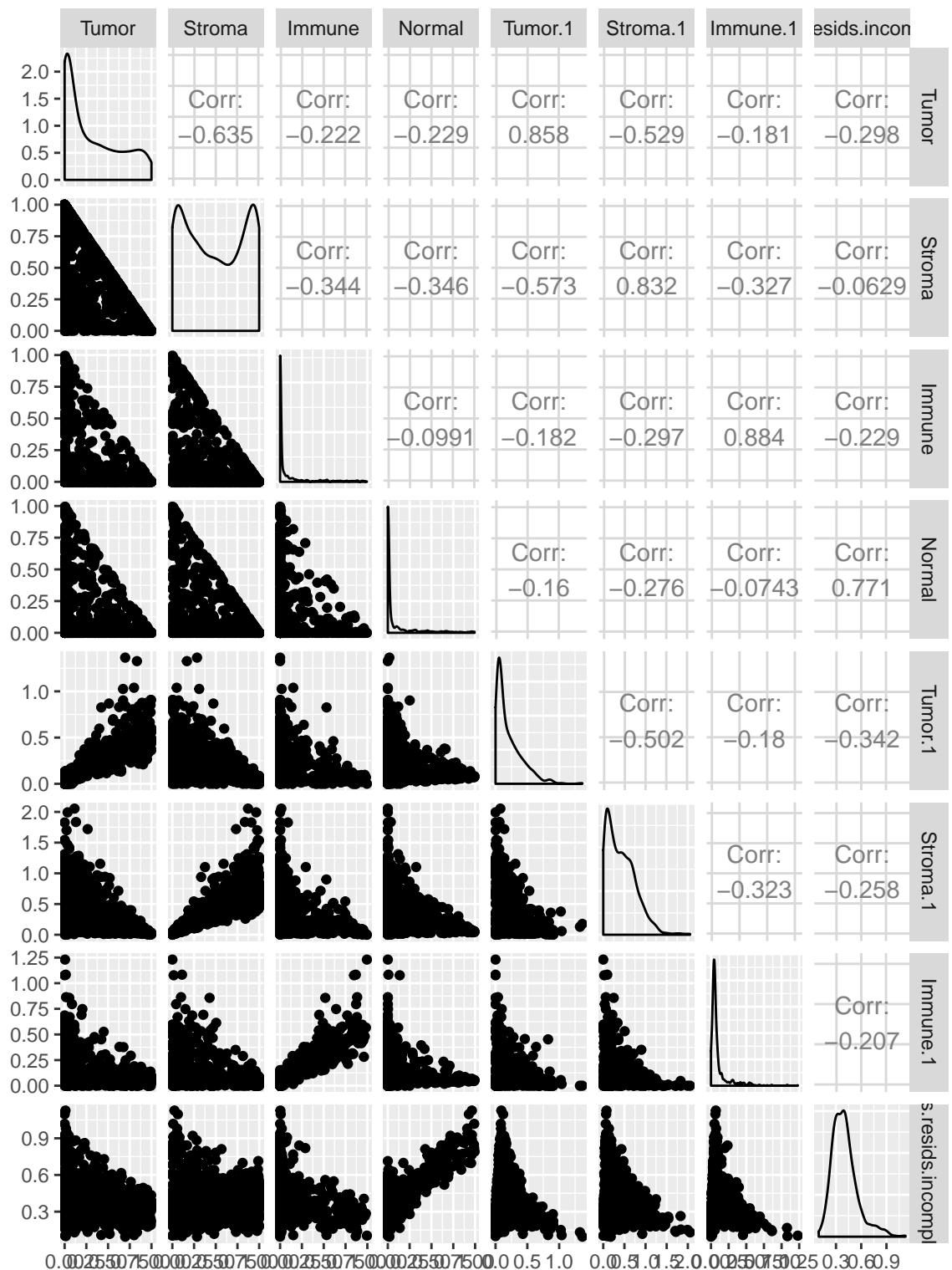
  nmf.H.incomplete[,i] <- nmf.result$H
  nmf.resids.incomplete[i] <- nmf.result$residual
}

}
```

## Compare results

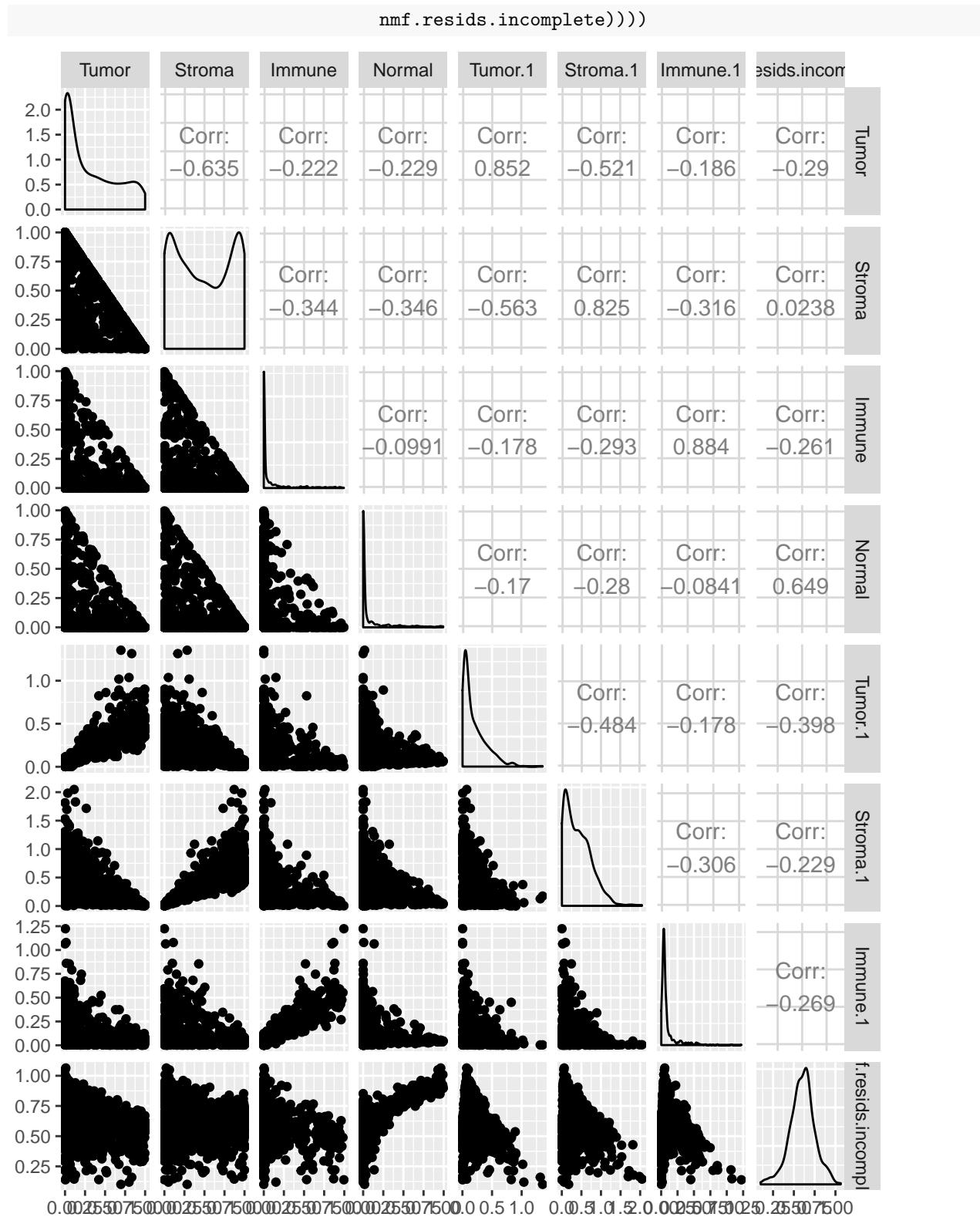
truth versus nnls incomplete

```
ggpairs(data = data.frame(t(rbind(H,
                                    nnls.H.incomplete,
                                    nnls.resids.incomplete))))
```



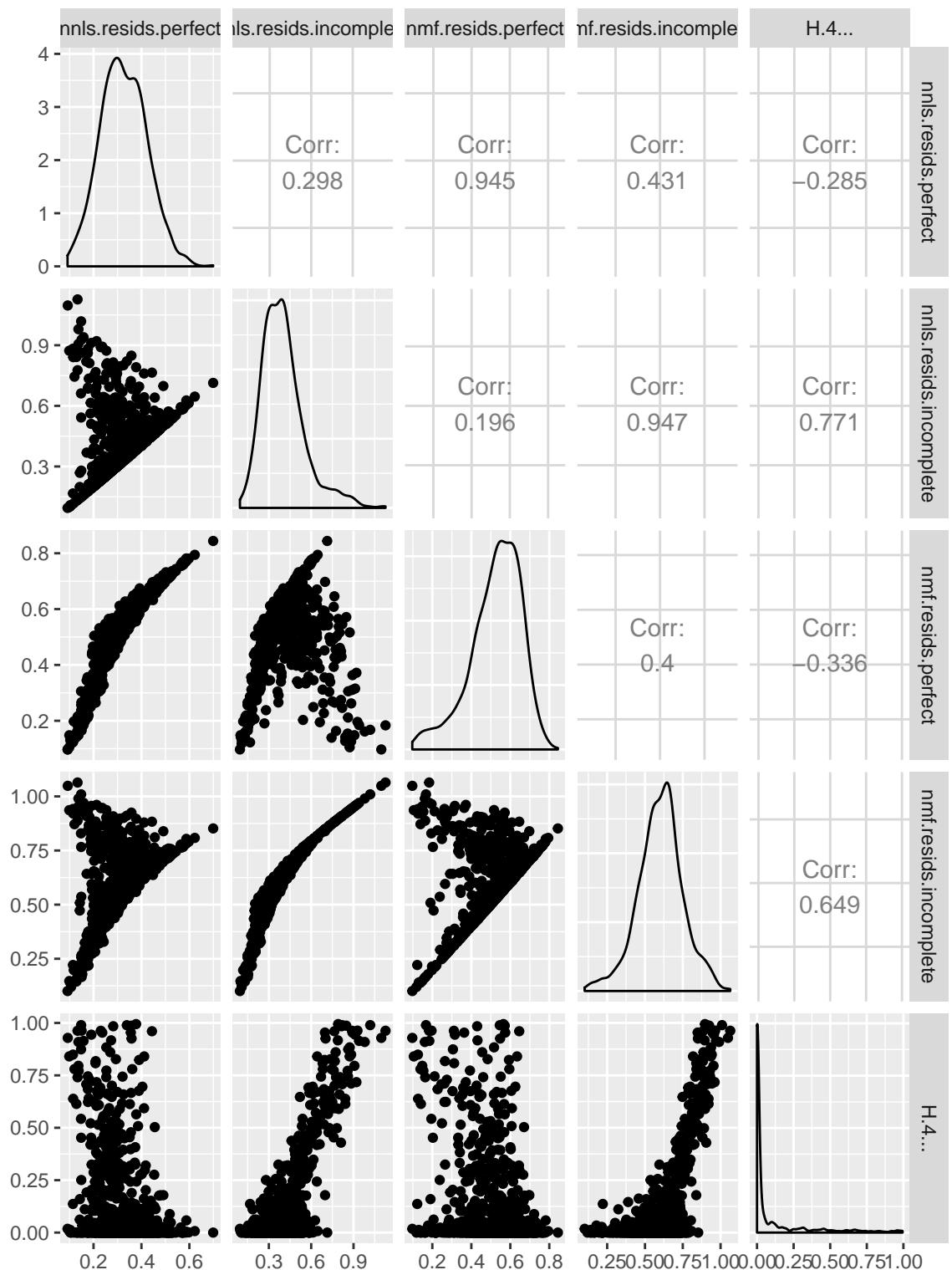
truth versus nmf imperfect

```
ggpairs(data = data.frame(t(rbind(H,
                                    nmf.H.incomplete,
```



## comparison of residuals

```
ggpairs(data = data.frame(nnls.resids.perfect,
                           nnls.resids.incomplete,
                           nmf.resids.perfect,
                           nmf.resids.incomplete,
                           H[4,]))
```



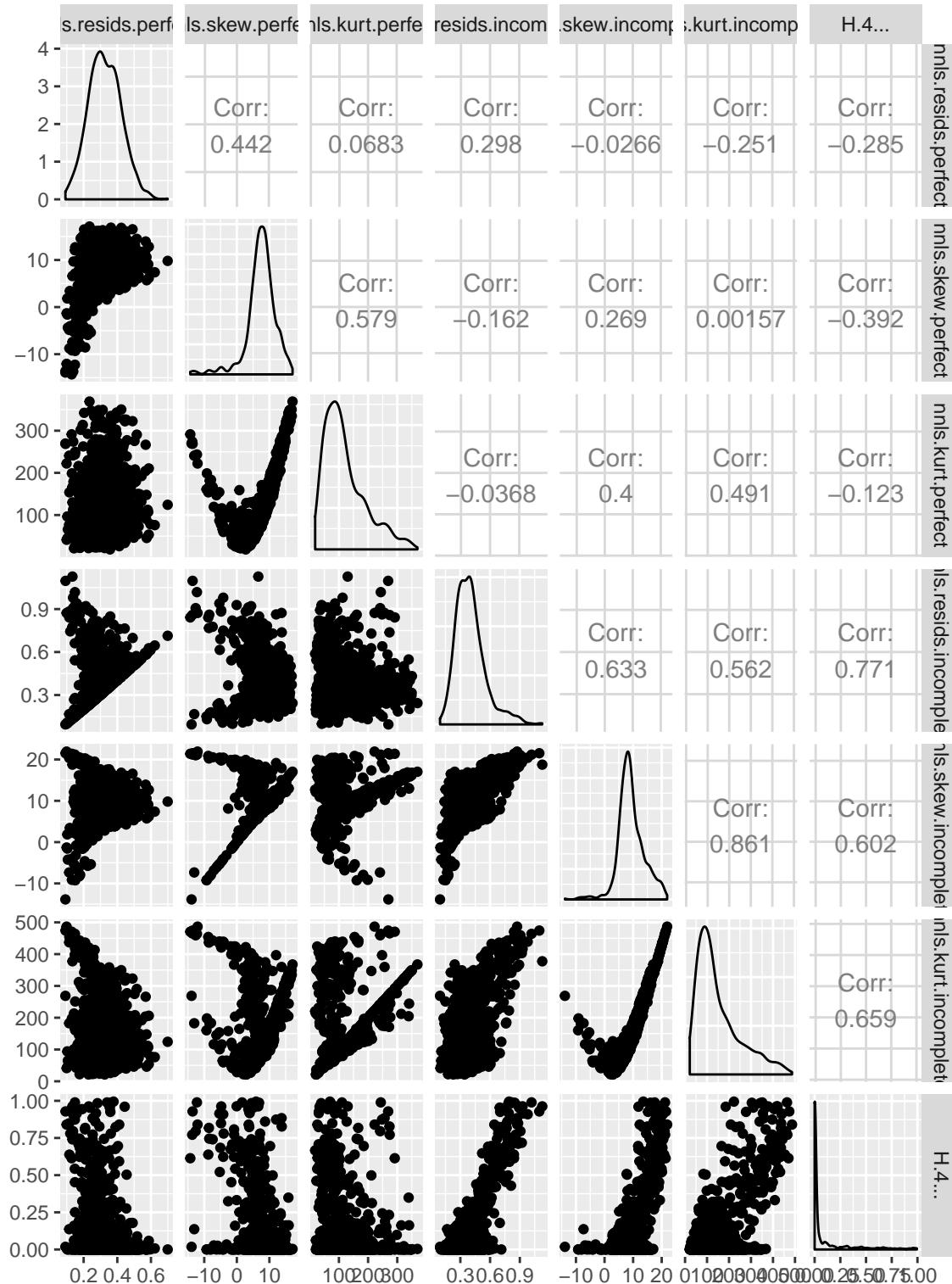
### comparison of gaussianity

```
ggpairs(data = data.frame(nnls.resids.perfect,
                           nnls.skew.perfect,
```

```

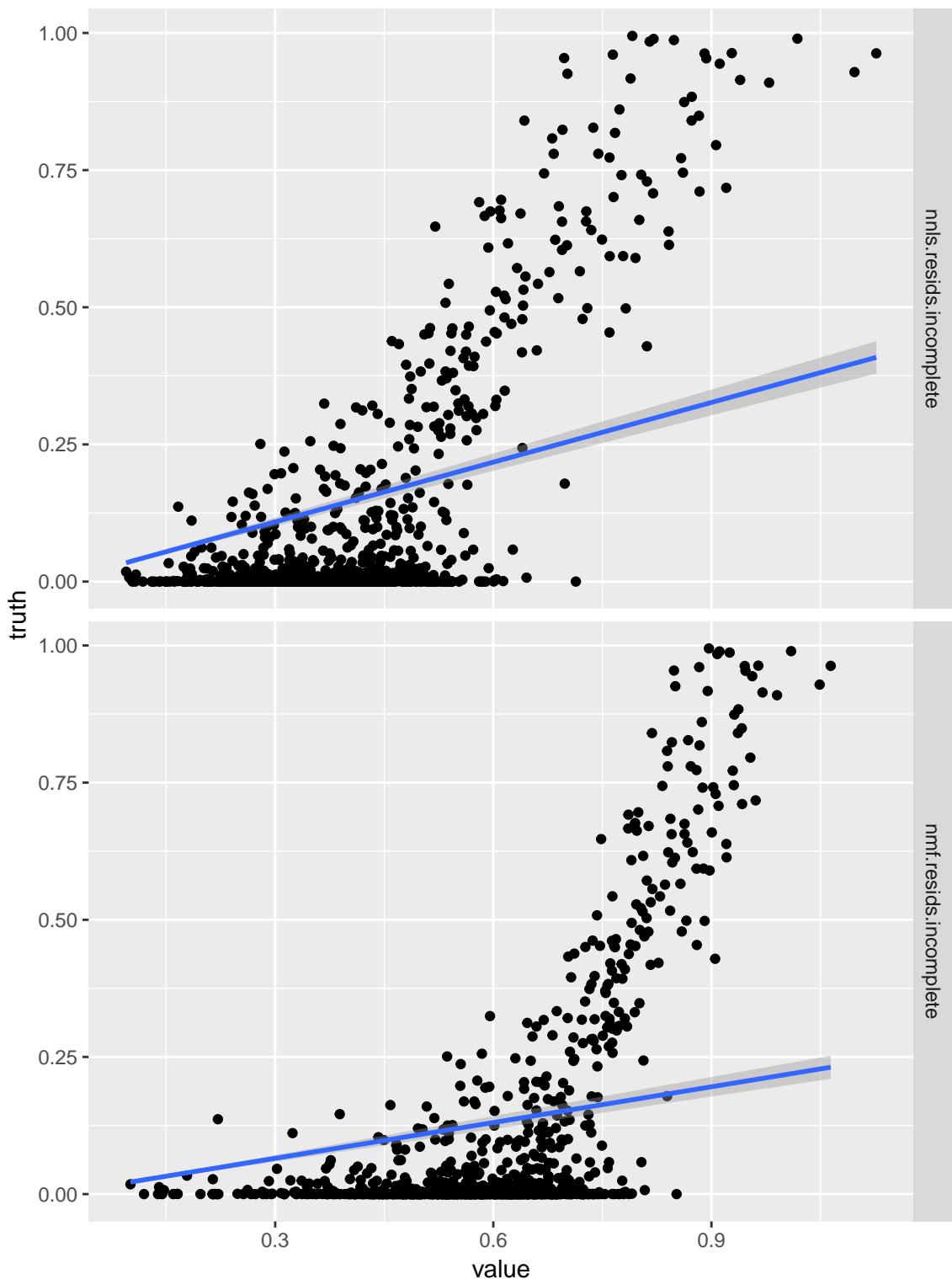
nnls.kurt.perfect,
nnls.resids.incomplete,
nnls.skew.incomplete,
nnls.kurt.incomplete,
H[4,])

```



## Correlation to unknown

```
my_df <- data.frame(nnls.resids.incomplete,
                      nmf.resids.incomplete,
                      truth = H[4,])
my_df <- melt(my_df,id.vars = "truth")
p <- ggplot(data = my_df, aes(y = truth,x = value))
p <- p + facet_grid(facets = variable ~ .)
p <- p + geom_point()
p <- p + geom_smooth(method = "glm",formula = y ~ x -1)
print(p)
```



### predictive power nnls metrics

```
my_df <- data.frame(nnls.resids.perfect,  
                      nnls.resids.incomplete,
```

```

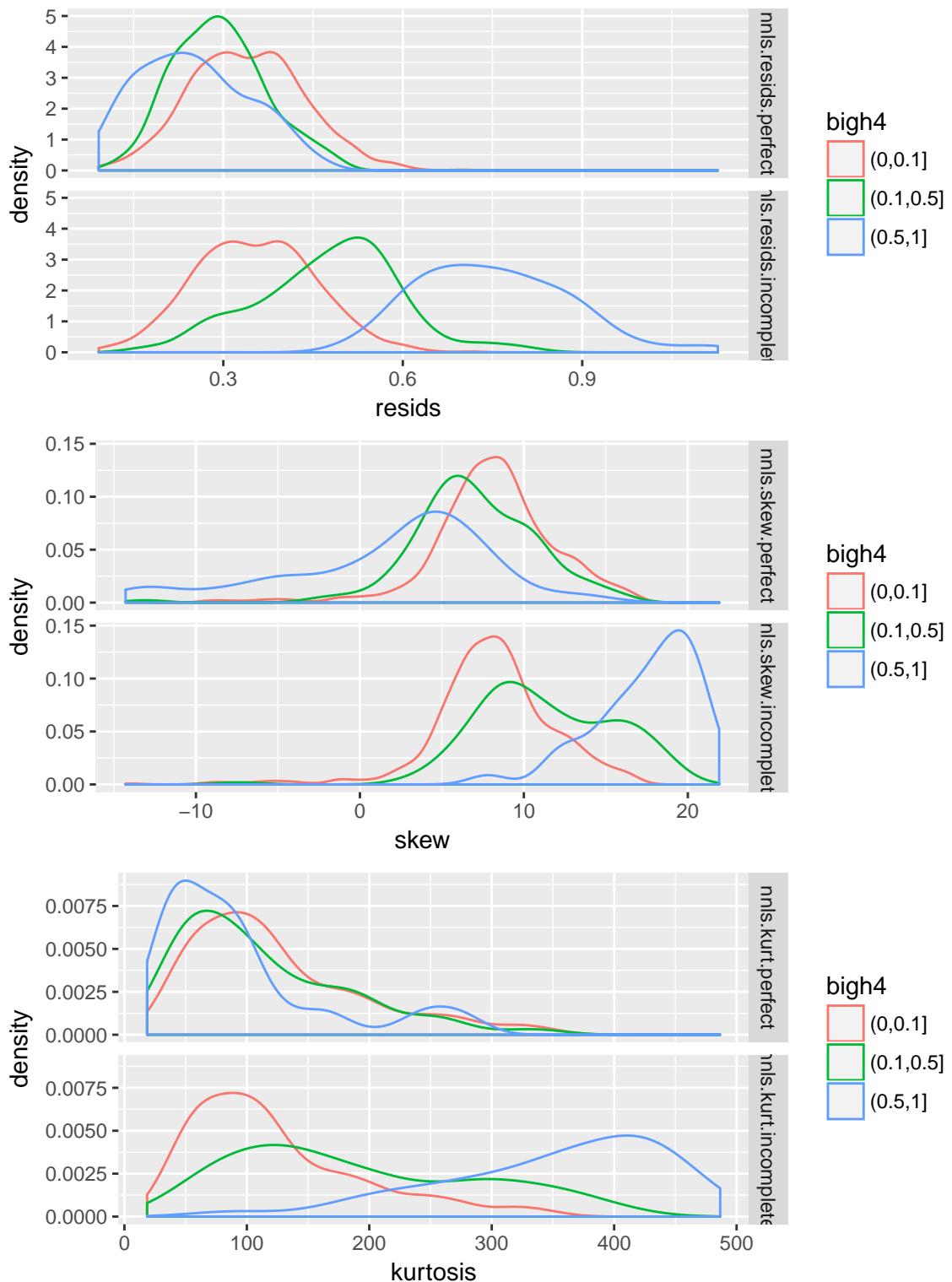
        bigh4 = cut(H[4,],breaks = c(0, 0.1, 0.5 ,1)))
my_df <- melt(my_df,id.vars = "bigh4")
p1 <- ggplot(data = my_df, aes(x = value,color = bigh4))
p1 <- p1 + facet_grid(facets = variable ~ .)
p1 <- p1 + geom_density()
p1 <- p1 + labs(x = "resids")

my_df <- data.frame(nnls.skew.perfect,
                      nnls.skew.incomplete,
                      bigh4 = cut(H[4,],breaks = c(0, 0.1, 0.5 ,1)))
my_df <- melt(my_df,id.vars = "bigh4")
p2 <- ggplot(data = my_df, aes(x = value,color = bigh4))
p2 <- p2 + facet_grid(facets = variable ~ .)
p2 <- p2 + geom_density()
p2 <- p2 + labs(x = "skew")

my_df <- data.frame(nnls.kurt.perfect,
                      nnls.kurt.incomplete,
                      bigh4 = cut(H[4,],breaks = c(0, 0.1, 0.5 ,1)))
my_df <- melt(my_df,id.vars = "bigh4")
p3 <- ggplot(data = my_df, aes(x = value,color = bigh4))
p3 <- p3 + facet_grid(facets = variable ~ .)
p3 <- p3 + geom_density()
p3 <- p3 + labs(x = "kurtosis")

grid.arrange(p1,p2,p3)

```



### predictive power nmf

```
my_df <- data.frame(nnls.resids.perfect,
                      nnls.resids.incomplete,
```

```
nmf.resids.perfect,
nmf.resids.incomplete,
bigh4 = cut(H[4,],breaks = c(0, 0.1, 0.5 ,1)))
my_df <- melt(my_df,id.vars = "bigh4")
p <- ggplot(data = my_df, aes(x = value,color = bigh4))
p <- p + facet_grid(facets = variable ~ .)
p <- p + geom_density()
print(p)
```

