

# autoplot : ready made plots with ggplot2

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Deuxièmes rencontres R

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Jean-Olivier Irisson



# The problem

```
head(env, 1)
obj <- prcomp(env, scale.=T)

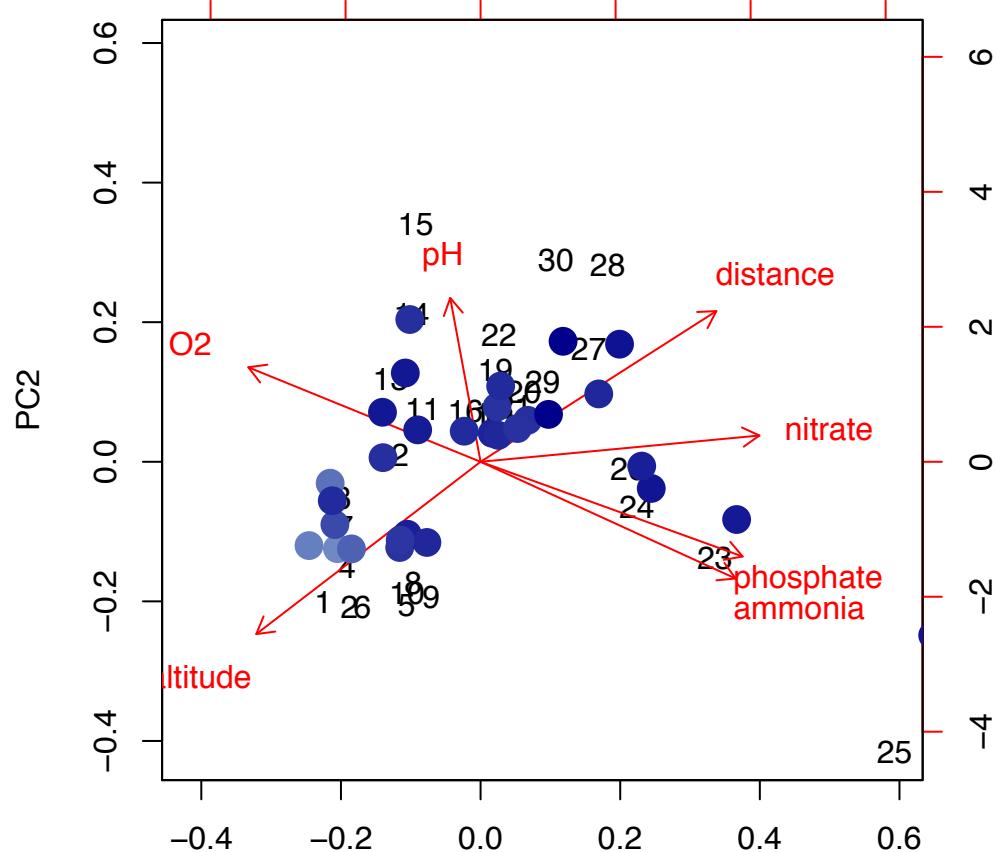
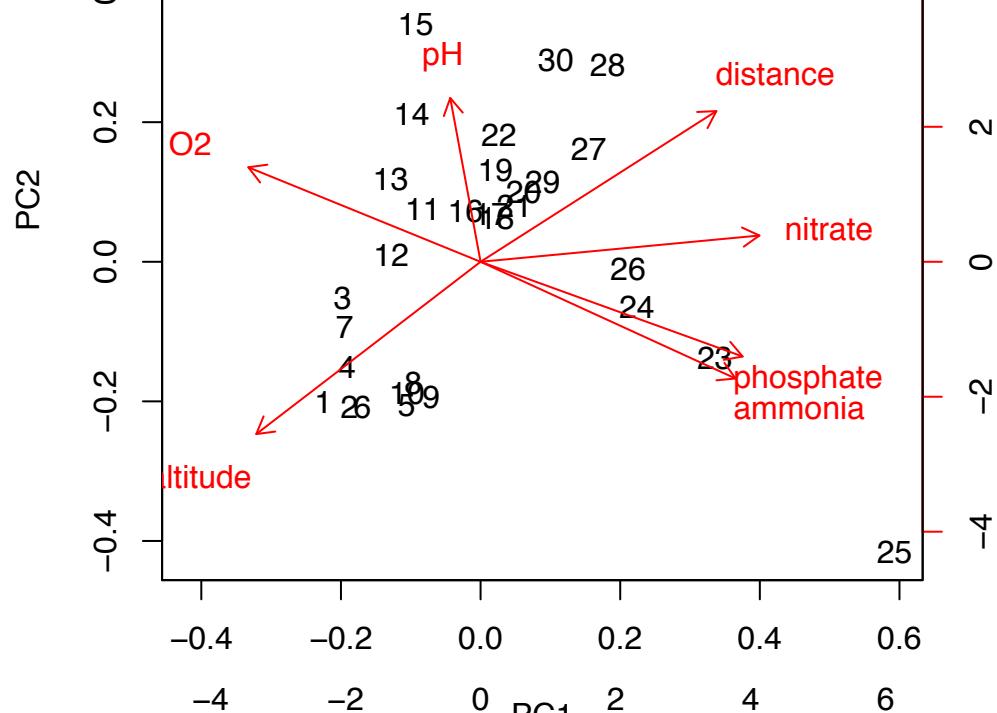
class(obj)
[1] "prcomp"

summary(obj)
biplot(obj)

blues <- colorRamp(
  c("lightblue", "darkblue"))

cols <- rgb(blues(allEnv$hard /
  max(allEnv$hard)), max=255)

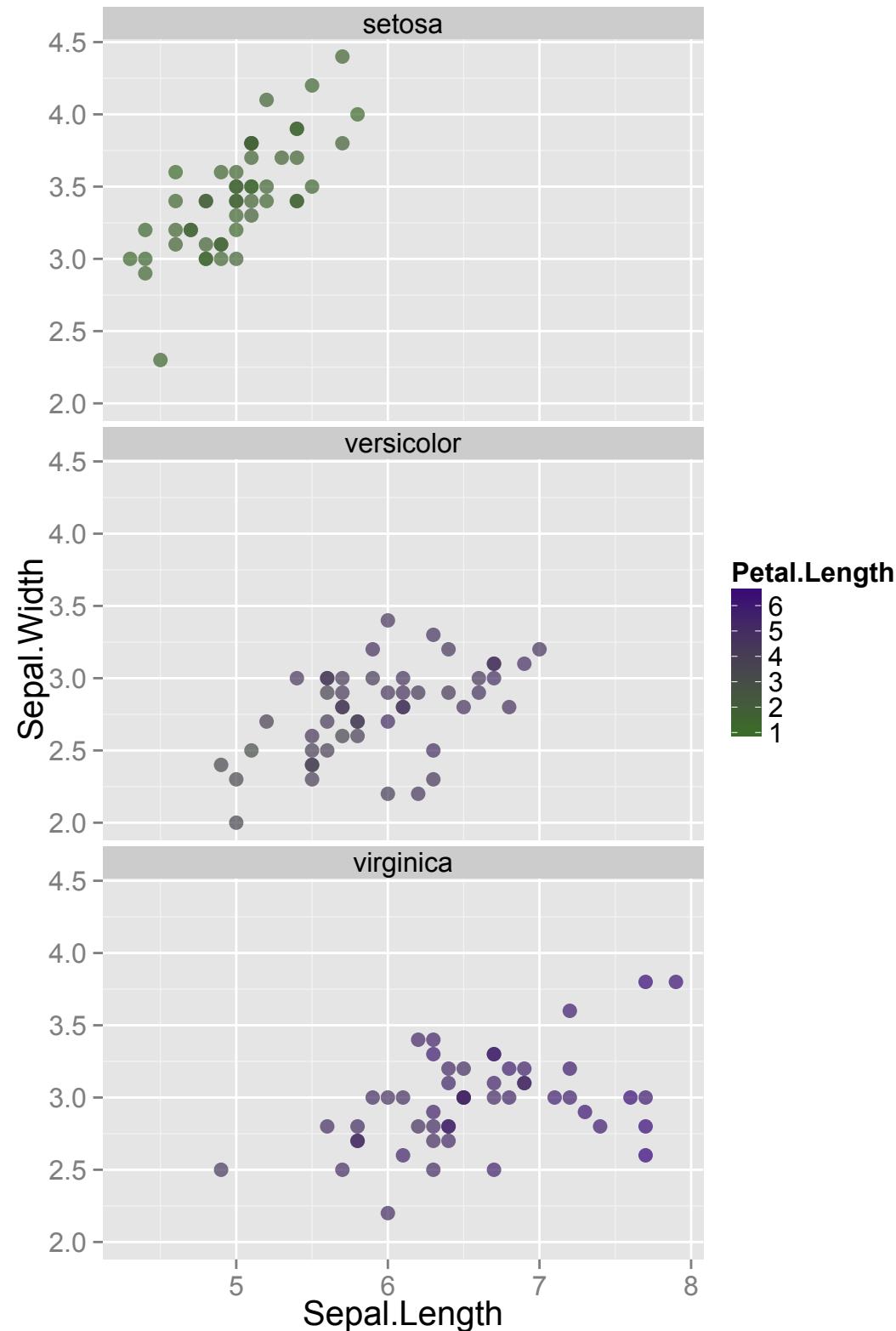
points(obj $x[,1], obj $x[,2],
  col=cols, pch=16, cex=2)
```



# Why ggplot?

- Visually pleasing
- Automatic everything
- Customizable after the fact

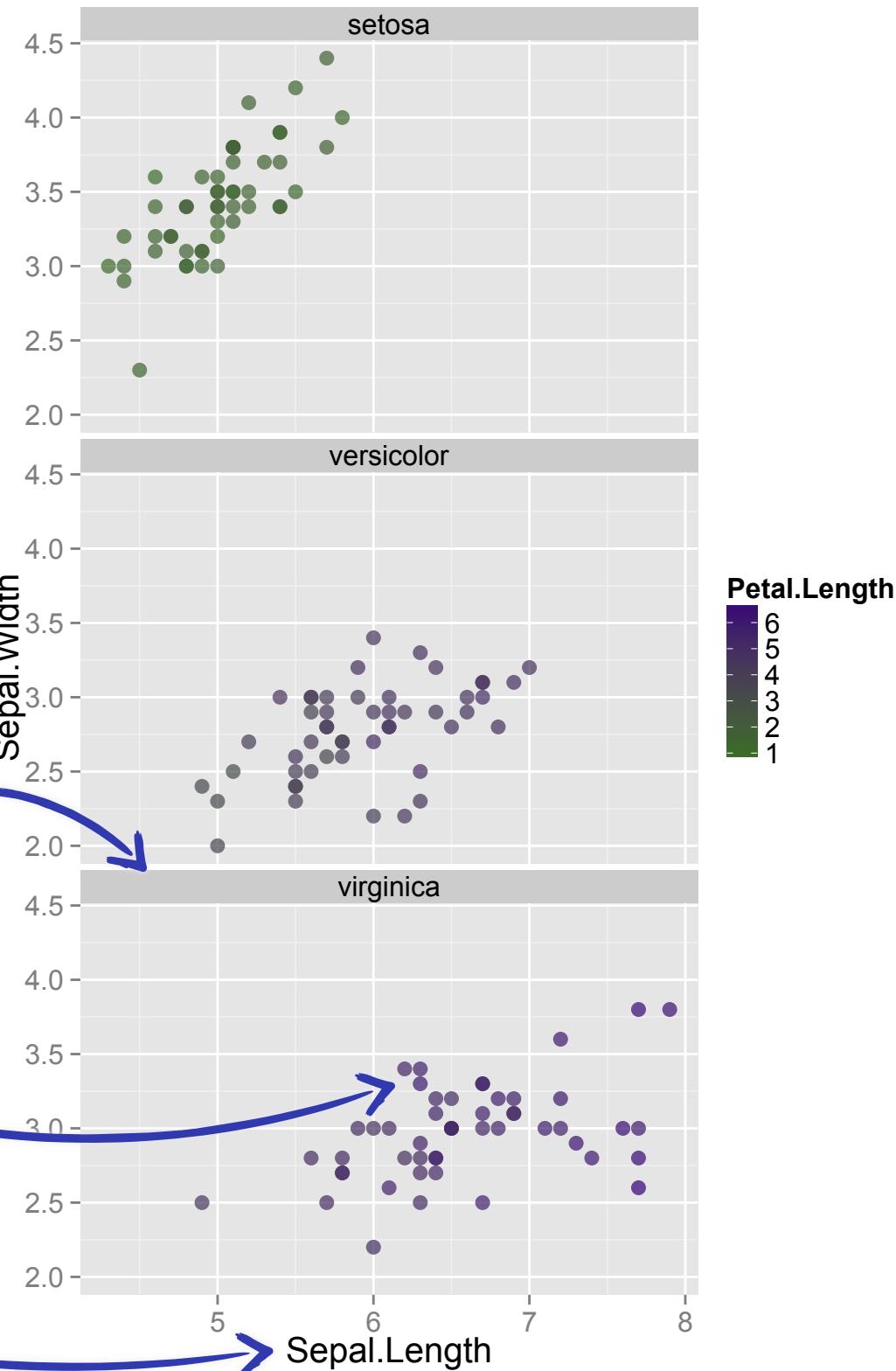
```
p <- ggplot(iris) +  
  geom_point(aes(  
    x=Sepal.Length,  
    y=Sepal.Width,  
    colour=Petal.Length),  
    alpha=0.7, size=3)  
  
p +  
  facet_wrap(~Species, ncol=1)+  
  scale_colour_continuous(  
    low="#3C6B2A", high="#350079")
```



# How does ggplot2 work?

- grammar of graphics
- mapping data  $\leftrightarrow$  aesthetics

Sepal. length	Sepal. width	Petal. length	Peta. width	Species
6.0	3.0	4.8	1.8	virginica
6.4	2.9	4.3	1.3	versicol
4.6	3.6	1.0	0.2	setosa
4.3	3.0	1.1	0.1	setosa



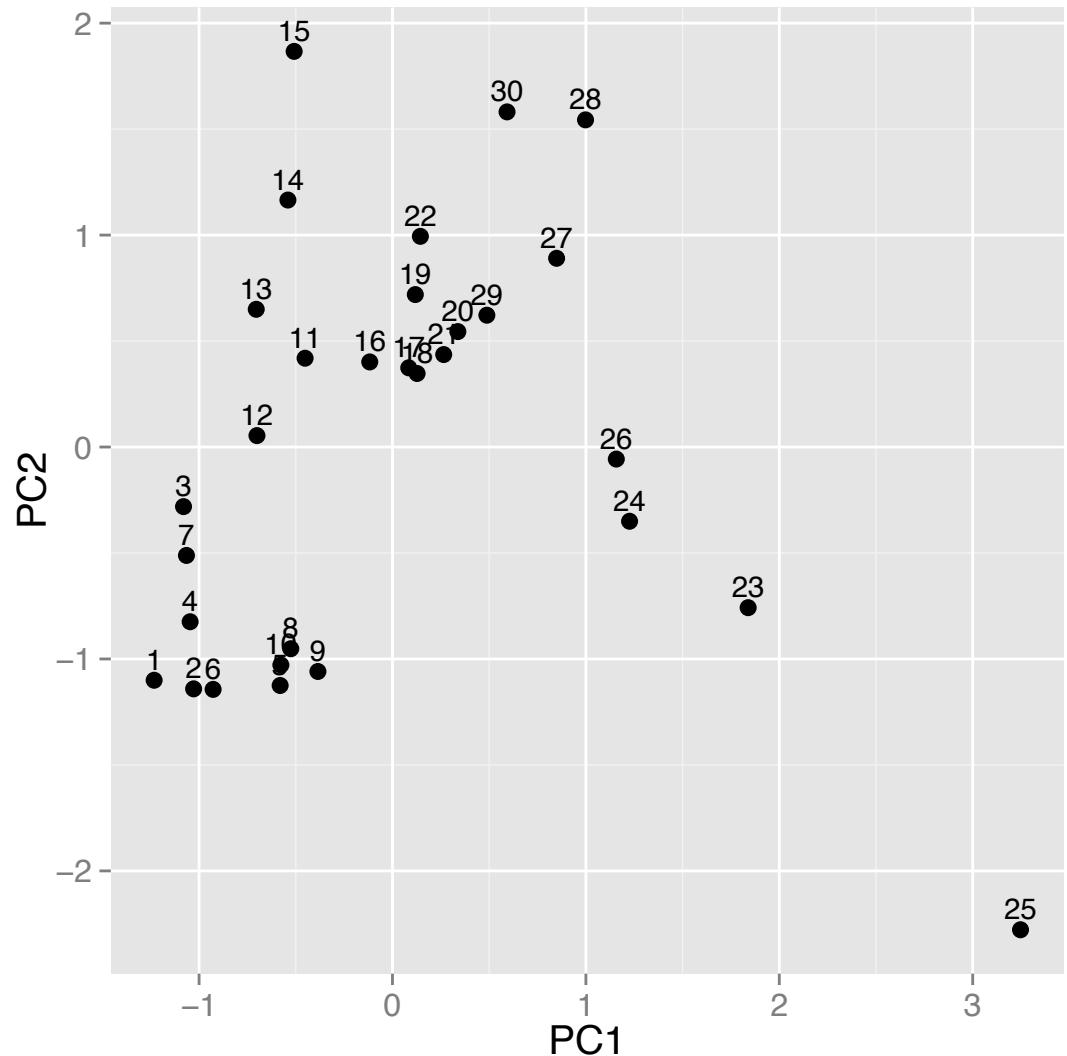
# Use ggplot2 for multivariate analyses output

```
obj <- prcomp(env, scale.=T)

# prepare a data.frame
scores <- obj$x[,1:2]
lambda <- obj$sdev[1:2]
d <- data.frame(id=1:nrow(d),
  t(t(scores)/lambda))

# plot it
ggplot(d, aes(x=PC1, y=PC2)) +
  geom_point() +
  geom_text(aes(label=id),
    size=3, vjust=-0.5)

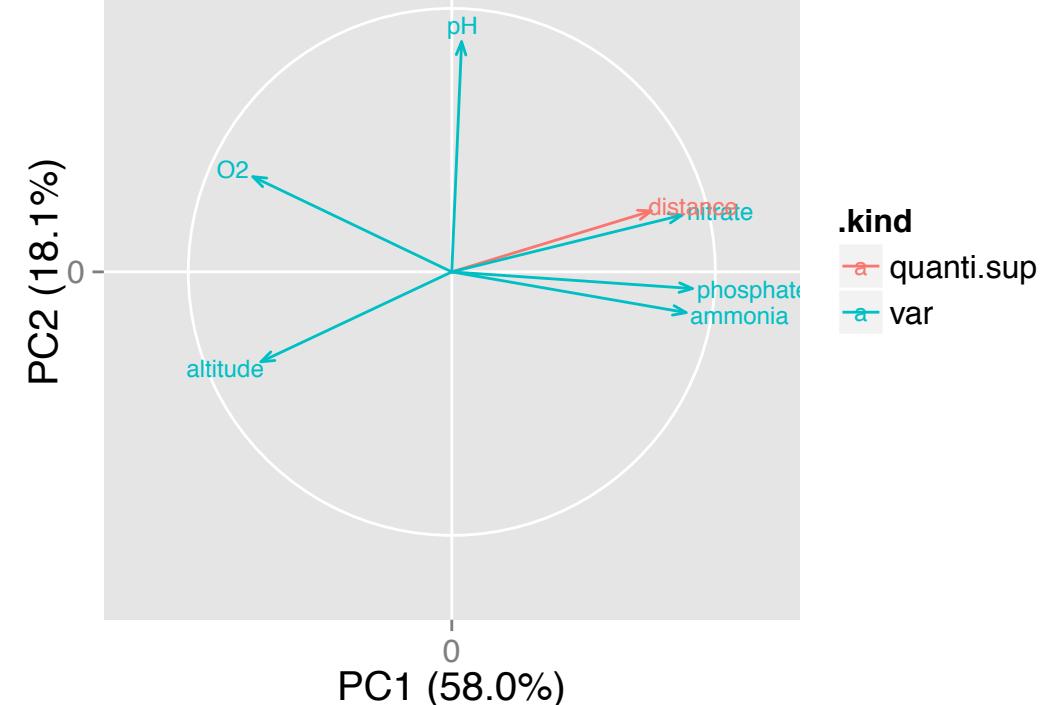
# or
autoplot(obj, type="obs")
```



# Automatic

- computation of percentage of variance explained
- better scales
- automatic colours

```
library("FactoMineR")  
  
obj <- PCA(env, graph=F,  
          quanti.sup=1, ind.sup=1)  
  
autoplot(obj, type="var")  
  
autoplot(obj, type="obs")
```



# Yet customizable

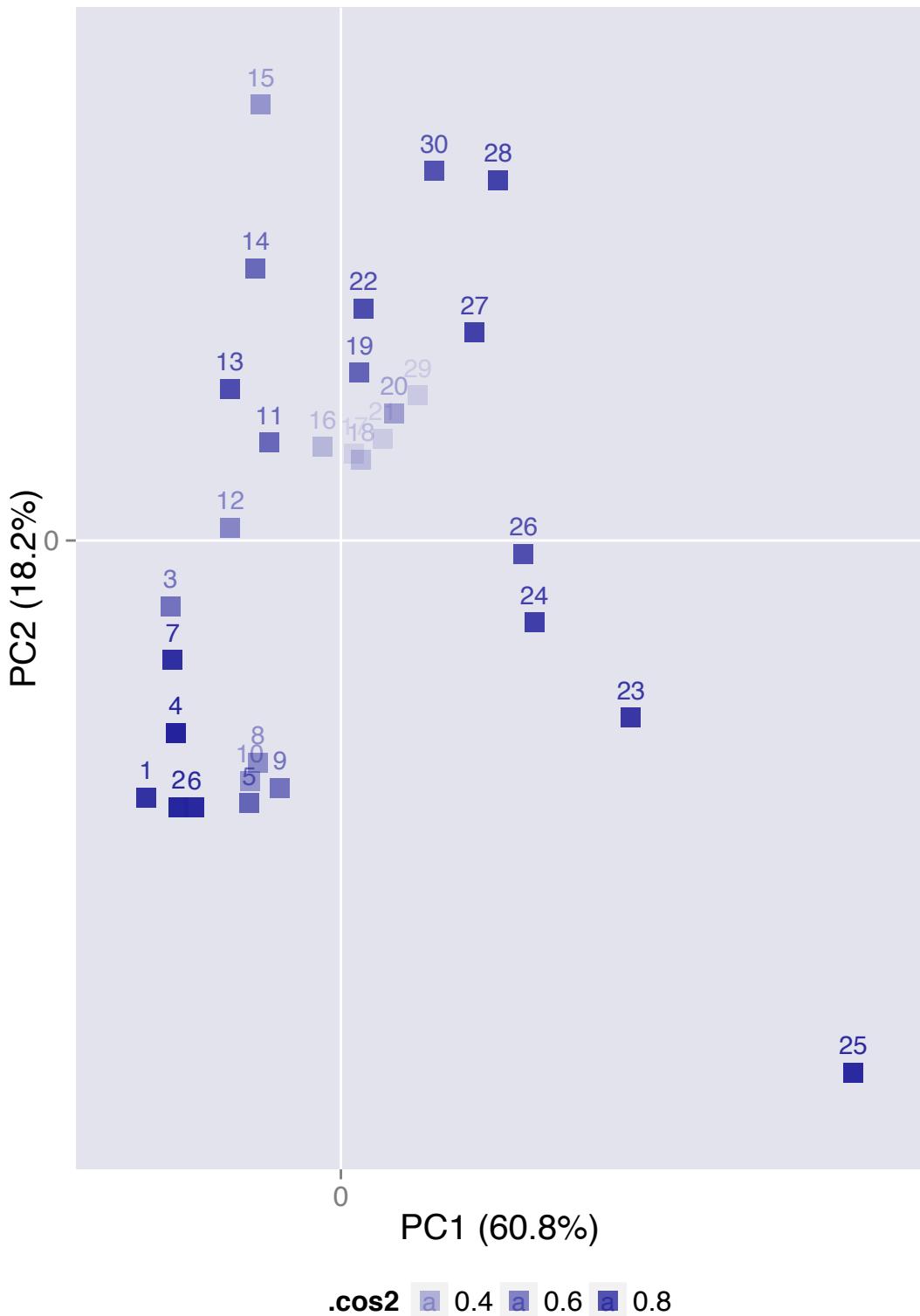
```
obj <- PCA(env, graph=F)

p <- autoplot(obj, type="obs",

  # map additional variables
  mapping=aes(alpha=.cos2),

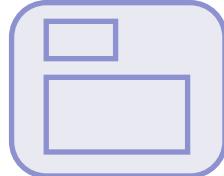
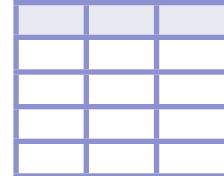
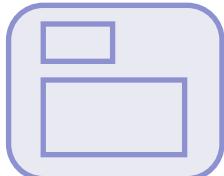
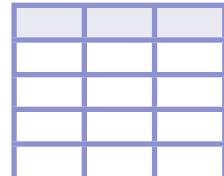
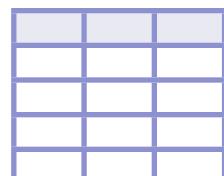
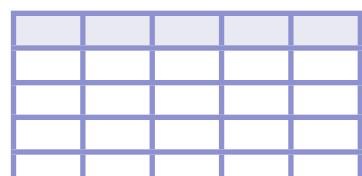
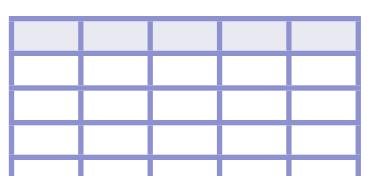
  # pass settings to ggplot calls
  shape=15, colour="#24239D"
)

# change the look after the fact
p + theme(
  panel.background=
  element_rect(fill="#E3E3EE"),
  legend.position="bottom"
)
# yay, beamer-like ggplot!
```



## Two step process

---

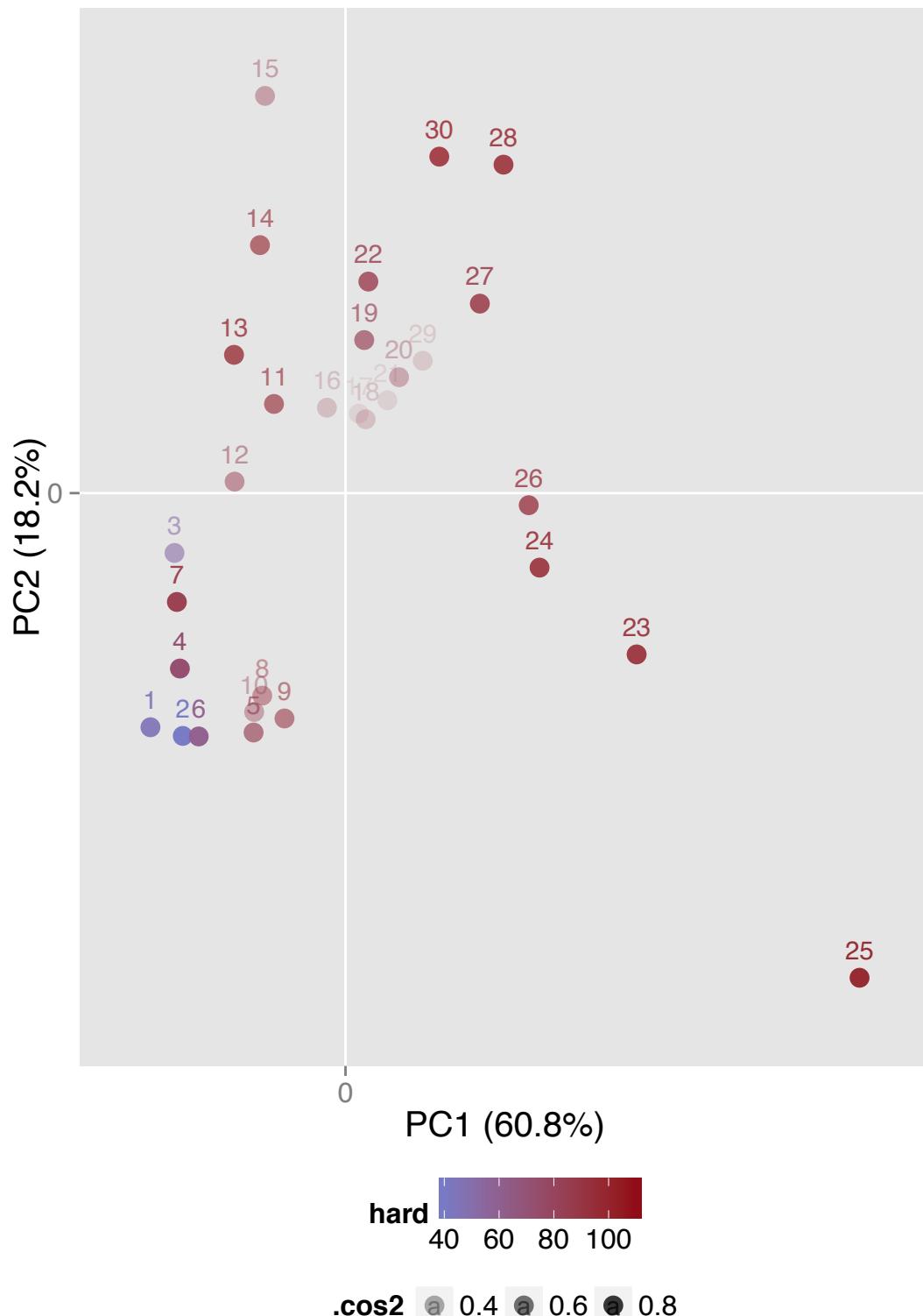
```
autoplot(model=  , data=  , ...) {  
  fortify(model=  , data=  ) {  
    cbind(  ,  )  
    return(  )  
  }  
  ggplot(  , ...) + geom_???(aes(...))  
}
```

# Map original variables, outside the analysis

```
head(env, 1)
  distance altitude pH phosphate
1      3        934 79
  nitrate ammonia O2
1     20          0 122
```

```
head(allEnv, 1)
  distance altitude pH phosphate
1      3        934 79
  nitrate ammonia O2 slope flow
1     20          0 122 6.176 84
hard O2_demand
1    45          27
```

```
autoplot(obj, data=allEnv,
  type="obs",
  mapping=aes(alpha=.cos2,
  colour=hard)
)
```

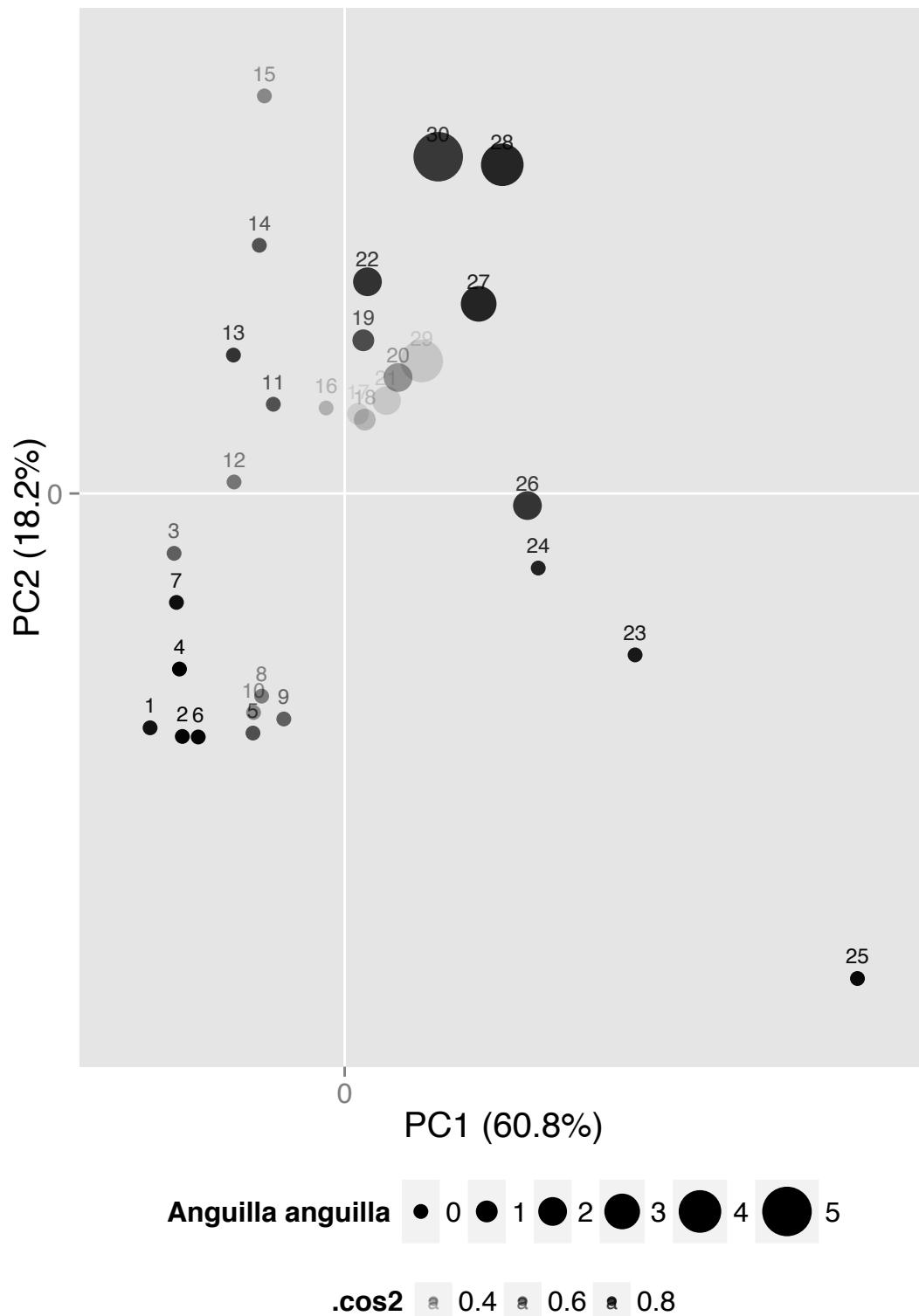
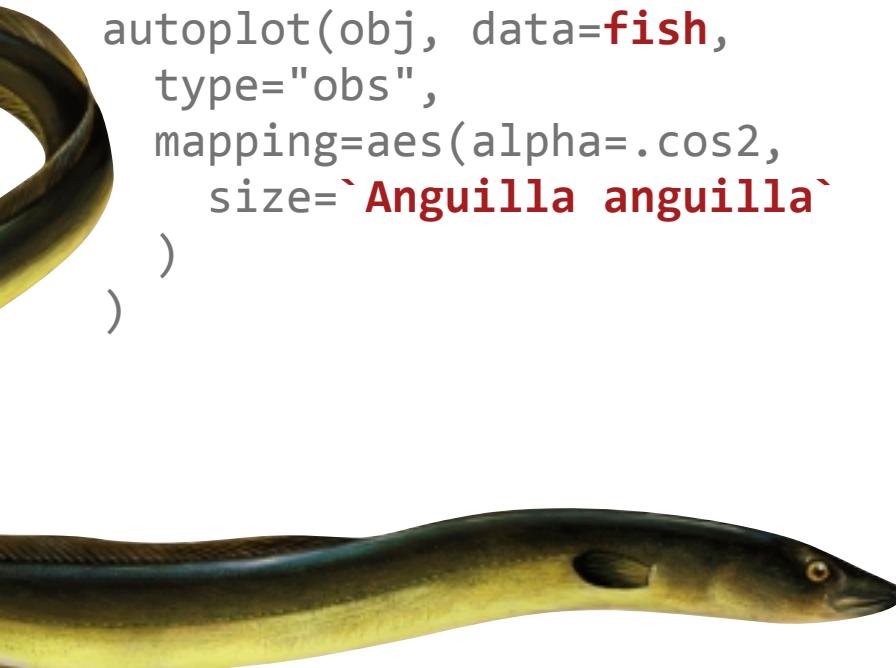


# Map variables from other datasets

```
dim(env)  
[1] 30 7
```

```
dim(fish)  
[1] 30 27
```

```
autoplot(obj, data=fish,  
        type="obs",  
        mapping=aes(alpha=.cos2,  
                    size=`Anguilla anguilla`  
        ))
```

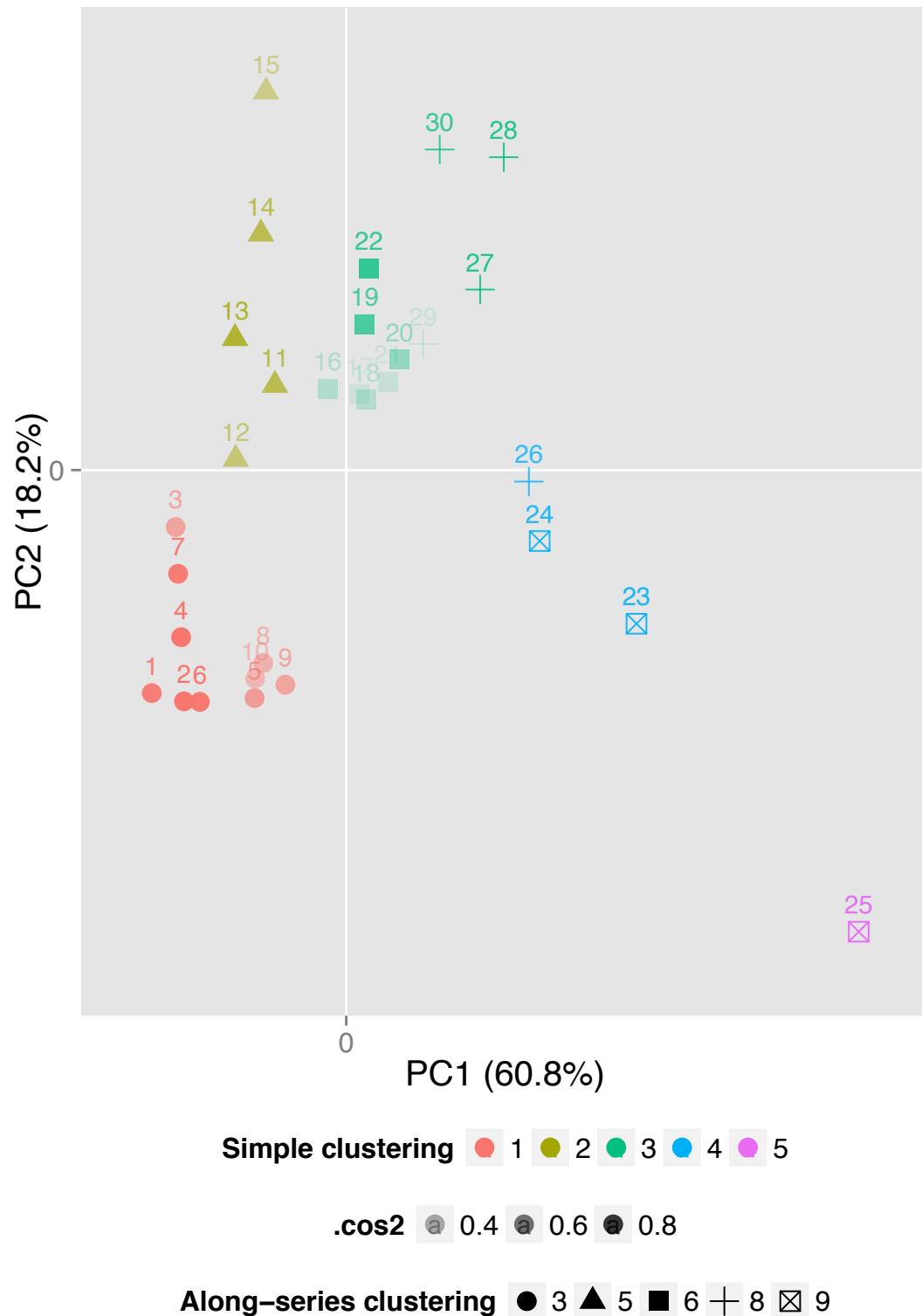


# Easily visualize several results

```
scores <- as.matrix(fortify(...))

# un-constrained clustering
clust <- hclust(
  dist(scores), method="ward")
env$clust <- cutree(clust, 5)
# clustering based on distance
library("mvpart")
dClust <- mvpart(
  scores ~ distance,
  data=env, size=5)
env$dclust <- dClust$where

autoplot(obj, data=env,
  type="obs",
  mapping=aes(alpha=.cos2,
    colour=clust,
    shape=dclust))
)
```



# Homogenization at little cost

- Several fortify methods
- One plotting function

```
obj <- stats::prcomp(env,  
  scale.=TRUE)
```

```
autoplot(obj, type="obs")
```

```
obj <- FactoMineR::PCA(env,  
  graph=FALSE)
```

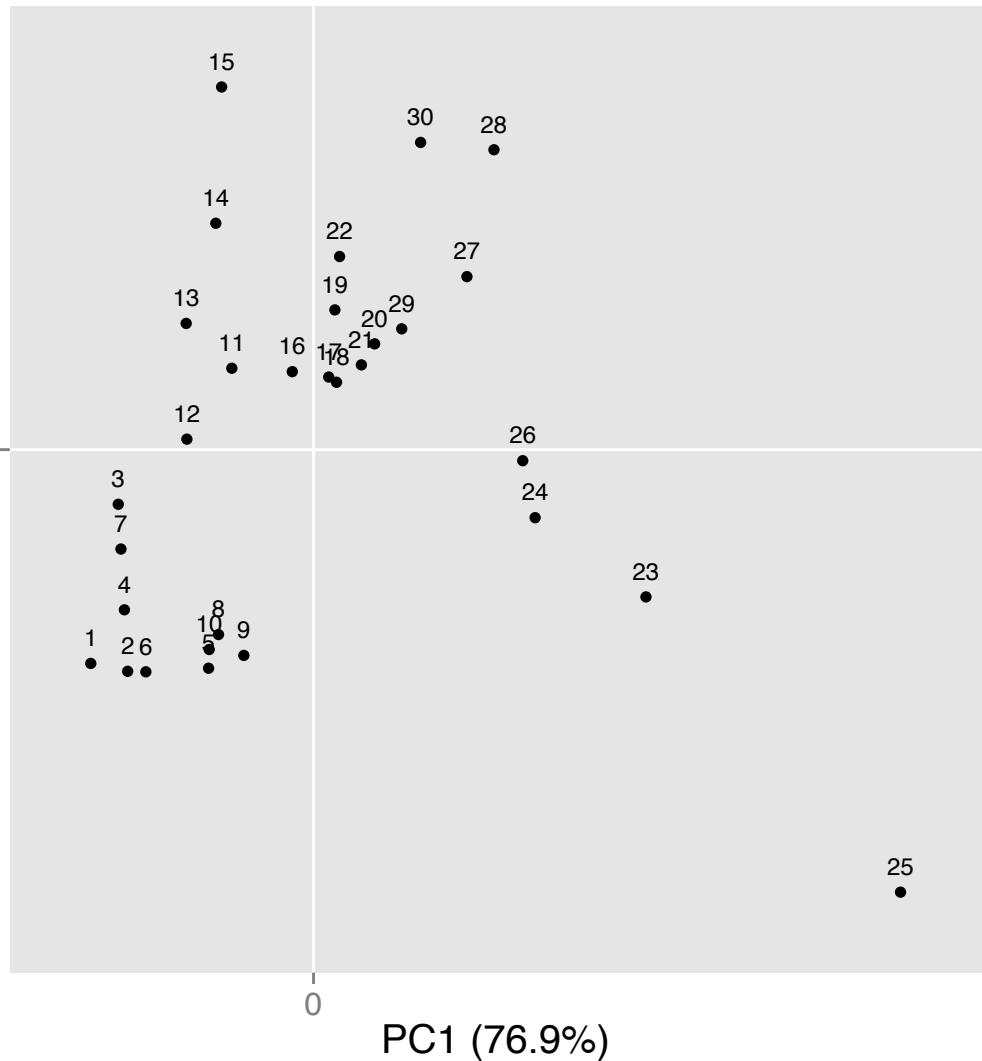
```
autoplot(obj, type="obs")
```

```
obj <- pcaMethods::pca(env,  
  scale="uv")
```

```
autoplot(obj, type="obs")
```

- In progress: fortify.pca (ade4)  
and fortify.rda (vegan)

PC2 (23.1%)



# Future directions

- Scaling + biplot in PCA
- More multivariate functions (CA, MCA are in progress)
- Code autoplot.lm() (fortify.lm() is already in ggplot2)
- Quantile regression (rq, rqs)
- Bayesian stats (mcmc)
- <https://github.com/jiho/autoplot>

```
library("devtools")
install_github("autoplot","jiho")
```

The screenshot shows a GitHub repository page for 'jiho/autoplot'. The repository has 50 commits, 1 branch, and 0 tags. The master branch is selected. A yellow sticky note in the bottom right corner says 'Thanks Hadley!'. The page includes a list of files and their descriptions:

- R: Make score extraction work when only one element is extracted
- inst: Join data and pca results, rather than rbind them
- man: Improve package-wide documentation
- tests: Fix tests which need to load autoplot and not ggplot2
- .gitignore: Ignore the scratch folder in git
- DESCRIPTION: Fix extraneous comma
- NAMESPACE: Add methods for ...