

# Metapopulation Genetics

Helene Wagner, University of Toronto

## Goal: Test genetic effects of metapopulation dynamics

- Quantify spatial genetic structure
- Test effects of population size and connectivity
- Test effect of recent extinction events

*D. depressissimum*



Photos: J-P Pointier

## Methodological Challenges

1. 'EcoGenetics': multi-source data
2. AMOVA with R
3. Regression analysis in R:
  - Interpret results
  - Check assumptions
4. Video 2: graphics with 'ggplot2'



# Freshwater Snail Data

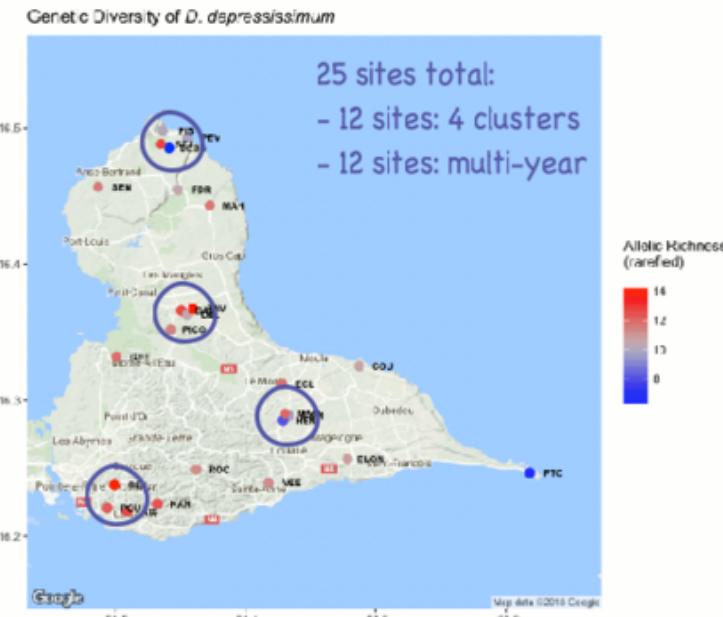
Lamy et al. (2012), Molecular Ecology 21: 1394 – 1410

## Genetic data:

- 22-32 individuals per site
  - 10 microsatellites

## Population-level:

- RA: allelic richness
  - He: expected heterozygosity
  - FST.GESTE: site-level FST
  - Fst.temp: temporal FST



## Hierarchical data set

$$4 \text{ clusters} \times 3 \text{ ponds} = 12$$

## AMOVA:

## Differentiation within vs. between clusters

## Within < Between

## Spatial data set

25 sites (out of 2059 ponds)

## Multiple regression:

$$RA \sim NLT + C$$

## FST.GESTE ~ NLT + C

Diversity increases with NLT  
Differentiation decreases with C

### Predictor variables:

- NLT: long-term pop size
  - Size: pond size
  - C: hydrological connectivity
  - D: wetland area within 2 km
  - APE: apparent extinction event?

## Temporal data set

12 ponds x 2-4 years  
(5 with APE, 7 without)

## Two-sample t-test:

Fst.temp ~ APE

Higher Fst.temp if extinct

# Package 'EcoGenetics'

## || ECOGEN CLASS OBJECT ||

ecoslot.G (obj)  
obj[[ "G" ]]  
Not: obj @ G

Access to slots: <ecoslot.> + <name of the slot> + <(name of the object)>  
See: `help("EcoGenetics accessors")`

Spatial coordinates	slot XY:	--> 1270 x 2	coordinates
Phenotypic traits	slot P:	--> 0 x 0	phenotypic variables
Genetic data	slot G:	--> 1270 x 10	loci >> ploidy: 2    codominant
Allele frequencies	slot A:	--> 1270 x 372	alleles
Site variables	slot E:	--> 0 x 0	environmental variables
Strata (sampling)	slot S:	--> 1270 x 4	structures >> 4 structures found
Anything else?	slot C:	--> 0 x 0	variables
Result output	slot OUT:	--> 0	results

SiteID, SITE  
Cluster, YEAR

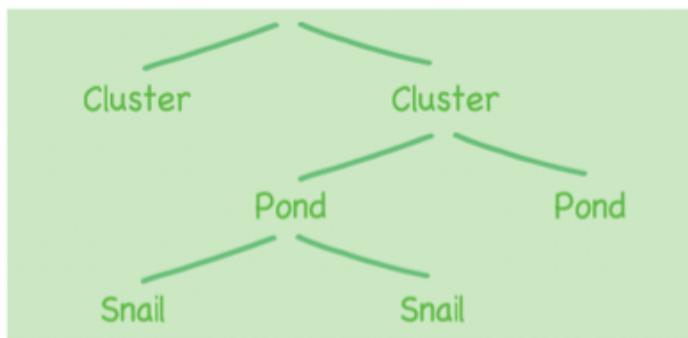
### Why use 'EcoGenetics'?

1. Integrate multi-source data
2. Export / import, e.g.: 'ecogen2genind'
3. Calculate spatial statistics (Week 5)

### Data Management:

- Slicing: `obj[1:3, ]`
- Constructing 'ecogen': row names must match
- Can join by column: 'eco.fill\_ecogen\_with\_df'
- Aggregate to 'ecopop' object: 'ecogen2ecopop'

## Hierarchical Levels



## Why?

FST:

- Genetic structure relative to fixation
- Parameters rely on pop gen assumptions

AMOVA:

- No underlying pop gen model

## Interpretation

1. Variance explained at each level?
2. Significantly > 0?
3. Phi = population differentiation statistics

\$statphi

	Phi	
Phi-samples-total	0.094235649	PhiST = a + b
Phi-samples-Cluster	0.089553260	PhiSC = b/(b+c)
Phi-Cluster-total	0.005142958	PhiCT = a

# AMOVA

A-nalysis of  
MO-lecular  
VAriance

```
> amova.result
```

```
$call
```

```
ade4::amova(samples = xtab, distances = xdist, structures = xstruct)
```

```
$results
```

	Df	Sum Sq	Mean Sq
Between Cluster	3	106.2413	35.413781
Between samples Within Cluster	8	249.6422	31.205270
Within samples	353	2764.2124	7.830630
Total	364	3120.0959	8.571692

```
$components of covariance
```

	Sigma	%
Variations Between Cluster	0.04446256	0.5142958
Variations Between samples Within Cluster	0.77023554	8.9092691
Variations Within samples	7.83062999	90.5764351
Total variations	8.64532809	100.0000000

```
> amova.test
```

```
class: krandtest lightkrandtest
```

```
Monte-Carlo tests
```

```
Call: randtest.amova(xtest = amova.result, nrepet = 999)
```

```
Number of tests: 3
```

```
Adjustment method for multiple comparisons: none
```

```
Permutation number: 999
```

	Test	Obs	Std.Obs	Alter	Pvalue
1	Variations within samples	7.83062999	-80.175606	less	0.001
2	Variations between samples	0.77023554	57.660752	greater	0.001
3	Variations between Cluster	0.04446256	1.781495	greater	0.050

# Regression Interpretation

## Cohen's Effect Size

	Correlation	Regression
S	$r > 0.1$	$R^2 > 0.01$
M	$r > 0.3$	$R^2 > 0.09$
L	$r > 0.5$	$R^2 > 0.25$

lm ( FST.GESTE ~ NLT + C, data = dd.site )

Direction?

Sign of slope

As expected?

Effect Size?

Multiple R-squared

Beta coefficients

Biologically relevant?

Significant?

p-values of t-tests

Rule out chance?

Correlation is not causation!

## Variation Partitioning



## With Original Variables

### Coefficients: Slopes

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	0.197237	0.034235	5.761	8.53e-06	***
NLT	-0.042685	0.012774	-3.342	0.00296	**
C	-0.030047	0.009946	-3.021	0.00628	**

Signif. codes: 0 ‘\*\*\*’ 0.001 ‘\*\*’ 0.01 ‘\*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 0.04187 on 22 degrees of freedom

Multiple R-squared: 0.4599, Adjusted R-squared: 0.4108

F-statistic: 9.367 on 2 and 22 DF, p-value: 0.00114 **Adj R2**

## All Variables Standardized

### Coefficients: Beta coeffs

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-2.576e-17	1.535e-01	0.000	1.00000	
scale(NLT)	-5.254e-01	1.572e-01	-3.342	0.00296	**
scale(C)	-4.750e-01	1.572e-01	-3.021	0.00628	**

Signif. codes: 0 ‘\*\*\*’ 0.001 ‘\*\*’ 0.01 ‘\*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 0.7676 on 22 degrees of freedom

Multiple R-squared: 0.4599, Adjusted R-squared: 0.4108

F-statistic: 9.367 on 2 and 22 DF, p-value: 0.00114

# Checking Assumptions

## OLS Assumptions

### Sample:

- Simple random sample
- Randomized experiment

### Response Y:

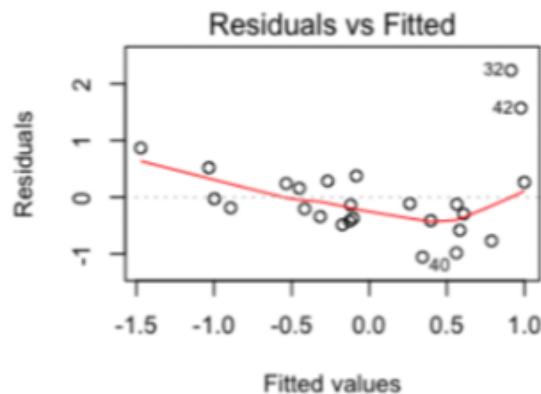
- Linear relationship with X
- Errors normally distributed
- Errors with constant variance
- Errors identically distributed
- Errors independent

### Predictors X:

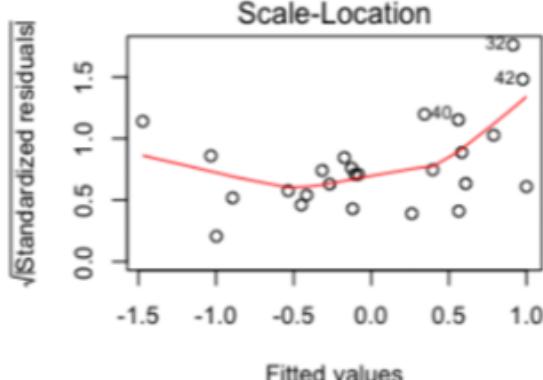
- Measured without error
- No multicollinearity
  - strict: no redundant vars
  - wide: low cor among X

## Variance Inflation Factor (VIF)

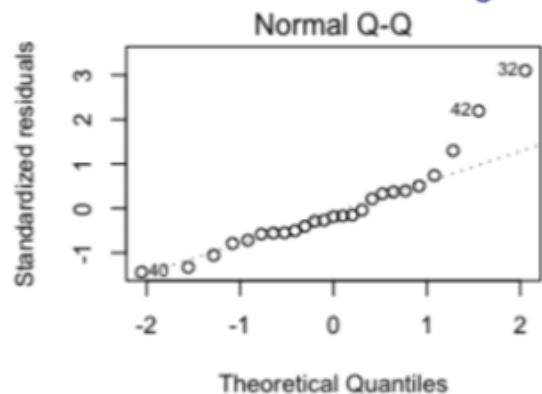
Linear relationship with X:  
Red line should be horizontal



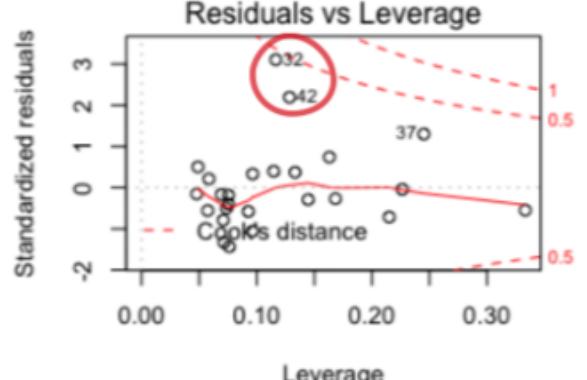
Errors with constant variance:  
Red line should be horizontal



Errors normally distributed:  
Points should follow straight line



No influential points:  
No point with Cook's D > 1



Interpretation:

- Inflation of slope variance
- $\text{SE}(\text{slope}) * \sqrt{(\text{VIF})}$

Rule of thumb:

- < 5: moderate collinearity
- > 10: problematic!

# The R Roller Coaster

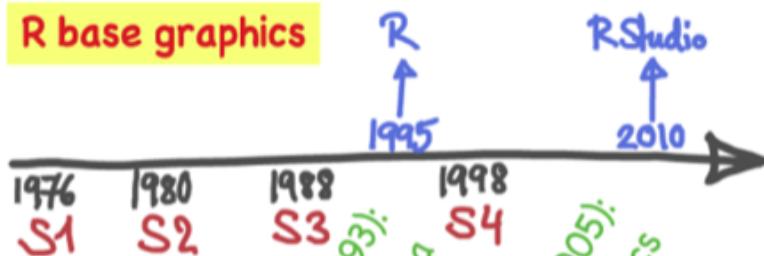
1. Understand how R thinks
2. Google it!
3. Save and re-use



'ggplot2': Grammar of Graphics for R

'lattice': Trellis graphics for R

R base graphics



Cleveland (1993):  
Visualizing Data

Wilkinson (1999, 2005):  
Grammar of Graphics



Hadley Wickham  
Chief Scientist at RStudio

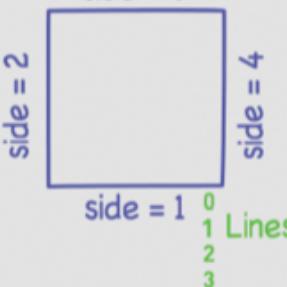
# Basic R Graphics

## File types

- png: for import
- pdf: for printing
- NOT: jpg

## Figure margins

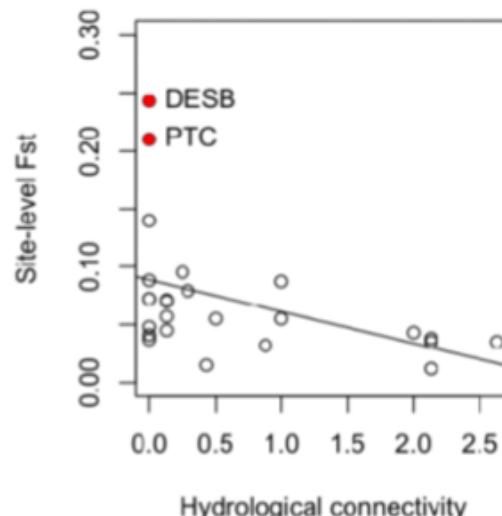
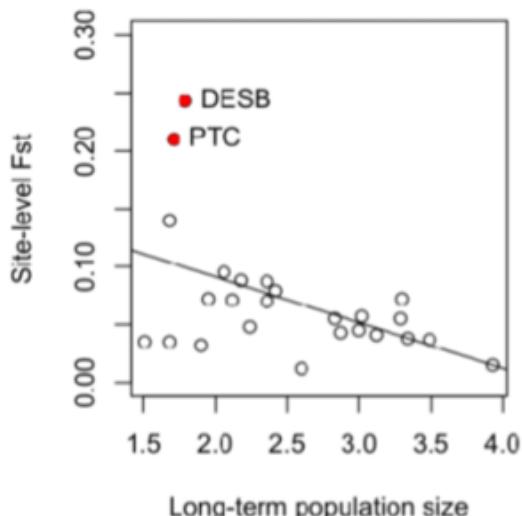
side = 3



```
png( "MyFigure.png",
    width = 7, height = 3.5, units = "in",
    pointsize = 12, res = 300)

par( mfrow = c( 1, 2 ), mar = c( 4, 4, 1, 1 ) )

plot( FST.GESTE ~ NLT, data = df )
plot( FST.GESTE ~ C, data = df )
dev.off()
```



pch	Point Types	lty	Line Types
○ 1	□ 14	—	1
△ 2	■ 15	---	2
+ 3	● 16	....	3
× 4	▲ 17	-----	4
◊ 5	◆ 18	- - -	5
▽ 6	● 19	- . - . -	6
◻ 7	● 20		
* 8	○ 21	. . . . .	.1
◊ 9	□ 22	—	.25
⊕ 10	◊ 23	—	.5
⋮ 11	△ 24	—	1
田 12	▽ 25	—	3
⊗ 13		—	6
			you can also use any character

Define axes

```
plot( y ~ x, data = df, ylim = c( 0, 0.3 ), type = "n",
      xlab = "Label x", ylab = "Label y")
```

Add data points

```
points( y ~ x, data = df, cex = 1)
```

Add fitted line

```
abline( lm( y ~ x, data = df ) )
```

Create index 'a'

```
a <- is.element( row.names( df ), c( "32", "42" ) )
```

Add red symbols

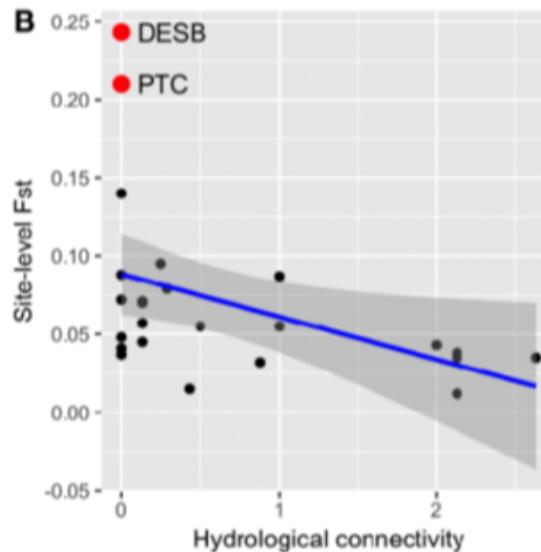
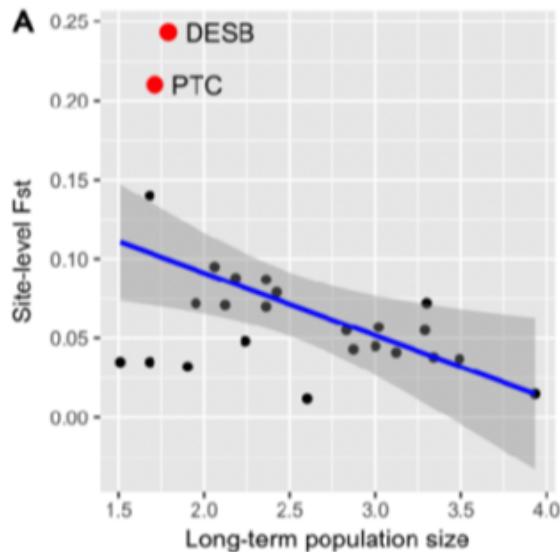
```
points( y ~ x, data = df [ a ], pch = 16, col = "red" )
```

Add labels

```
with( df [ a ], text( x , y, labels = SITE, pos = 4 ) )
```

# Grammar of Graphics: 'ggplot2'

See RStudio cheat sheet: Data Visualization with ggplot2



## Compose and store plot

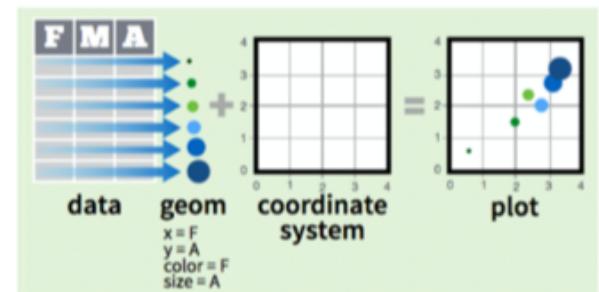
Define axes

```
Plot1 <- ggplot(data = df, aes( x, y )) +  
  xlab("Label x") + ylab("Label y") +  
  geom_point() +  
  geom_smooth(col = "blue", method="lm") +  
  geom_point(data = df[a,], color = "red",size = 3) +  
  geom_text(data = df[a,], mapping=aes(x, y, label = SITE),  
            size = 4, hjust = 0, nudge_x = 0.1)
```

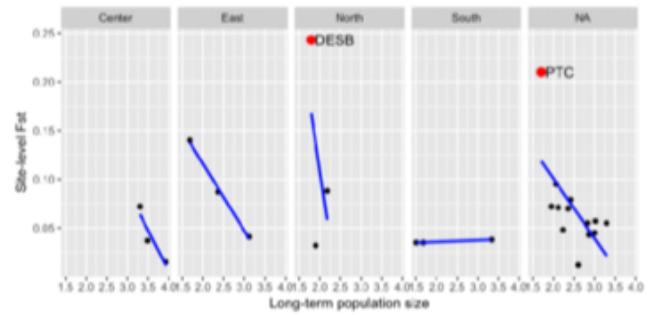
## Assemble and export figure

```
ggsave(filename = "ggplot.png", width = 7, height = 3.5, unit = "in", dpi = 300,  
       plot = cowplot::plot_grid(Plot1, Plot2, labels = c("A", "B")))
```

## Fundamentals



## Faceting: split by group



- t + facet\_grid(. ~ fl)**  
facet into columns based on fl
- t + facet\_grid(year ~ .)**  
facet into rows based on year
- t + facet\_grid(year ~ fl)**  
facet into both rows and columns
- t + facet\_wrap(~ fl)**  
wrap facets into a rectangular layout

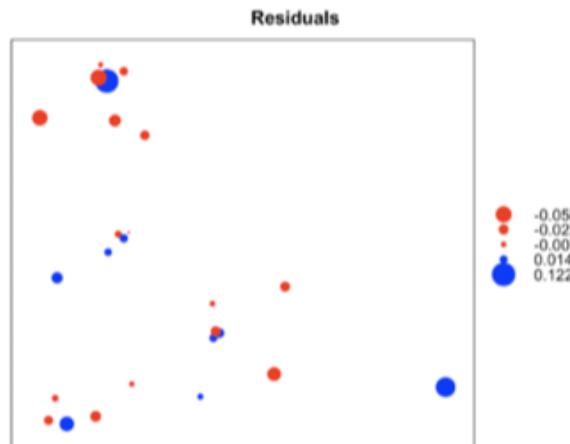
# Plotting Residuals in Space

```
mod <- lm ( FST.GESTE ~ NLT + C, data = dd.spatial )
```

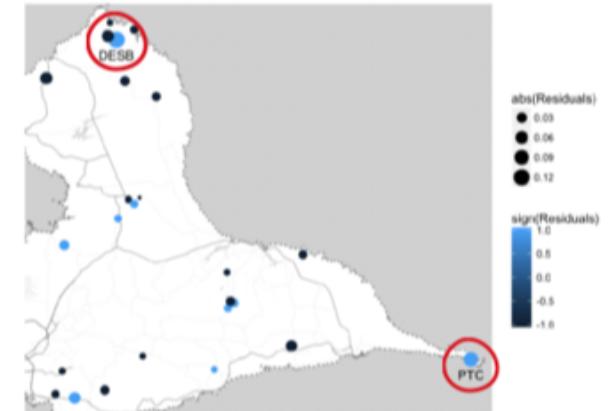
```
> attributes(mod)  
$names  
[1] "coefficients"   "residuals"        "effects"          "rank"  
[5] "fitted.values"  "assign"          "qr"              "df.residual"  
[9] "xlevels"         "call"            "terms"           "model"
```

```
Residuals <- mod $ residuals
```

lattice universe: package 'sp'



ggplot2 universe: package 'ggmap'

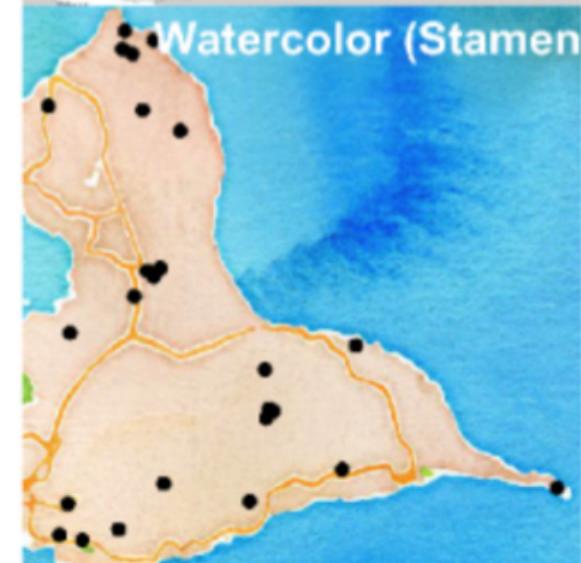
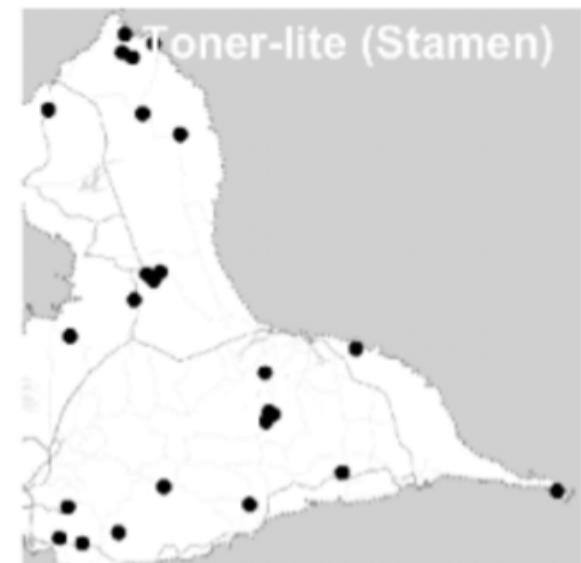


```
sp :: bubble ( dd.spatial, zcol = "Residuals",  
               col = c( "red", "blue" ) )
```

```
qmpplot ( Longitude, Latitude, data = df,  
           color = sign( Residuals ),  
           size = abs( Residuals ) ) +  
           geom_text( ... )
```

# Map Types from Internet

?ggmap :: get\_map



# Metapopulation Genetics

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- Test effect of recent extinction events

## Methodological Challenges

1. 'EcoGenetics': multi-source data
2. AMOVA with R
3. Regression analysis in R
4. Video 2: R graphics
  - Base R graphics
  - Grammar of graphics: 'ggplot2'
  - Plotting residuals in space

See also Worked Example "Week 0: R Graphics"

*D. depressissimum*



Photos: J-P Pointier

