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Verso l'Open Science in ISMAR

Giovedì 23 ottobre 2014

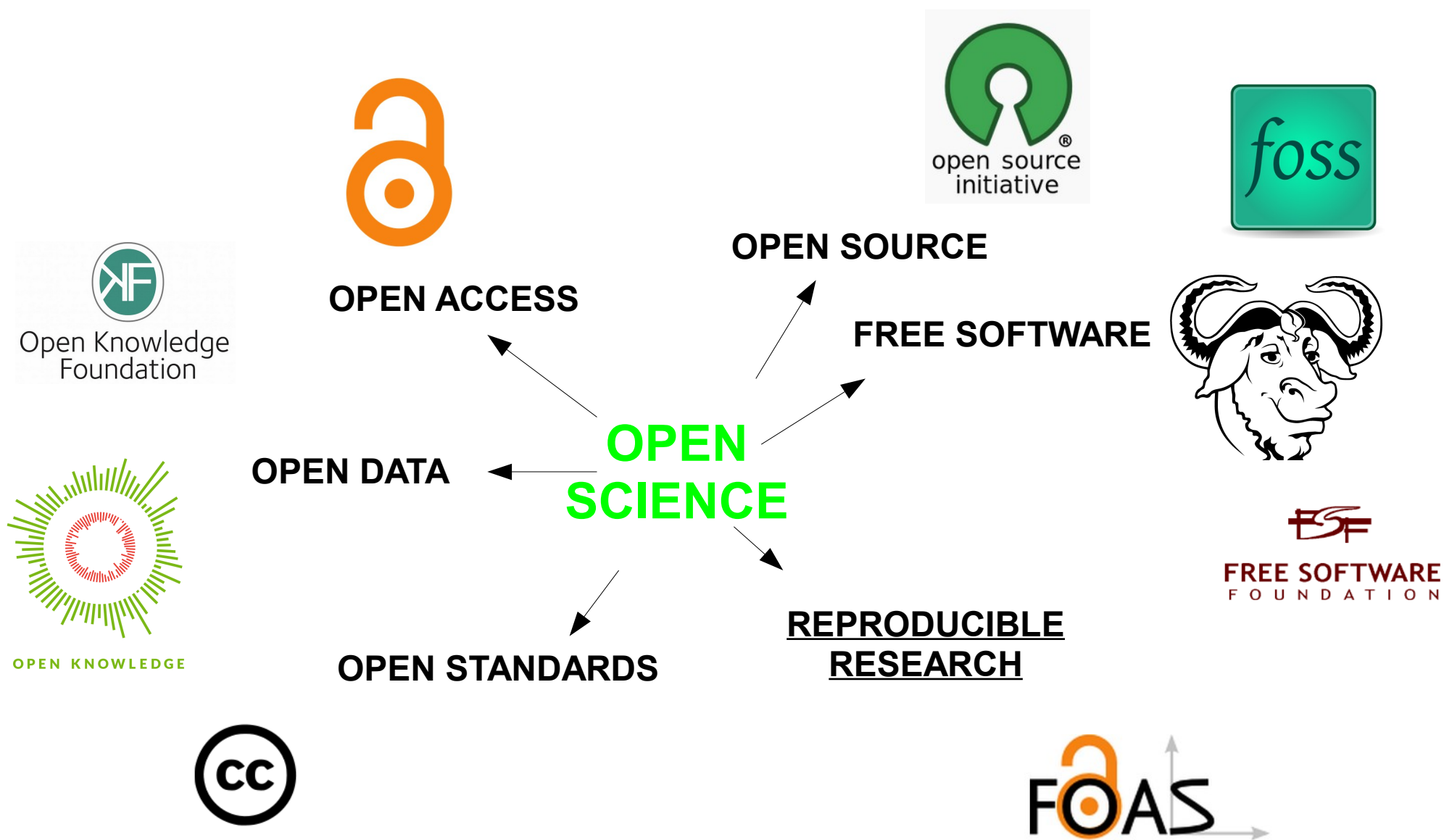


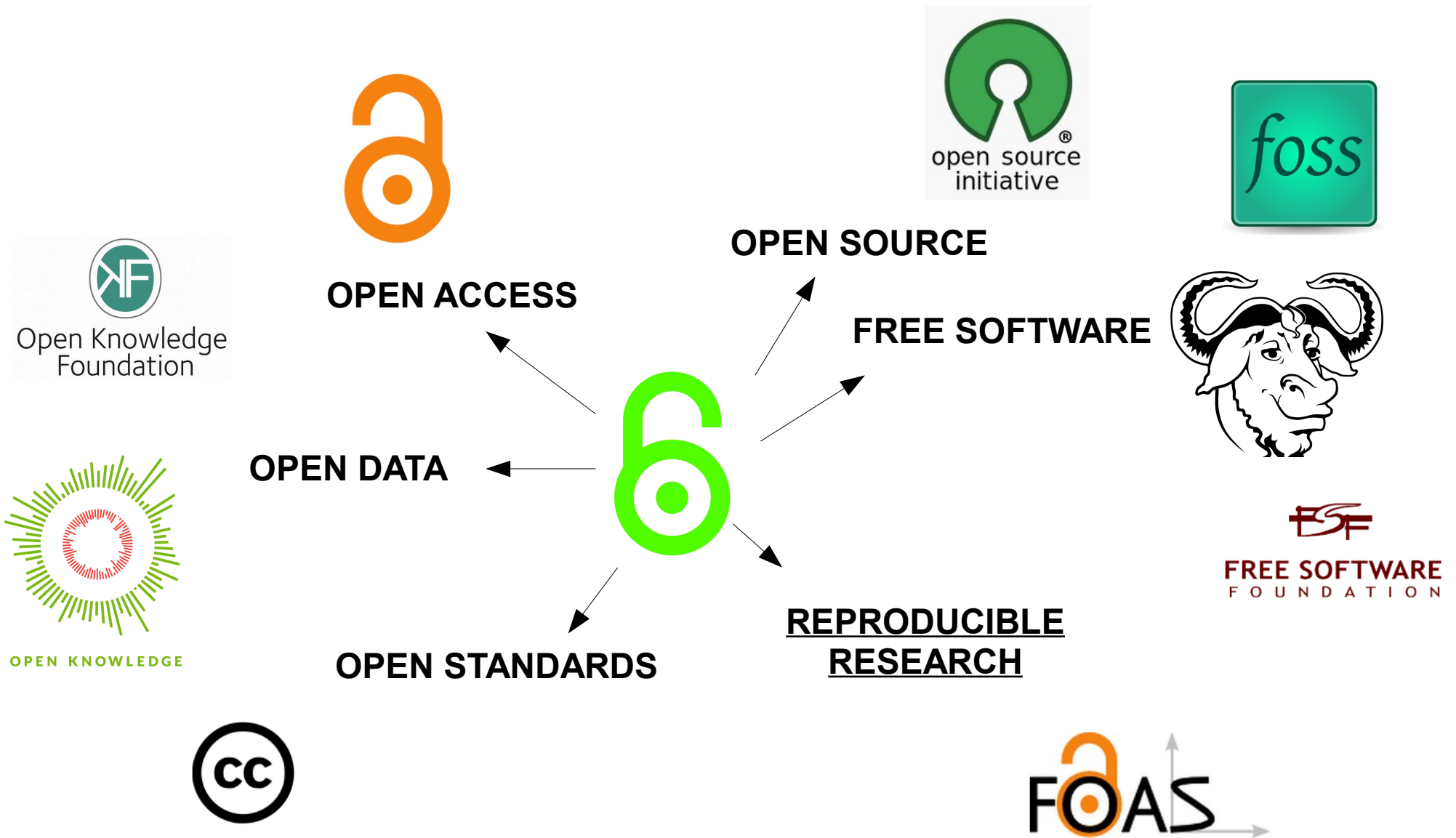
Riproducibilità nella ricerca e nelle applicazioni ecologiche

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REPRODUCIBILITY

Reproducibility is a main tenet of the scientific method.

It can be defined as “*the extent to which consistent results are obtained when produced repeatedly*” (Oxford English Dictionary). It is the ability of an experiment or study to be reproduced, either by the researcher or by a third party working independently.

Reproducibility should involve all steps in experimental studies, including the ability to recompute analysis on known data sets.

Ecology is characterized by high complexity and context-specific contingency, and it is often based on field observations and studies. Therefore, it is inherently harder to reproduce. Moreover, by aiming at maximizing reproducibility of ecological studies, relevance and generalisability of results is reduced. However, clearly, full reproducibility is expected for the analysis of raw data.

Reproducibility is here focused to the phase of data analysis.

REPRODUCIBLE RESEARCH

Reproducible research (computational science): *The ability to repeat the calculations for analyzing the data and obtaining the computational results* (Mesirov, 2010).

Big data sets and complex mathematical approaches (including inherently stochastic methods) reduce reproducibility even when data are provided and methods reported!

Gilbert et al. (2012) analyzed 23 papers with available data sets about genetic structure (software STRUCTURE): results were not reproduced in 30% of them.

In practice, RR is the paper along with the full computational environment used to perform analysis, including the code, data, etc. that can be used to reproduce the results. When a new algorithm or computational approach is proposed, the code should be fully available.

Both raw data and code need to be properly licensed (public copyright).

REPRODUCIBLE RESEARCH

In Ecology:

- methods and algorithms for ecological indices in bioindication and biomonitoring
- multivariate data analysis in community studies (many similar methods with confused nomenclature, different existing version of the algorithms, a number of parameters, complex plots, ...)
- time series analysis
- spatial analysis
- ...

REPRODUCIBLE RESEARCH

Reproducible Research allows:

- traceability and full control of methods, approaches and results
- proper and independent interpretation of the results
- linking provided data to results
- testing and validation of methods and results
- teaching and disseminating methods
- updating the analysis or an entire manuscript or manuscript when changes occur in software, source data, analysis
- coordinating different researchers
- reuse and improvement of the code in the same or other contexts
- promoting the debate

RR approaches include eg. the **Science Code Manifesto** (Barnes, 2013) and **Linked Open Science** (Kauppinen & de Espindola, 2011).

REPRODUCIBLE RESEARCH

Science Code Manifesto

- Code** All source code written specifically to process data for a published paper must be available to the reviewers and readers of the paper.
- Copyright** The copyright ownership and license of any released source code must be clearly stated.
- Citation** Researchers who use or adapt science source code in their research must credit the code's creators in resulting publications.
- Credit** Software contributions must be included in systems of scientific assessment, credit, and recognition.
- Curation** Source code must remain available, linked to related materials, for the useful lifetime of the publication.

REPRODUCIBLE RESEARCH

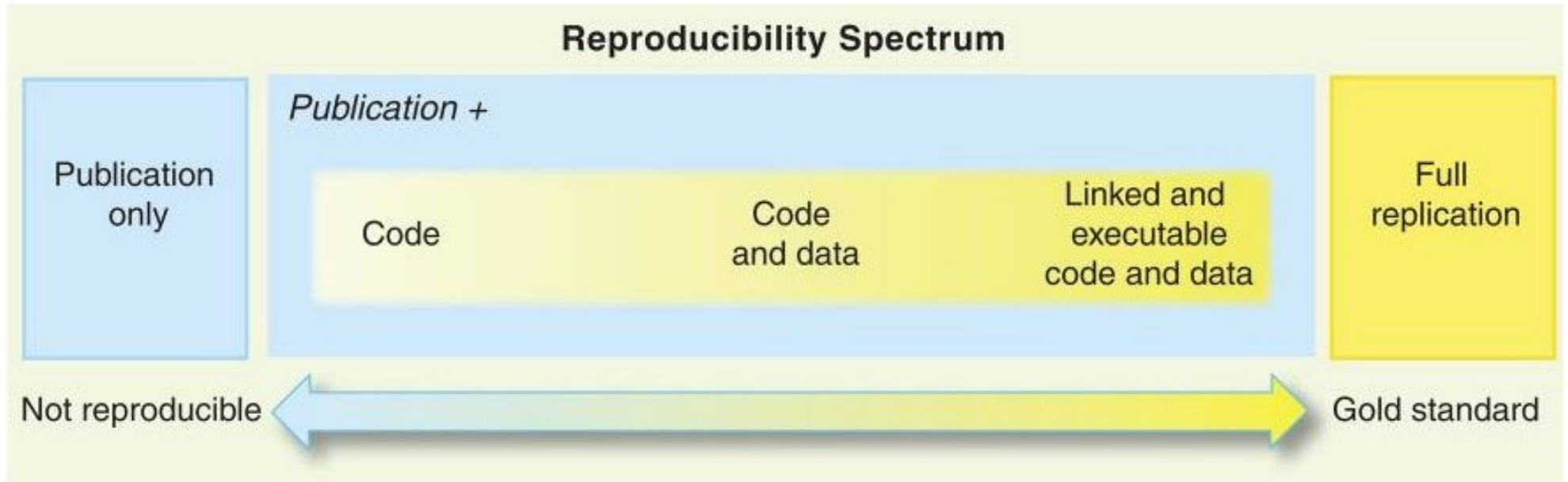
Linked Open Science

An approach in the context of Reproducible Research to solve challenges of an “executable paper”:

- 1) publication of scientific data, metadata, results, and provenance information using **Linked Data** principles,
- 2) **open source and web-based environments** for executing, validating and exploring research,
- 3) **Cloud Computing** for efficient and distributed computing,
- 4) **Creative Commons** for the legal infrastructure.

(Kauppinen & de Espindola, 2011)

REPRODUCIBLE RESEARCH



(Peng, 2011)

CASE STUDY

Reproducibility in bioindication: the M-AMBI quality index

M-AMBI

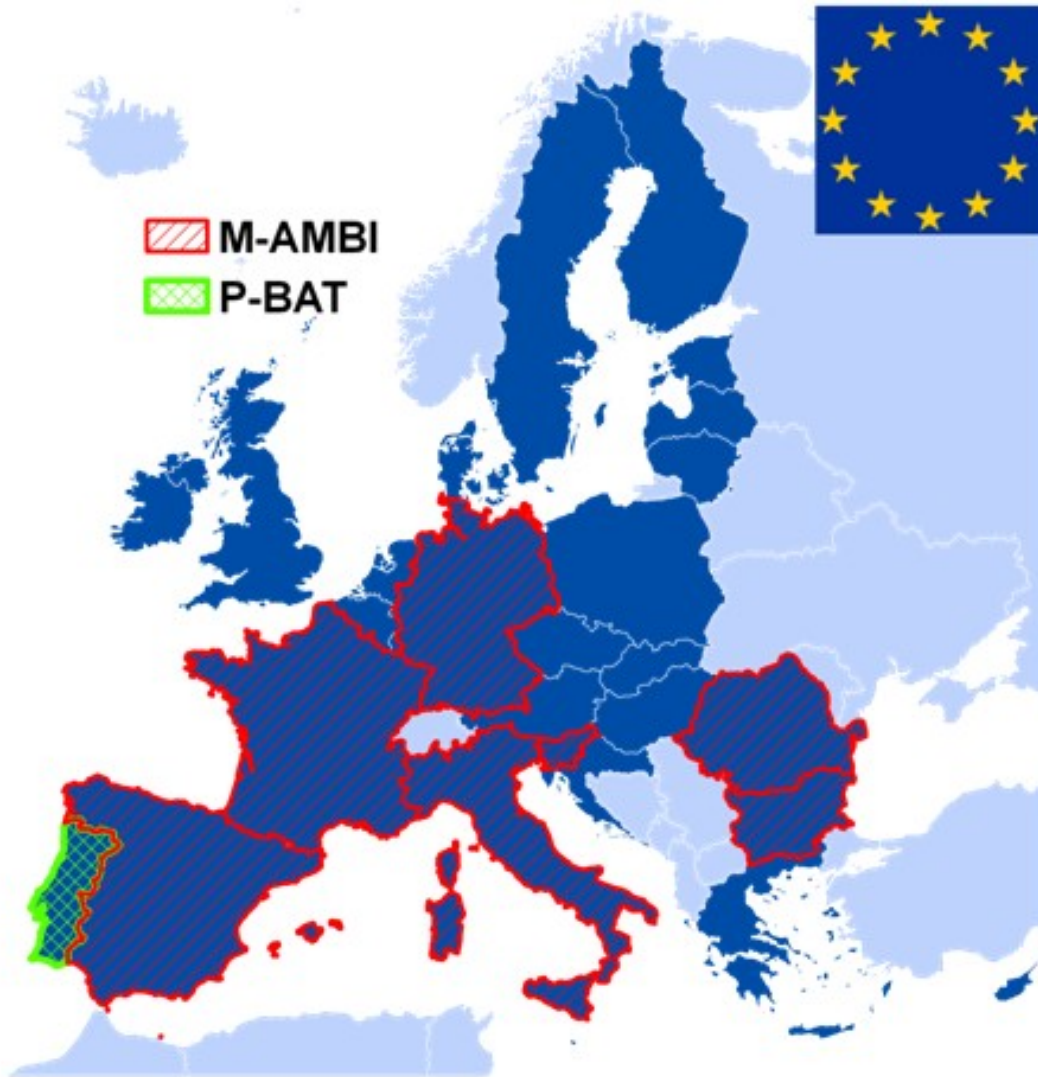
- **multivariate index of ecological quality** (Muxika et al., 2007)
- **integrates biotic index AMBI with richness (S) and Shannon diversity (H')**
- **macrozoobenthos** (soft bottom) in marine and transitional waters
- **WFD**: officially enforced in 7 EU countries
(+ Portuguese index P-BAT, Teixeira *et al.*, 2009)



biotic index AMBI (Borja *et al.*, 2000): taxa attributed to 5 ecological groups (EG, based on sensitivity); biotic coefficient BC [0; 6] is calculated based on relative abundance of EG:

$$\begin{aligned} &= \{(0 \times \% \text{GI}) + (1.5 \times \% \text{GII}) \\ &\quad + (3 \times \% \text{GIII}) + (4.5 \times \% \text{GIV}) \\ &\quad + (6 \times \% \text{GV})\}/100. \end{aligned}$$

Water Framework Directive 2000/60/CE



SOFTWARE



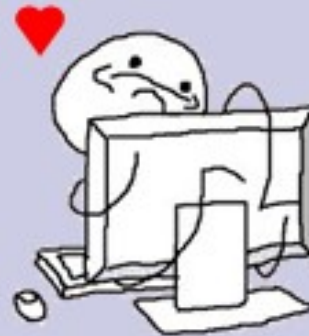
CODE



FREWARE?



USER-FRIENDLY?



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Go to M-AMBI

Work Path



Species List

v. Mar2012

View

Data Input

Input File

Browse

example.xls
(Hoja1)

Test

? Format

OK

Acquisition

Run

The species 'Eteone spetsbergensis' was not found in the library

- you can ignore it (no relevant species),
- you can consider it as Not Assigned,
- you can change the species name by the Selected one,
- or you can cancel the acquisition and correct the Input file

Note: Assignment process will be saved in results excel file

Ignore

Do not assign

Change

- Eteone sp. (III)
- Eteone tchangsii (III)
- Ethusa mascarone (Not assigned)
- Eualus cranchii (I)
- Eualus lineatus (II)
- Eualus occultus (I)
- Eualus pusiolus (I)
- Eualus sollaudi (I)
- Eubranchus sp. (Not assigned)
- Euchone analis (II)
- Euchone incolor (II)
- Euchone limnicola (III)



Developed by AZTI-Tecnalia Marine Research Division

Cancel

Summarized Results

Station	%I	%II	%III	%IV	%V	%n.a.	AMBI	BI	Disturbance Classification
1	12.5	75.0	12.5	0.0	0.0	0.0	1.475	2	Slightly disturbed
2	41.0	11.5	30.8	15.4	1.3	2.5	1.901	2	Slightly disturbed
3	1.2	0.0	97.6	0.0	1.2	0.0	2.987	2	Slightly disturbed
4	0.0	0.0	100.0	0.0	0.0	5.9	4.333	4	Moderately disturbed
5	3.3	16.7	47.2	29.4	3.3	0.0	3.130	2	Slightly disturbed
6	0.0	0.0	98.9	1.1	0.0	0.0	3.010	2	Slightly disturbed
7	0.4	0.0	93.8	0.3	5.5	0.0	3.119	2	Slightly disturbed
8	0.0	12.0	32.0	8.0	48.0	0.0	4.089	3	Moderately disturbed
9	0.1	1.1	1.2	1.4	96.2	0.0	5.887	6	Heavily disturbed
10	0.0	5.6	9.5	0.6	84.4	0.0	5.458	5	Heavily disturbed
11	0.0	8.7	4.8	2.9	83.7	0.0	5.186	5	Heavily disturbed

Select: - a station, for details

- an ensemble of stations for graphical representation, selecting them in order with Control or Shift key

Selected station results

Histogram

Space / Time distribution

Summarized Results

Station	%I	%II	%III	%IV	%V	%n.a.	AMBI	BI
1	12.5	75.0	12.5	0.0	0.0	0.0	1.475	2
2	41.0	11.5	30.8	15.4	1.3	2.5	1.901	2
3	1.2	0.0	97.6	0.0	1.2	0.0	2.987	2
4	0.0	0.0	100.0	0.0	0.0	5.9	4.333	4
5	3.3	16.7	47.2	29.4	3.3	0.0	3.130	2
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8	0.0	12.0	32.0	8.0	48.0	0.0	4.089	3
9	0.1	1.1	1.2	1.4	96.2	0.0	5.887	6
10	0.0	5.6	9.5	0.6	84.4	0.0	5.458	5
11	0.0	8.7	4.8	2.9	83.7	0.0	5.186	5

Select: - a station, for details

- an ensemble of stations for graphical representation, selecting them in order with Control or Shift key

M-AMBI

Stations	AMBI	Diversity	Richness
BAD	6	0	0
HIGH	1.47	3.54	23

1	1.475	1.80	6
2	1.901	3.54	22
3	2.987	1.88	8
4	4.333	1.82	7
5	3.130	3.54	23
6	3.010	1.67	5
7	3.119	2.29	12
8	4.089	2.50	9
9	5.887	1.15	15
10	5.458	1.84	12
11	5.186	1.68	10

Run factorial analysis :

Summarized Results

Station	%I	%II	%III	%IV	%V	%VI	%VII	%VIII	%IX	%X
1	12.5	75.0								
2	41.0	11.5								
3	1.2	0.0								
4	0.0	0.0								
5	3.3	16.7								
6	0.0	0.0								
7	0.4	0.0								
8	0.0	12.0								
9	0.1	1.1								
10	0.0	5.6								
11	0.0	8.7								

Select: - a station, for details

- an ensemble of stations for graph
selecting them in order with Con

M-AMBI

M-AMBI Results

Stations	AMBI	Diversity	Richness	X	Y	Z	M-AMBI	Status
BAD	6	0	0	3.37	2.70	1.40	0.00	Bad
HIGH	1.47	3.54	23	-2.84	-2.33	-1.07	1.00	High
1	1.475	1.80	6	0.68	-1.05	0.04	0.56	Good
2	1.901	3.54	22	-2.65	-2.04	-1.02	0.96	High
3	2.987	1.88	8	0.55	-0.23	0.10	0.51	Moderate
4	4.333	1.82	7	0.91	0.64	0.26	0.41	Moderate
5	3.130	3.54	23	-2.62	-1.33	-0.93	0.90	High
6	3.010	1.67	5	1.11	-0.04	0.25	0.44	Moderate
7	3.119	2.29	12	-0.26	-0.45	-0.15	0.60	Good
8	4.089	2.50	9	0.13	0.11	-0.12	0.53	Moderate
9	5.887	1.15	15	0.52	1.70	0.59	0.35	Poor
10	5.458	1.84	12	0.38	1.18	0.27	0.41	Moderate
11	5.186	1.68	10	0.72	1.14	0.36	0.38	Poor

Summarized Results

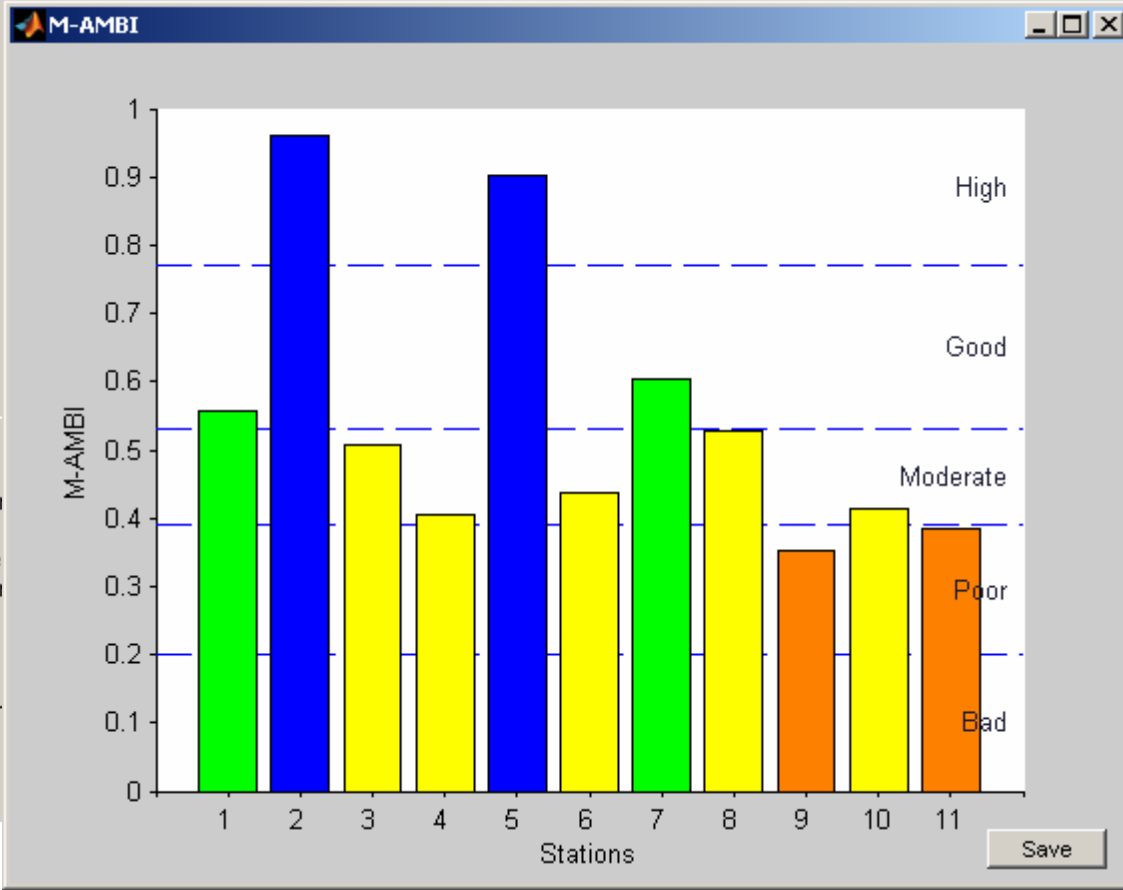
Station	%I	%II	%III	%IV	%V	%VI	%VII	%VIII	%IX	%X	%XI	%XII
1	12.5	75.0										
2												
3												
4												
5												
6												
7												
8												
9												
10												
11												

M-AMBI

Stations AMBI Diversity Richness

M-AMBI Results

Z	M-AMBI	Status
1.40	0.00	Bad
-1.07	1.00	High
0.04	0.56	Good
-1.02	0.96	High
0.10	0.51	Moderate
0.26	0.41	Moderate
-0.93	0.90	High
0.25	0.44	Moderate
-0.15	0.60	Good
-0.12	0.53	Moderate
0.59	0.35	Poor
0.27	0.41	Moderate
0.36	0.38	Poor



Select: - a station, for
- an ensemble
- selecting the

Export

--- graphs ---

Boundaries

SOFTWARE



free software environment for statistical computing and graphics

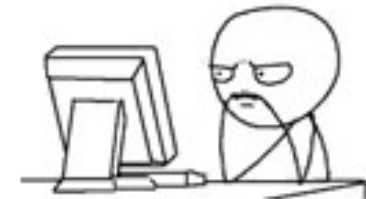
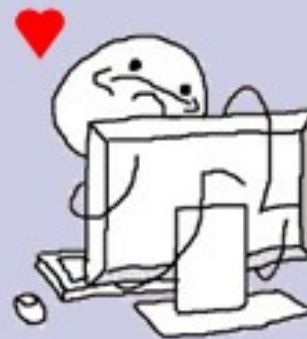
CODE



FREWARE?



USER-FRIENDLY?



OPEN?

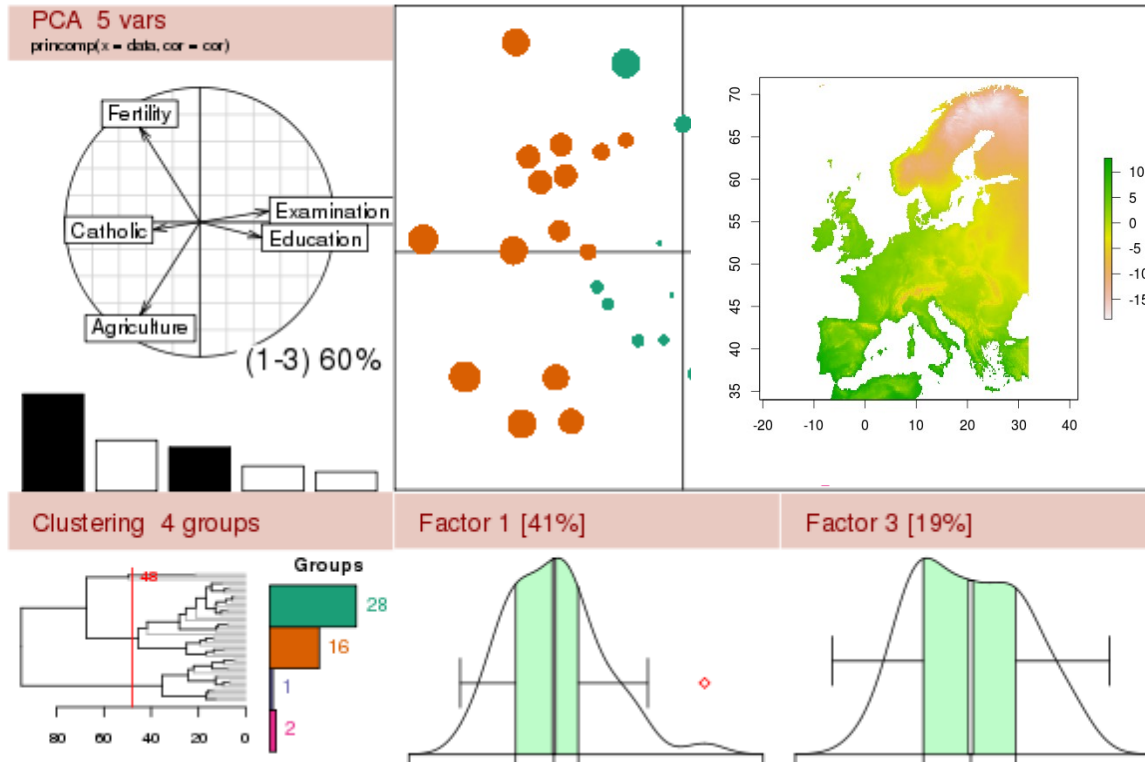


GPLv2

The R Project for Statistical Computing



www.r-project.org



R is a language and environment for statistical computing and graphics. It is a [GNU project](#) which is similar to the S language and environment which was developed at Bell Laboratories (formerly AT&T, now Lucent Technologies) by John Chambers and colleagues. R can be considered as a different implementation of S. There are some important differences, but much code written for S runs unaltered under R.

R provides a wide variety of statistical (linear and nonlinear modelling, classical statistical tests, time-series analysis, classification, clustering, ...) and graphical techniques, and is highly extensible. The S language is often the vehicle of choice for research in statistical methodology, and R provides an Open Source route to participation in that activity.

R is available as Free Software under the terms of the [Free Software Foundation's GNU General Public License](#) in source code form. It compiles and runs on a wide variety of UNIX platforms and similar systems (including FreeBSD and Linux), Windows and MacOS.

The term "environment" is intended to characterize it as a fully planned and coherent system, rather than an incremental accretion of very specific and inflexible tools, as is frequently the case with other data analysis software.

ORIGINAL ALGORITHM

Instructions for the use of the AMBI index software (Version 5.0)

Ángel Borja, Julien Mader, Iñigo Muxika¹

... When running the software, it uses **richness, diversity and AMBI as variables**. Then, the **variables are standardized**, by subtracting the mean and dividing by the standard deviation. The **Factorial Analysis uses principal components analysis as extraction method, using the correlation matrix and extracting 3 factors**. The software uses **the Varimax rotation, the regression as method for calculating the factor scores**, and missing values are excluded using **listwise**. The **factor scores (X, Y, Z)** are the new coordinates of each sampling station in the new factor space. These coordinates are used in deriving the **EQR or M-AMBI value**.

If you are using other software to calculate the EQRs, you must be aware that each software may use different algorithms for the Factor Analysis, providing different results (see Borja et al., 2008a).

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...

principal()	princomp()	prcomp()	factanal()
psych	stats	stats	?
eigendec. on cov	eigendec. on cov	svd on centered scaled data matrix	?
var: /(n-1)	var: /(n)	var: /(n-1)	?
eigenval	sdev=sqrt(eigenval)sdev=sqrt(eigenval)?		
eigenvec rescaled by sdev	eigenvec	eigenvec	?

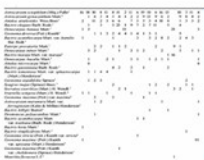
Main function in R for factor analysis



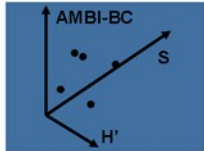
```
mambisimpl <- function(x, eg, m.values, metrics = c("S", "H1", "BC"), trasf = "F", high = "def", bad = "def",
  st = rownames(X)) {
  ## 1) METRICS
  if (missing(m.values)) {
    x <- as.matrix(X)[order(rownames(X)), order(colnames(X))]
    AMBI <- function(x, eg) {
      eg <- eg[which(rownames(eg) %in% colnames(x)), , drop = F]
      if (length(which(colnames(x) %in% rownames(eg))) < length(colnames(x))) {
        eg.na <- colnames(x)[-which(colnames(x) %in% rownames(eg))]
        eg <- rbind(eg, rep(NA, length(eg.na)))
        rownames(eg)[1:nrow(eg) - length(eg.na) + 1:nrow(eg)] <-
          colnames(x)[-which(colnames(x) %in% rownames(eg))]
      }
      eg <- eg[order(rownames(eg)), , drop = F]
      eg.matr <- as.data.frame(matrix(0, nrow = ncol(x), ncol = 6))
      colnames(eg.matr) <- c("EG1", "EG2", "EG3", "EG4", "EG5", "NA")
      rownames(eg.matr) <- rownames(eg)
      options(warn = -1)
      eg.matr[which(eg[, 1] == 1), 1] <- 1
      eg.matr[which(eg[, 1] == 2), 2] <- 1
      eg.matr[which(eg[, 1] == 3), 3] <- 1
      eg.matr[which(eg[, 1] == 4), 4] <- 1
      eg.matr[which(eg[, 1] == 5), 5] <- 1
      eg.matr[which(eg[, 1] == 0), 6] <- 1
      eg.matr[which(is.na(eg[, 1]) == T), 6] <- 1
      options(warn = 0)
      ambi <- as.data.frame(x %>% as.matrix(eg.matr))
      ambi <- cbind("N" = apply(x, 1, sum), ambi, "BC" = rep(NA, nrow(x)))
      for (i in 1:nrow(x)) {
        for (j in 2:6) {
          ambi[i, j] <- ambi[i, j] / (ambi[i, "N"] - ambi[i, "NA"]) * 100
        }
      }
      for (i in 1:nrow(x)) {
        ambi[i, "BC"] <- round((1.5*ambi[i, "EG2"] + 3*ambi[i, "EG3"] + 4.5*ambi[i, "EG4"] +
          5*ambi[i, "EG5"]) / 100, 3)
      }
      colnames(ambi)[2:6] <- c("EG1(X)", "EG2(X)", "EG3(X)", "EG4(X)", "EG5(X)")
      ambi[, 2:6] <- round(ambi[, 2:6], 2)
      return(ambi)
    }
    ambi <- AMBI(x, eg)
    ## samples with N = 0
    if (any(is.nan(ambi$BC))) {ambi$BC[which(is.nan(ambi$BC))] <- 7}
    ## calculation of AMBI-BC on each replicate (other metrics are calculated on pooled replicates)
    st <- st[order(rownames(X))]
    st <- as.factor(st)
    ambi.mu <- round(aggregate(ambi$BC ~ st, FUN = mean)[2], 3)[, 1]
    x.st <- aggregate(x ~ st, FUN = sum)
    rownames(x.st) <- x.st[, 1]
    x.st <- x.st[-1]
    colnames(x.st) <- colnames(x)
    ## calculation of richness/diversity metrics
    require(vegan)
    m.values <- data.frame(
      "S" = specnumber(x.st),
      "d" = round((specnumber(x.st) - 1) / (log(apply(x.st, 1, sum))), 3),
      "H1" = round(diversity(x.st, index = "shannon", base = 2), 3),
      "BC" = ambi.mu)
    ## richness
    ## Margalef index
    ## Shannon index
    ## AMBI-BC
    sample.names <- rownames(x.st)
    METRICS <- NULL
    for (i in 1:length(metrics)) METRICS <- cbind(METRICS, m.values[, metrics[i]])
    colnames(METRICS) <- metrics
    rownames(METRICS) <- sample.names
  }
  else {
    ambi <- NULL
    sample.names <- rownames(m.values)
    METRICS <- as.matrix(m.values)
    colnames(METRICS) <- metrics
    rownames(METRICS) <- sample.names
  }
  ## 2) "HIGH" AND "BAD" REFERENCES VALUES ADDED AS FICTITIOUS SAMPLES
  options(warn = -1)
  if (is.numeric(high) == T) METRICS.high <- high
  if (is.numeric(bad) == T) METRICS.bad <- bad
  if (high == "def") {
    METRICS.high <- rep(NA, length(metrics))
    for (i in 1:length(metrics)) {
      if (metrics[i] == "BC") METRICS.high[i] <- min(METRICS[, i])
      else METRICS.high[i] <- max(METRICS[, i])
    }
  }
  if (bad == "def") {
    METRICS.bad <- rep(NA, length(metrics))
    for (i in 1:length(metrics)) {
      if (metrics[i] == "BC") METRICS.bad[i] <- 5
      else METRICS.bad[i] <- 0
    }
  }
  options(warn = 0)
}
```


M-AMBI

(Muxika et al., 2007; Borja et al., 2012)

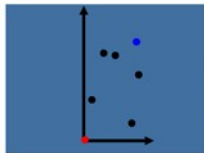


1. species / abundance matrix

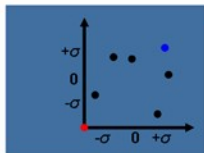


2. three metrics (S, H', BC)

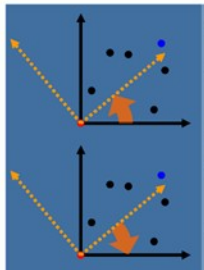
(plot 3D → 2D for displaying purposes)



3. High/Bad reference values added

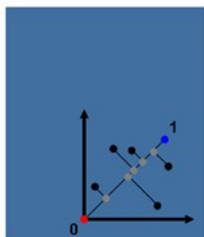


4. standardisation



5. factor analysis:

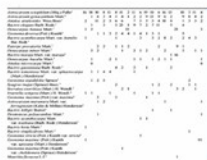
- i. eigendecomposition (→ PCA)
- ii. factor selection
- iii. eigenvectors rescaling by $\sqrt{\text{eigenvalues}}$
- iv. Varimax rotation



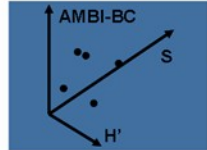
6. scores (new coordinates) projected on the line identified by ref. values and normalised 0-1

M-AMBI

(Muxika et al., 2007; Borja et al., 2012)

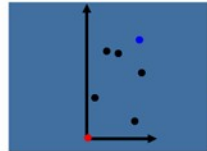


1. species / abundance matrix

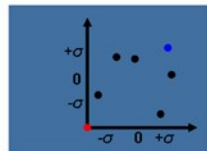


2. three metrics (S, H', BC)

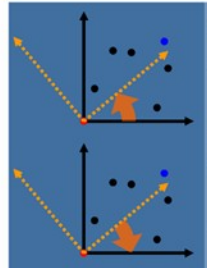
(plot 3D → 2D for displaying purposes)



3. High/Bad reference values added

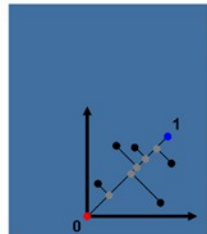


4. standardisation



5. factor analysis:

- i. eigendecomposition (→ PCA)
- ii. factor selection
- iii. eigenvectors rescaling by $\sqrt{\text{eigenvalues}}$
- iv. Varimax rotation



6. scores (new coordinates) projected on the line identified by ref. values and normalised 0-1



M-AMBI^{*} (m)



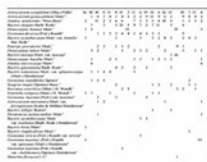
4. normalisation



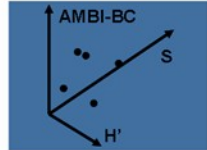
(mathematically identical to the simple mean)

M-AMBI

(Muxika et al., 2007; Borja et al., 2012)

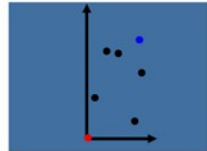


1. species / abundance matrix

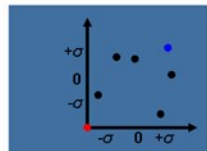


2. three metrics (S, H', BC)

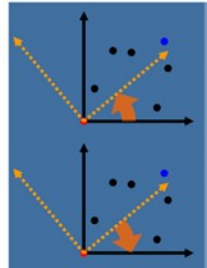
(plot 3D → 2D for displaying purposes)



3. High/Bad reference values added

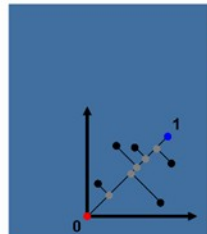


4. standardisation



5. factor analysis:

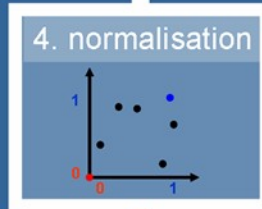
- i. eigendecomposition (→ PCA)
- ii. factor selection
- iii. eigenvectors rescaling by $\sqrt{\text{eigenvalues}}$
- iv. Varimax rotation



6. scores (new coordinates) projected on the line identified by ref. values and normalised 0-1

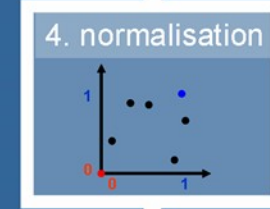
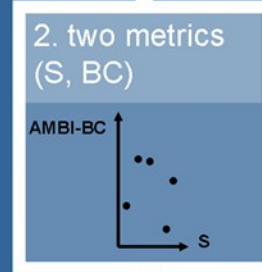


M-AMBI^{*}_(m)



(mathematically identical to the simple mean)

S-AMBI_(m)

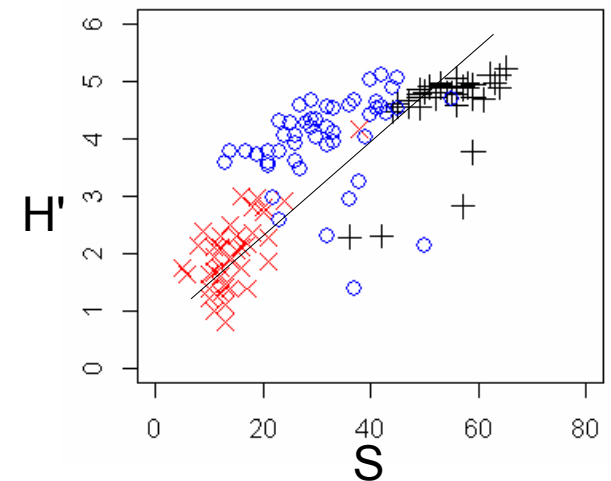


PROPOSED ALGORITHMS

$$\text{M-AMBI}^*_{(n)} = \frac{S_{(n)} + H'_{(n)} + \text{AMBI BC}_{(n)}}{3}$$

$$\text{S-AMBI}^*_{(n)} = \frac{S_{(n)} + \text{AMBI BC}_{(n)}}{2}$$

The less mutually correlated the metrics are, the more the multimetric index is interpretable.



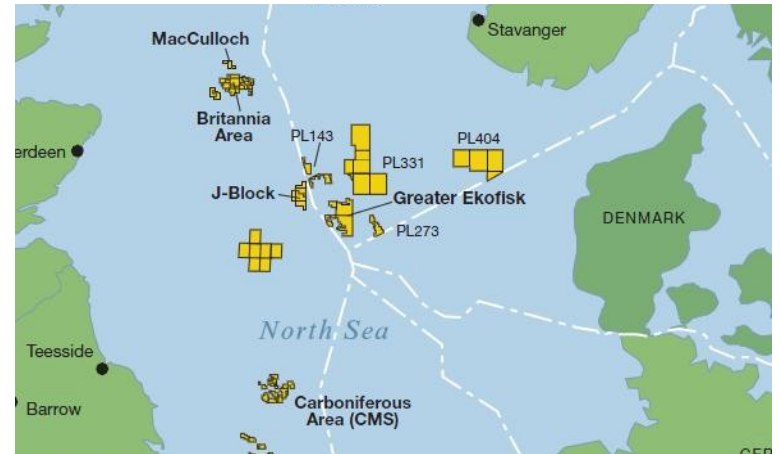
DATASETS

1. Ekofisk

Survey of the Ekofisk oil field (Norway), 1987

Gray *et al.*, 1990; in PRIMER software for community analysis (Clarke & Gorley, 2006)

39 samples (0.1 m², 3 rep., 1 mm mesh), 13,883 individuals, 173 taxa.

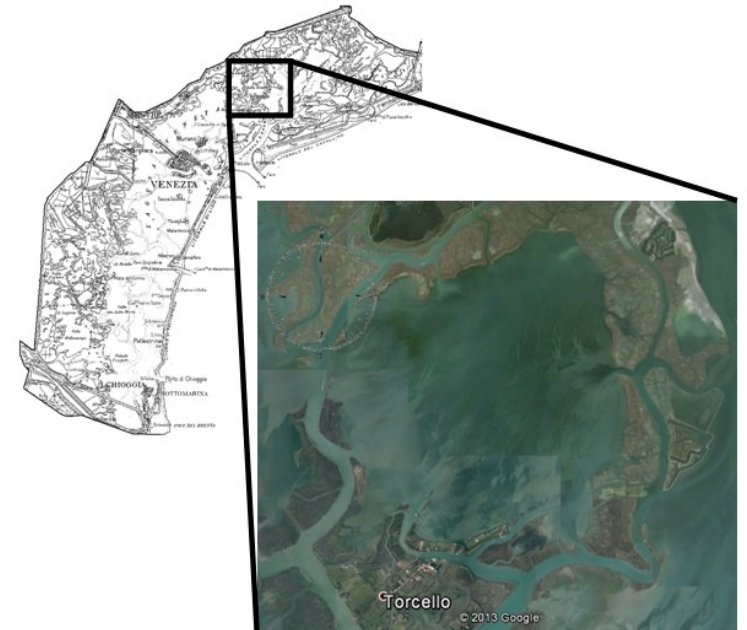


2. Venice

Palude della Rosa, Lagoon of Venice (Italy), 1991

Tagliapietra *et al.*, 1998

42 stations (0.1 m², 1 mm mesh), 34,732 individuals, 62 taxa.

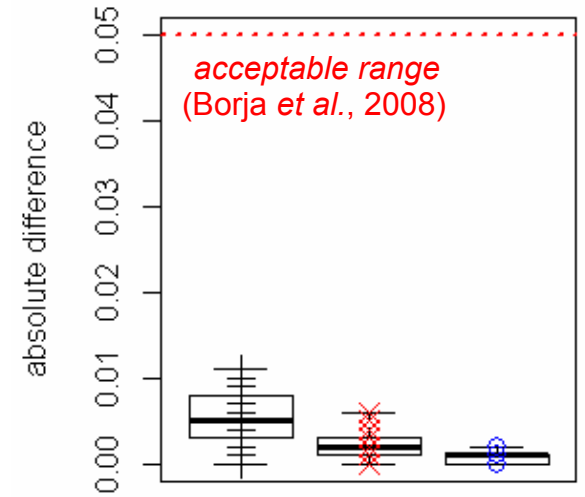
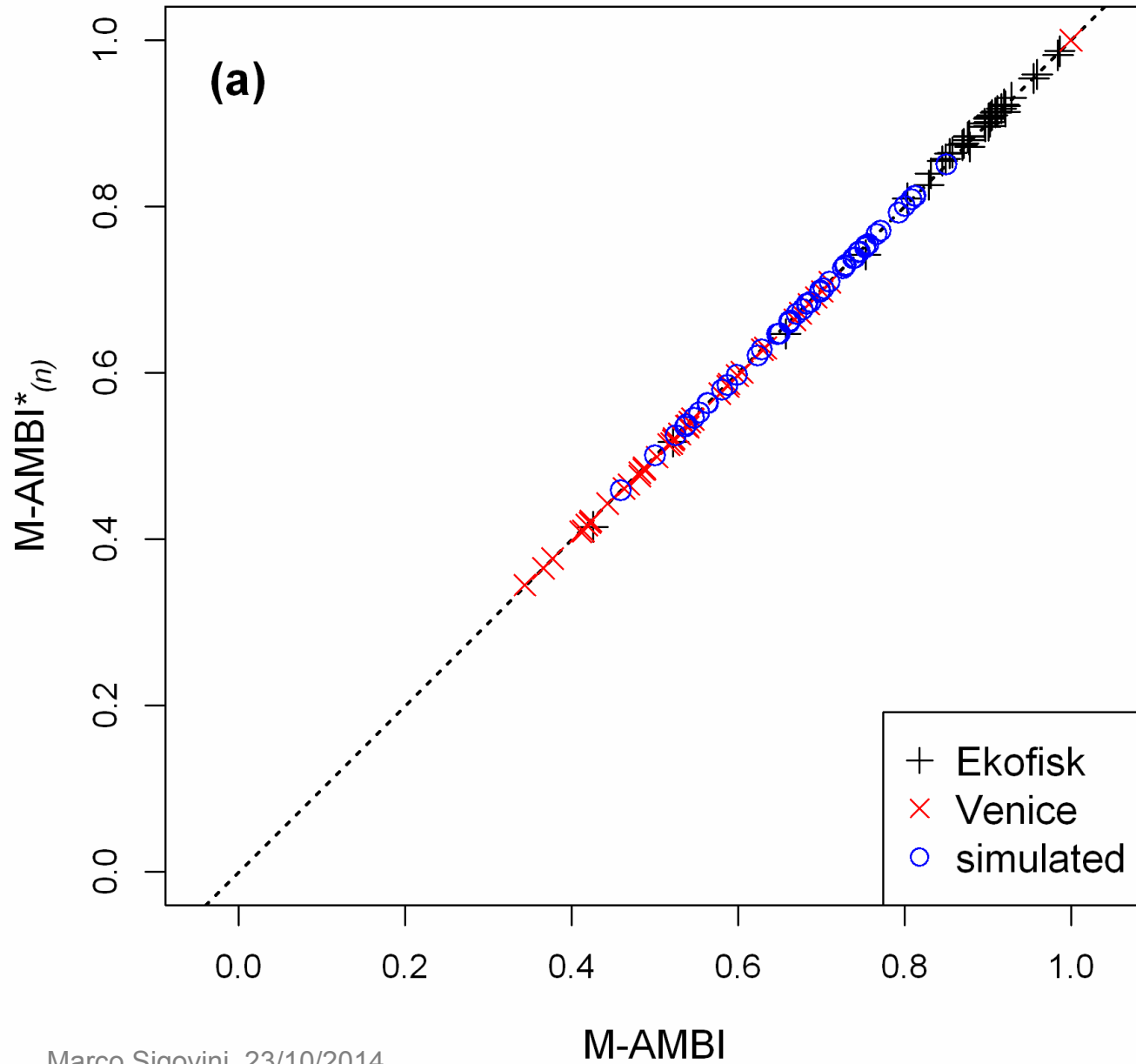


3. Simulated

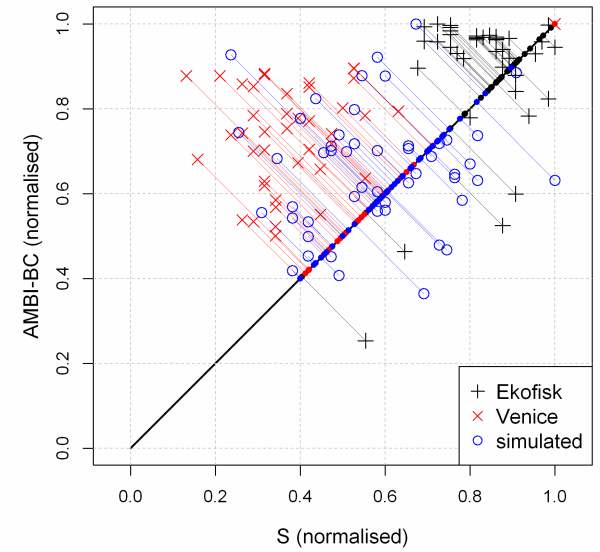
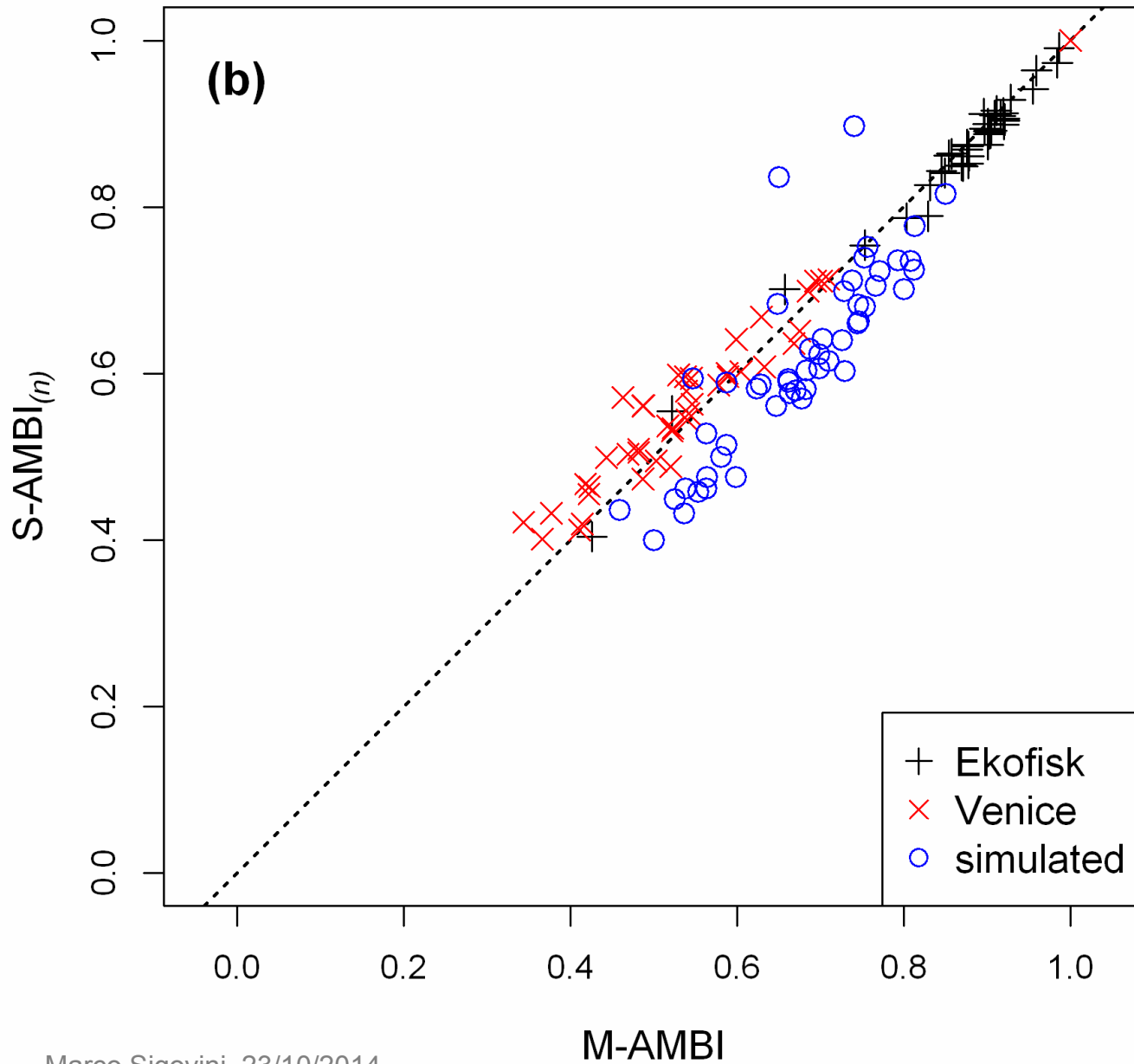
Poisson lognormal distribution, with parameters μ and σ randomly generated (from a normal distribution) for each sample. EG randomly assigned. 50 samples, 120 fictitious species.

-> independence among AMBI and S, H'

RESULTS: M-AMBI*_(n)



RESULTS: S-AMBI_(n)



Sensitivity + Diversity components

CONCLUSIONS

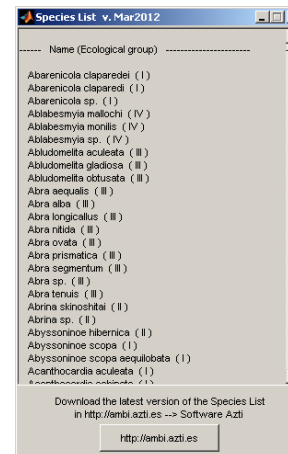
1) Factor Analysis is demonstrated not functional to M-AMBI

FA results are affected by the criteria used to choose the model, decide how many factors to retain, select the rotation method.; addition of new data gives different results (in fact, very small deviations).

2) M-AMBI is closely approximated by the simple mean of the normalised metrics independent of the number of sample

3) Bivariate version of M-AMBI (S-AMBI: diversity + species sensitivity index) still highly correlated.

4) → free the species (list)!



CONCLUSIONS

lex parsimoniae (Ockham's Razor)

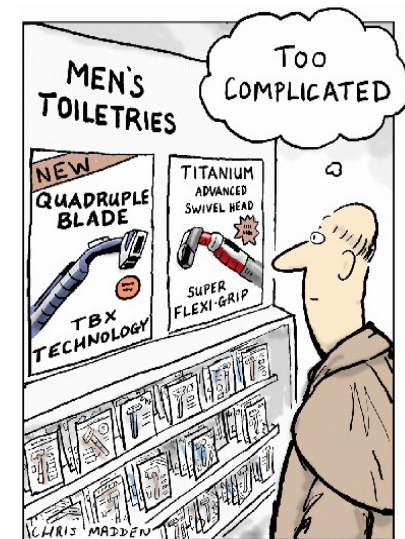
“The model with the fewest assumptions should be preferred“

Desirable proprieties (proposed algorithms):

- Transparency, Reproducibility, Openness
- Testability, Falsifiability
- Simplicity
- Generalisability
- Robustness
- Ecological meaningfulness (and significance)

Reproducing the algorithm on open source software was instrumental to understand and test its features, identify flaws and propose improvements.

Ecologists should consider to provide source code (and data) when proposing mathematical tools or publishing results (eg. as suppl. material)



M-AMBI revisited: looking inside a widely-used benthic index

Marco Sigovini · Erica Keppel ·
Davide Tagliapietra





www.r-project.org



ONLINE SOURCES about Reproducible Research with R:

<http://cran.r-project.org/web/views/ReproducibleResearch.html>

<http://www.r-bloggers.com/reproducible-research-is-still-a-challenge/>

<http://www.r-bloggers.com/making-reproducible-research-enjoyable/>

<http://www.r-bloggers.com/brief-introduction-on-sweave-and-knitr-for-reproducible-research/>

<https://osf.io/s9tya/> (Stodden et al., 2013, Implementing Reproducible Research, 448 pp.)

<http://ropensci.org/blog/2014/06/09/reproducibility/>

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Gray et al., 1990. Marine Ecology Progress Series, 66: 285–299

Kauppinen & de Espindola, 2011. Procedia Computer Science, 4: 726–731

Mesirov, 2010. Science, 327(5964)

Muxika et al., 2007. Marine Pollution Bulletin, 55: 16–29

Peng, 2011. Science, 334(6060): 1226–1227

Tagliapietra et al., 1998. Estuarine Coast and Shelf Science, 47: 217–226

Teixeira et al., 2009. Marine Pollution Bulletin, 58: 1477–1486

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