

e-Learning Statistics – A Selective Review

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e-Learning

- web-based teaching materials and hypermedia
- authoring systems
- simulations and animations
- e-mail, blogs and wikis
- learning management/Learning content management systems
- videoconferencing and teleteaching



e-Learning

- e-learning is a big theme in modern education
- large-scale projects: NYU: 20'000'000 \$, eCornell: 40'000'000 \$, UK e-Universities: 62'000'000 £
- Institute for Statistics and Econometrics (ISE) involved in different smaller projects

“E-learning has not really revolutionized learning and teaching to date. Far-reaching, novel ways of teaching and learning, facilitated by ICT, remain nascent or still to be invented.”

OECD (2005)



Outline

1. Motivation ✓
2. A typical education in statistics
3. e-Learning materials
 - ▶ MM*Stat
 - ▶ Electronic books
 - ▶ Q & A
 - ▶ e-stat
 - ▶ Moodle
4. Conclusion



Course Structure – Bachelor (B.Sc.)

Group	Class	e-Media
Introduction	Statistics I & II	MM*Stat Q & A
Multivariate Statistics	Multivariate Statistics I XploRe Introductory Course	e-stat XploRe
Applied Statistics	Computerbased Statistics I & II Data Mining/Statistical Learning Numerical Introductory Course	e-Books Excel R
Privatissimum	Privatissimum Bachelor Thesis	



Course Structure – Master (M.Sc.)

Group	Class	e-Media
Multivariate Statistics	Multivariate Statistics I XploRe Introductory Course	e-stat XploRe
Statistics of Fin. Markets	Statistics of Fin. Markets I & II	R
Advanced Statistics	Multivariate Statistics II Non- and Semiparametrics I & II Applied Quantitative Methods	Matlab
Privatissimum	Privatissimum Master Thesis	



Course Structure - PhD

Group	Class	e-Media
Financial Statistics	Quantitative Finance Seminar	XploRe
	Adv. Stat. Methods in Finance	eBooks
	Mathematical Statistics Seminar	Matlab
	Statistical Tools in Finance and Insurance	



MM*Stat

- HTML/JavaScript/CSS
- generated with \LaTeX
- filing-card structure
- contents:
 - ▶ lectures, explanations
 - ▶ examples, multiple choice questions
 - ▶ interactive examples



1.5 Measurement Scales

The values random variables take can differ distinctively:

Symbol	Variable	Sample space
X	Age (rounded to years)	$\{0, 1, 2, \dots\}$
S	Sex	{female, male}
T	Marital status	{single, married, divorced}
Y	Monthly income	$[0, \infty)$

They can be classified into quantitative, i.e. numerically valued (age and income) and qualitative, i.e. categorical, (sex, marital status) variables. As numerical values are usually assigned to observations of qualitative variables, they may appear quantitative. Yet such synthetic assignments aren't of the same quality as numerical measurements that naturally arise in observing a phenomenon. The crucial distinction between quantitative and qualitative variables lies in the properties of the actual **scale of measurement**, which in turn is crucial to the applicability of statistical methods. In developing new tools statisticians make assumptions about permissible measurement scales.

A measurement is a numerical assignment to an observation. Some measurements appear more natural than others. By measuring the height of persons, for example, we apply a yardstick that ensures comparability between observations up to almost any desired precision—regardless of the units (such as inches or centimeters). School grades, on the other hand, represent a relatively rough classification indicating a certain ranking, yet putting many pupils into the same category. The values assigned to qualitative statements like 'very good', 'average' etc. are an arbitrary yet practical short cut in assessing people's achievements. As there is no conceptual reasoning behind a school grade scale, one should not try to interpret the 'distances' between grades.

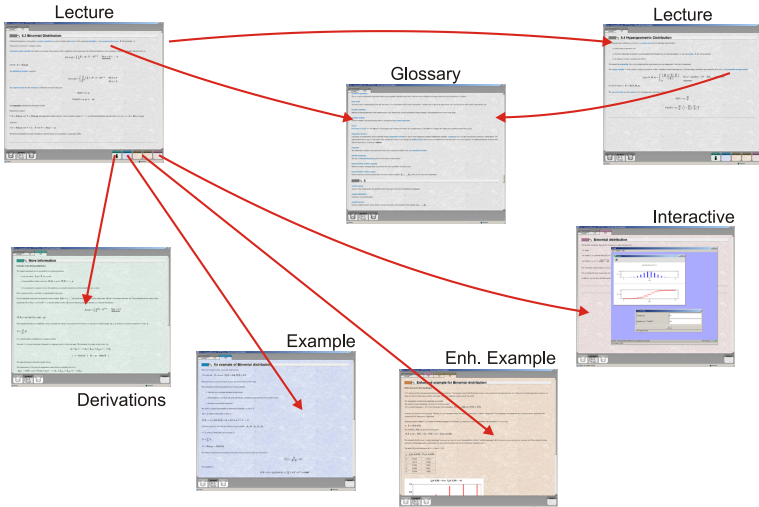
Clearly, height measurements convey more information than school marks, as distances between measurements can consistently be compared. Statements such as 'Tom is twice as tall as his son' or 'Manuela is 35 centimeters smaller than her partner' are permissible.

As statistical methods are developed in mathematical terms, the applicable scales are also defined in terms of mathematical concepts. These are the transformations that can be imposed on them without loss of information. The wider the range of permissible transformations, the less information the scale can convey. The following table lists common measurement scales in increasing order of information content. Scales carrying more information can always be transformed into less informative scales.

Variable	Measurement Scale	Statements	Permissible Transformations
Qualitative	Nominal Scale	equivalence	any equivalence preserving mapping



MM*Stat Linking of Topics



MM*Stat – Review

- + positive feedback from students, partly replaced traditional media such as books
- + simple access via webbrowser
 - no significant improvement of grades
 - complicated technical structure (browser war)
 - focus on economics
 - cultural differences and accessibility issues not considered
 - sustainability



Electronic Books (MD*Book Framework)


- 'golden solution', used for various books
 - ▶ embed special links in the \LaTeX source
 - ▶ generate respective HTML-pages
 - ▶ pages for 'edit', 'execute' and download of XploRe files
- HTML pages use Java applets with XploRe Quantlet Server
- additionally: easily generate PDF and PostScript versions



Applied Multivariate Statistical Analysis - Microsoft Internet Explorer

Datei Bearbeiten Ansicht Favoriten Extras ?

Adresse <http://www.quantlet.org/mdstat/codes/mva/MVAregpull.html> Wechsln zu



$$f_k(x) = G_k \left(\omega_{k0}^{(k)} + \sum_{j=1}^m \omega_{kj}^{(k)} G_j \right) G_0$$

home execute edit help

MVAregpull

Description: MVAregpull computes a linear regression of sales (X1) on price (X2) from the pullovers data set ("pullover.dat")

Download: [MVAregpull.xpl](#)

Code:

```

library("xplore")
x<-read("pullover")           ; reads the pullovers data
y=x[,1]                       ; prices (X2)
x<-matrix(rows(x))~x[,2]     ; constant & sales (X1)
beta=gls(x, y)               ; computes beta (lin. regression)
d=x[,2]~y                    ; data points
line=min(x[,2])|max(x[,2])
m=line~((1|1)-line)*beta     ; regression line
setmaskl(m, (1: rows(m))', 0, 2, 2)
setmaskp(m, 0, 0, 0)
di=createdisplay(1, 1)
show(di, 1, 1, m, d)         ; shows data and regression line
setgopt(di, 1, 1, "title", "pullovers data")
setgopt(di, 1, 1, "xlabel", "price (X2)", "ylabel", "sales (X2)")

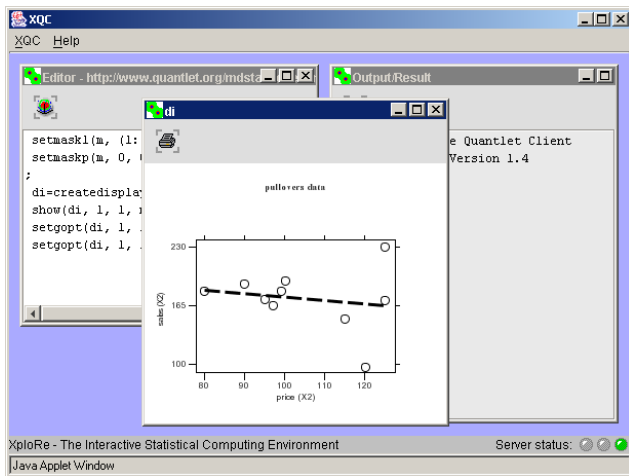
```

MD*Tech Method and Data Technologies

Internet

⇒ [Link](#)

Electronic Books – Example Applet



Electronic Books – Review

- + based mostly on standard software (\LaTeX)
- + different output formats, searchable index (Google)
- + allows real-world data to be used
 - complex structure, requires profound knowledge
 - not independent from underlying computing engine
 - accessibility not considered



e-stat

- developed by a group of German universities
- focus: provide materials for students and people with interest in statistics
- content delivered as XML modules following a DTD
- XML code arranged in views and scenarios with three levels of abstraction: introductory (A), applied (B), advanced (C)
- different ways of learning, based on methods, problems or views
- <http://www.emilea.de>



http://emileA-stat.rwth-aachen.de - EMILeA-stat - Microsoft Internet Explorer

Stöbern in EMILeA-stat

Maschine Learning > Klassifikation > Support Vektor Maschinen

Support Vektor Maschinen | SVM Lernen | Support Vektoren | Soft Margin SVM | Lernen von Soft Margin SVM | Support Vektoren Soft C

Level A | Level B | Level C

Inhaltsverzeichnis

- EMILeA-stat Modulwelt
 - Amliche Statistik
 - Assoziation
 - Beschreibende Statistik
 - Entropie
 - Explorative Datenanalyse
 - Finanzmathematik
 - Lineare Strukturgleichungen
 - Maschine Learning
 - Datengruppierung
 - Dichteschätzung
 - Klassifikation
 - Boosting
 - Fishers Lineare Diskriminanzanalyse
 - Fishers Multiple Diskriminanzanalyse
 - Funktionenzahlen
 - Generalisierte Lineare Diskriminanzfunktion
 - K-nächste Nachbarklassifikation
 - Klassifikation mit Mischverteilungen
 - Kleinste-Quadrate-Lösung
 - Linear Separierbare Objekte
 - Lineare Diskriminanzfunktionen
 - Lineare Klassifikation
 - Mehrklassenfall Allgemein
 - Neuronale Netze
 - Optimierungs Theorie
 - Perzeptron-Kriterium
 - Support Vektor Maschinen**
 - Validierung
 - Mathematische Grundlagen
 - Methodenkritische Begleitung zu PISA 2000
 - Numerische Methoden
 - Qualitätsoptimierung
 - Ausreißertests
 - Einführung in die robuste Statistik
 - Robuste Regressionsverfahren
 - Robuste Schätzer für Lageparameter
 - Schließende Statistik
 - Sequenzielle Methoden
 - Statistik der Finanzmärkte
 - Stochastik in der Schule

Abbildung Das Prinzip einer Support Vektor Maschine: Es soll die Hyperebene gefunden werden, die den Rand m maximiert.

Abbildung 1 veranschaulicht das Prinzip der Support Vektor Maschine. Der Rand m einer Hyperebene ist der Abstand zum nächst Trainingsvektor. Die Länge des Gewichtsvektors \mathbf{a} wird normiert, ansonsten könnte man den Rand durch Skalierung des Gewichtsvektors über jede Schranke heben.

Die Trainingsvektoren \mathbf{x}_i die auf dem Rand liegen, d.h. für die gilt

$$\frac{y_i \mathbf{a}^T \mathbf{x}_i}{\|\mathbf{a}\|} = m \quad \text{Formel : (3.131)}$$

heßen **Supportvektoren** (Stützvektoren). Die Supportvektoren bestimmen die separierende Hyperebene. Läßt man die übrigen Vektoren weg, so ändert sich die optimale Entscheidungsgrenze im Sinne der Supportvektormaschine nicht.

XploRe | SPSS | Formaldruckung als Bild | Internet

e-stat – Review

- + covers different courses of studies
- + different levels of abstraction and ways of accessing topics
- + sustainability
 - complex framework required to add new content
 - for only a subset of topics all levels have been implemented
 - accessibility not considered

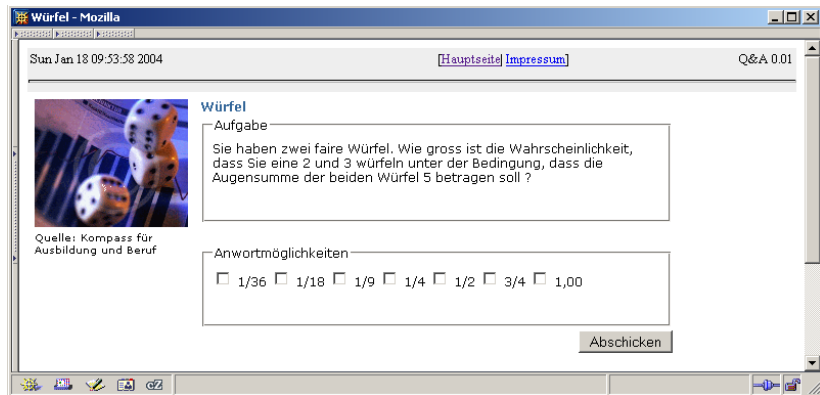


Q & A (Questions and Answers)

- idea: interactive, web-based environment for practicing
- typically 300-400 students per year in undergraduate statistics
- evaluation of exams showed 'typical' errors
- students can choose via checkbox correct solution
- if wrong answer is given, hint on the error is presented



Q & A (Questions and Answers)



Sun Jan 18 09:53:58 2004 [\[Hauptseite\]](#) [\[Impressum\]](#) Q&A 0.01

Würfel

Aufgabe

Sie haben zwei faire Würfel. Wie gross ist die Wahrscheinlichkeit, dass Sie eine 2 und 3 würfeln unter der Bedingung, dass die Augensumme der beiden Würfel 5 betragen soll ?

Quelle: Kompass für Ausbildung und Beruf

Anwortmöglichkeiten

1/36 1/18 1/9 1/4 1/2 3/4 1,00



Q & A – Review

- + covers typical error sources
- + positive feedback and hints from students
- + follows structure of the examination
 - requires huge amount of time from the lecturer
 - students tend to trial-and-error to find correct solution
 - accessibility not considered



Moodle

- learning management software
- open source software based on PHP and MySQL
- helps lecturers create effective online learning communities
- Moodle allows to
 1. structure lectures according to a set of topics or a time table,
 2. upload and manage documents (slides, exercises) for lecturers and students
 3. set up and maintain forums for student/student and student/teacher communication
 4. generate online exercises with direct evaluation.



Course: Selected Topics in Computational Statistics: Numerical Introductory Course - Microsoft Internet Explorer

Datei Bearbeiten Ansicht Favoriten Extras ? Preispielen

Advanced search ?

Administration

- Grades
- Edit profile
- Change password
- Unenrol me from 70195-WS0506

My courses

- Nonparametric Statistical Methods
- [Selected Topics in Computational Statistics: Numerical Introductory Course](#)
- Statistik I & II (SS 05 + WS 05/06)
- XploRe Introductory Course
- Moodle-Kursanmeldung
- [Selected Topics in Advanced Statistics \(Applied Quantitative Methods\)](#)

[All courses...](#)

Thursday 18-20: 1

Grading will composed of

- 1.) 45 min presentation (35%)
 - description of the method
 - advantages/disadvantages
 - short proofs only (if you like)
 - application of the method to (economic) data
- 2.) participation in the seminar with questions or comments (25%)
- 3.) contributing to the "Numerical Methods in Statistics" (40%)

[Nachrichtenforum](#)

[Web page of the course](#)

[Numerical Methods in Statistics](#)

[Template: LaTeX slides, see "Slides: new MD*Stat Style \(BEAMER\)"](#)

[Lecture MIT 18.335: Introduction to Numerical Methods](#)

[Numerical Recipes online](#)

15.11.2005

[Basic linear algebra and Gram-Schmidt orthogonalization](#)

[Sorting algorithms \(158611\)](#)

22.11.2005

[Numerical differentiation \(501351\)](#)

[Numerical integration \(188281\)](#)

Sigbert Klinke
Stellenausschreibung für
studentische Hilfskraft
[more...](#)

6 Nov, 21:23
Sigbert Klinke
Seminar work / Presentation
[more...](#)

24 Okt, 16:28
Sigbert Klinke
Date and time of seminar
[more...](#)
[Older topics ...](#)

Upcoming Events

There are no upcoming events

[Go to calendar...](#)
[New Event...](#)

Recent Activity

Activity since Donnerstag,
12 Januar 2006, 08:52
[Full report of recent activity...](#)

Nothing new since your last login

Internet

Moodle – Review

- + gives good overview about the number of students
- + allows communication with all registered students
- + simplified file management (no HTML editing necessary)
- + accessibility for physically handicapped
 - requires a certain initial amount of time
 - advantage for students



Other Packages

Microsoft Excel available everywhere, well-known interface

DoLStat@d by Okayama University, Japan: web based learning system using real world data together with their analysis stories

'Neue Statistik' by Freie Universität Berlin: problem-oriented and practical approach using animations, diagrams and video sequences



Conclusion

- e-Learning may provide valuable help for students & lecturers
- often huge time investment on the teachers' side without noticeable improvement in students' understanding
- consequent evaluation of is rarely done, accesslogs are insufficient
- require willingness to adjust the own behavior from students and teachers
- students only use them if they feel a real advantage
- best practice: start with simple solution, have students evaluate, evaluate, evaluate. . .



References



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