

# Syllabus

## *Descrizione corso*

<b>Titolo insegnamento</b>	Statistical Methods
<b>Codice insegnamento</b>	27502
<b>Titolo aggiuntivo</b>	
<b>Settore Scientifico-Disciplinare</b>	SECS-S/01
<b>Lingua</b>	Inglese
<b>Corso di Studio</b>	Corso di laurea magistrale in Data Analytics for Economics and Management
<b>Altri Corsi di Studio (mutuati)</b>	M1 Statistical methods for business analysis is loaned from course 25559 – Master in Entrepreneurship and Innovation (LM-77 EI) M2 Advanced statistics is loaned from course 73006 – Master in Computing for Data Science (LM-18)
<b>Docenti</b>	prof. Alessandro Casa, Alessandro.Casa@unibz.it <a href="https://www.unibz.it/en/faculties/economics-management/academic-staff/person/46549">https://www.unibz.it/en/faculties/economics-management/academic-staff/person/46549</a>
<b>Assistente</b>	
<b>Semestre</b>	Secondo semestre
<b>Anno/i di corso</b>	1
<b>CFU</b>	12
<b>Ore didattica frontale</b>	M1: 36 hours M2: 40 hours
<b>Ore di laboratorio</b>	M1: 18 hours M2: 20 hours
<b>Ore di studio individuale</b>	-
<b>Ore di ricevimento previste</b>	M1: 18 hours M2: 18 hours
<b>Sintesi contenuti</b>	M1: This module begins with a review of key principles of statistical inference and then introduces core concepts in statistical learning. Topics include linear regression and its extensions, advanced

	<p>regression techniques such as decision trees, logistic regression, classification methods, model selection strategies, and unsupervised learning approaches like principal component analysis and clustering. Throughout the course, students will work hands-on in R, applying techniques to real-world datasets drawn from business scenarios. By the end, students will be able to choose suitable statistical models, apply them to a range of business problems, and effectively communicate their analytical insights</p> <p>M2:</p> <ul style="list-style-type: none"> <li>• Parameter estimation: maximum likelihood methods</li> <li>• Parameter estimation: Bayesian inference</li> <li>• Time series: components and forecasting</li> <li>• Time series: causal relationship tests</li> <li>• Missing data</li> <li>• Elements of statistics for Big Data</li> </ul>
<b>Argomenti dell'insegnamento</b>	<p>M1:</p> <ul style="list-style-type: none"> <li>- Review of statistical inference: random variables, confidence intervals, and hypothesis testing.</li> <li>- Introduction to statistical learning concepts: basic vocabulary and notions, parametric and nonparametric approaches, predictive and inferential objectives, bias-variance trade off, supervised and unsupervised learning</li> <li>- Linear regression and extensions: simple and multiple linear regression, model estimation and assessment, model assumptions, inferential tools, qualitative predictors, interaction effects, polynomial regression, basic notions on nonparametric regression</li> <li>- Classification: introduction to classification, logistic regression, model estimation, evaluation of classifiers</li> <li>- Other supervised learning techniques: trees, splines, additive models</li> <li>- Model selection/assessment and evaluation of model complexity: resampling methods, cross-validation and information criteria</li> <li>- Unsupervised learning: clustering tools such as k-means and hierarchical clustering, principal component analysis</li> <li>- Applications with the R software</li> </ul> <p>M2:</p> <ul style="list-style-type: none"> <li>- Parameter estimation: maximum likelihood methods</li> <li>- Parameter estimation: Bayesian inference</li> </ul>

	<ul style="list-style-type: none"> <li>- Time series: components and forecasting</li> <li>- Time series: causal relationship tests</li> <li>- Missing data</li> <li>- Elements of statistics for Big Data</li> </ul>
<b>Parole chiave</b>	Statistical learning, regression, classification, clustering, dimensionality reduction, model selection , statistical inference, Bayesian statistics, time series modelling, big data
<b>Prerequisiti</b>	<p>M1: No formal prerequisites are required. Nonetheless, knowledge of basic concepts in descriptive and inferential statistics is useful, and attending a pre-course in mathematics/statistics is recommended.</p> <p>M2: the knowledge provided by a course in calculus and one in probability and statistics</p>
<b>Insegnamenti propedeutici</b>	
<b>Modalità di insegnamento</b>	<p>M1: In-person lectures and computer labs. Whenever possible, lectures will be structured to prioritize in-class time for discussions, and practical applications.</p> <p>M2: Frontal lectures, discussions and exercises on computer.</p>
<b>Obbligo di frequenza</b>	Recommended, but not required.
<b>Obiettivi formativi specifici e risultati di apprendimento attesi</b>	
<b>Obiettivi formativi specifici e risultati di apprendimento attesi (ulteriori info.)</b>	
<b>Modalità di esame</b>	<p>The overall exam mark will be determined by the assessment of the two modules (M1+M2).</p> <p>M1: Assessment (for both attending and non-attending students):</p> <ul style="list-style-type: none"> <li>- Written Exam: Exercises and review questions (65% of the final grade).</li> <li>- Data Analysis Project: Group project in which students select and analyze an interesting dataset using the tools learned in the</li> </ul>

	<p>course. Groups will present their work at the end of the course (35% of the final grade; optional).</p> <p>Notes:</p> <ul style="list-style-type: none"> <li>- For students who do not complete the project, the written exam will count for 100% of the final grade.</li> <li>- Project grades remain valid for one academic year.</li> </ul> <p>M2:</p> <p>The assessment is based on class and lab participation, home-work exercises and a final written exam. The final written exam will include open questions and exercises to be worked out by the students as well as computational exercises to be solved with R.</p>
<b>Criteri di valutazione</b>	<p>M1:</p> <ul style="list-style-type: none"> <li>- Written exam: understanding of statistical concepts, correct interpretation of results of statistical analyses, clarity and precision of explanations.</li> <li>- Data Analysis Project: Quality and clarity of the presentation, adequacy and appropriateness of analyses with respect to dataset characteristics</li> </ul> <p>M2:</p> <p>For attending students the final grade will be determined by the evaluation of homeworks, class and lab participation (20%) and the evaluation of a final written exam (80%).</p> <p>The homeworks and the final written exam are separately evaluated with a score expressed in 30/30.</p> <p>For non-attending students the final grade will be determined by the evaluation of a final written exam (100%). The final written exam is evaluated with a score expressed in 30/30.</p>
<b>Bibliografia obbligatoria</b>	<p><b>M1:</b></p> <p>James, G., Witten, D., Hastie, T., Tibshirani, R. An Introduction to Statistical Learning with Applications in R. Springer, 2013. Freely available at <a href="http://www-bcf.usc.edu/~gareth/ISL/">http://www-bcf.usc.edu/~gareth/ISL/</a></p> <p>Slides and lecture notes provided</p>

	<p><b>M2:</b></p> <p>Randall Pruim, 2018, <i>Foundations and Applications of Statistics An Introduction Using R</i>. American Mathematical Society, Providence. ISBN 9781470428488. From this book we discuss topics from chapters 4 and 5.</p> <p>Robert Shumway and David Stoffer, 2019. <i>Time Series: A Data Analysis Approach Using R</i>. CRC Press, Boca Raton. ISBN 9780367221096. From this book we discuss chapters 1 to 4 and some optional topics from chapters 5 and 8.</p>
<b>Bibliografia facoltativa</b>	<p><b>M1:</b></p> <p>Bishop, C. M. (2006). Pattern recognition and machine learning. New York: Springer.</p> <p>Agresti, A., Finlay, B. Statistica per le scienze sociali, Pearson, 2009.</p> <p>Hyndman, R.J. and Athanasopoulos, G. Forecasting: principles and practice, 2nd edition, OTexts: Melbourne, 2018.</p> <p>Cicchitelli, Giuseppe. Statistica. Principi e metodi. Pearson, 2008.</p> <p>Azzalini, Adelchi, and Bruno Scarpa. Data analysis and data mining: An introduction. OUP USA, 2012.</p> <p>Grigoletto, Matteo, Laura Ventura, and Francesco Pauli. Modello lineare: teoria e applicazioni con R. G Giappichelli Editore, 2017.</p> <p>Johnson, Richard A., and Dean W. Wichern. "Applied multivariate statistical analysis." New Jersey 405 (1992).</p>

	<b>M2:</b>  Additional material and readings provided in class by the lecturer.
<b>Altre informazioni</b>	
<b>Obiettivi di Sviluppo Sostenibile (SDGs)</b>	Buona salute, Lotta contro il cambiamento climatico, Ridurre le disuguaglianze, Buona occupazione e crescita economica

## *Modulo del corso*

<b>Titolo della parte costituente del corso</b>	M1 - Statistical methods for business analysis
<b>Codice insegnamento</b>	27502A
<b>Settore Scientifico-Disciplinare</b>	SECS-S/01
<b>Lingua</b>	Inglese
<b>Docenti</b>	prof. Alessandro Casa, Alessandro.Casa@unibz.it <a href="https://www.unibz.it/en/faculties/economics-management/academic-staff/person/46549">https://www.unibz.it/en/faculties/economics-management/academic-staff/person/46549</a>
<b>Assistente</b>	
<b>Semestre</b>	Secondo semestre
<b>CFU</b>	6
<b>Docente responsabile</b>	
<b>Ore didattica frontale</b>	36
<b>Ore di laboratorio</b>	18
<b>Ore di studio individuale</b>	-
<b>Ore di ricevimento previste</b>	18
<b>Sintesi contenuti</b>	This module begins with a review of key principles of statistical inference and then introduces core concepts in statistical learning. Topics include linear regression and its extensions, advanced regression techniques such as decision trees, logistic regression, classification methods, model selection strategies, and unsupervised learning approaches like principal component analysis and clustering. Throughout the course, students will work hands-on in R, applying techniques to real-world datasets drawn from business scenarios. By the end, students will be able to choose

	suitable statistical models, apply them to a range of business problems, and effectively communicate their analytical insights
<b>Argomenti dell'insegnamento</b>	<ul style="list-style-type: none"> <li>- Review of statistical inference: random variables, confidence intervals, and hypothesis testing.</li> <li>- Introduction to statistical learning concepts: basic vocabulary and notions, parametric and nonparametric approaches, predictive and inferential objectives, bias-variance trade off, supervised and unsupervised learning</li> <li>- Linear regression and extensions: simple and multiple linear regression, model estimation and assessment, model assumptions, inferential tools, qualitative predictors, interaction effects, polynomial regression, basic notions on nonparametric regression</li> <li>- Classification: introduction to classification, logistic regression, model estimation, evaluation of classifiers</li> <li>- Other supervised learning techniques: trees, splines, additive models</li> <li>- Model selection/assessment and evaluation of model complexity: resampling methods, cross-validation and information criteria</li> <li>- Unsupervised learning: clustering tools such as k-means and hierarchical clustering, principal component analysis</li> <li>- Applications with the R software</li> </ul>
<b>Modalità di insegnamento</b>	In-person lectures and computer labs. Whenever possible, lectures will be structured to prioritize in-class time for discussions, and practical applications.
<b>Bibliografia obbligatoria</b>	<p>James, G., Witten, D., Hastie, T., Tibshirani, R. An Introduction to Statistical Learning with Applications in R. Springer, 2013. Freely available at <a href="http://www-bcf.usc.edu/~gareth/ISL/">http://www-bcf.usc.edu/~gareth/ISL/</a></p> <p>Slides and lecture notes provided</p>
<b>Bibliografia facoltativa</b>	<p>Bishop, C. M. (2006). <i>Pattern recognition and machine learning</i>. New York: Springer.</p> <p>Agresti, A., Finlay, B. Statistica per le scienze sociali, Pearson, 2009.</p>

	<p>Hyndman, R.J. and Athanasopoulos, G. Forecasting: principles and practice, 2nd edition, OTexts: Melbourne, 2018.</p> <p>Cicchitelli, Giuseppe. Statistica. Principi e metodi. Pearson, 2008.</p> <p>Azzalini, Adelchi, and Bruno Scarpa. Data analysis and data mining: An introduction. OUP USA, 2012.</p> <p>Grigoletto, Matteo, Laura Ventura, and Francesco Pauli. Modello lineare: teoria e applicazioni con R. G Giappichelli Editore, 2017.</p> <p>Johnson, Richard A., and Dean W. Wichern. "Applied multivariate statistical analysis." New Jersey 405 (1992).</p>
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## *Modulo del corso*

<b>Titolo della parte costituente del corso</b>	M2 - Advanced statistics
<b>Codice insegnamento</b>	27502B
<b>Settore Scientifico-Disciplinare</b>	SECS-S/01
<b>Lingua</b>	Inglese
<b>Docenti</b>	
<b>Assistente</b>	
<b>Semestre</b>	Secondo semestre
<b>CFU</b>	6
<b>Docente responsabile</b>	
<b>Ore didattica frontale</b>	40
<b>Ore di laboratorio</b>	20
<b>Ore di studio individuale</b>	-
<b>Ore di ricevimento previste</b>	18
<b>Sintesi contenuti</b>	<ul style="list-style-type: none"> <li>• Parameter estimation: maximum likelihood methods</li> <li>• Parameter estimation: Bayesian inference</li> <li>• Time series: components and forecasting</li> </ul>



	<ul style="list-style-type: none"> <li>• Time series: causal relationship tests</li> <li>• Missing data</li> <li>• Elements of statistics for Big Data</li> </ul>
<b>Argomenti dell'insegnamento</b>	
<b>Modalità di insegnamento</b>	Frontal lectures, discussions and exercises on computer.
<b>Bibliografia obbligatoria</b>	<p>Randall Pruim, 2018, <i>Foundations and Applications of Statistics An Introduction Using R</i>. American Mathematical Society, Providence. ISBN 9781470428488. From this book we discuss topics from chapters 4 and 5.</p> <p>Robert Shumway and David Stoffer, 2019. <i>Time Series: A Data Analysis Approach Using R</i>. CRC Press, Boca Raton. ISBN 9780367221096. From this book we discuss chapters 1 to 4 and some optional topics from chapters 5 and 8.</p>
<b>Bibliografia facoltativa</b>	Additional material and readings provided in class by the lecturer.