

# Summary Descriptions of Core Curriculums

Yukun (Edward) Zhang

## Table of Contents

<i>Graduate Courses: UBC</i> .....	3
<b>DSCI 591: Capstone Project - NLP Product Knowledge Graph</b> .....	4
<b>DSCI 575: Advanced Machine Learning</b> .....	4
<b>DSCI 574: Spatial and Temporal Models</b> .....	5
<b>DSCI 572: Supervised Learning II</b> .....	5
<b>DSCI 563: Unsupervised Learning</b> .....	6
<b>DSCI 562: Regression II</b> .....	6
<b>DSCI 554: Experimentation and Causal Inference</b> .....	7
<b>DSCI 553: Statistical Inference and Computation II</b> .....	7
<b>DSCI 542: Communication and Argumentation</b> .....	7
<b>DSCI 541: Privacy, Ethics, and Security</b> .....	8
<b>DSCI 532: Data Visualization II</b> .....	8
<b>DSCI 525: Web and Cloud Computing</b> .....	8
<b>DSCI 524: Collaborative Software Development</b> .....	8
<b>DSCI 573: Feature and Model Selection</b> .....	9
<b>DSCI 571: Supervised Learning I</b> .....	9
<b>DSCI 561: Regression I</b> .....	10
<b>DSCI 552: Statistical Inference and Computation I</b> .....	10
<b>DSCI 551: Descriptive Statistics and Probability for Data Science</b> .....	10
<b>DSCI 531: Data Visualization I</b> .....	10
<b>DSCI 523: Programming for Data Manipulation</b> .....	11

<b>DSCI 522: Data Science Workflows</b> .....	<b>11</b>
<b>DSCI 521: Computing Platforms for Data Science</b> .....	<b>11</b>
<b>DSCI 513: Databases and Data Retrieval</b> .....	<b>11</b>
<b>DSCI 512: Algorithms and Data Structures</b> .....	<b>12</b>
<b>DSCI 511: Programming for Data Science</b> .....	<b>12</b>
<i>Undergraduate Courses: UCLA &amp; UCSD</i> .....	<b>12</b>
<b>ECON 171: Decisions Under Uncertainty</b> .....	<b>12</b>
<b>ECON 147: Economics of Education</b> .....	<b>13</b>
<b>ECON 135 / USP 102: Urban Economics</b> .....	<b>13</b>
<b>ECON 130: Public Policy</b> .....	<b>13</b>
<b>ECON 120C: Econometrics C</b> .....	<b>13</b>
<b>ECON 120B: Econometrics B</b> .....	<b>13</b>
<b>ECON 120A: Econometrics A</b> .....	<b>14</b>
<b>ECON 110B: Macroeconomics B</b> .....	<b>14</b>
<b>ECON 102: Globalization</b> .....	<b>14</b>
<b>ECON 101: Microeconomic Theory</b> .....	<b>14</b>
<b>ECON 100B: Microeconomics B</b> .....	<b>14</b>
<b>ECON 100A: Microeconomics A</b> .....	<b>15</b>
<b>ECON 3: Economics - Macro</b> .....	<b>15</b>
<b>ECON 1: Economics - Micro</b> .....	<b>15</b>
<b>MATH 181A: Introduction to Mathematical Statistics I</b> .....	<b>16</b>
<b>MATH 180A: Introduction to Probability</b> .....	<b>16</b>
<b>MATH 174E: Mathematics of Finance</b> .....	<b>16</b>
<b>MATH 170B: Probability Theory</b> .....	<b>17</b>
<b>MATH 170A: Introduction to Numerical Analysis</b> .....	<b>17</b>
<b>MATH 164: Optimization</b> .....	<b>17</b>
<b>MATH 142B: Introduction to Analysis II</b> .....	<b>18</b>
<b>MATH 142A: Introduction to Analysis I</b> .....	<b>18</b>
<b>MATH 135: Ordinary Differential Equations</b> .....	<b>18</b>

<b>MATH 115A: Linear Algebra.....</b>	<b>19</b>
<b>MATH 109: Mathematical Reasoning .....</b>	<b>19</b>
<b>MATH 61: Discrete Structures.....</b>	<b>19</b>
<b>MATH 20F: Linear Algebra .....</b>	<b>20</b>
<b>MATH 20E: Vector Calculus.....</b>	<b>20</b>
<b>MATH 20D: Introduction to Differential Equations.....</b>	<b>20</b>
<b>MATH 20C: Calculus &amp; Analytic Geometry for Science and Engineering.</b>	<b>21</b>
<b>MATH 20B: Calculus for Science and Engineering .....</b>	<b>21</b>
<b>MATH 20A: Calculus .....</b>	<b>21</b>
<b>COMPTNG 10A: Introduction to Programming .....</b>	<b>21</b>
<b>CSE 3: Fluency in Information Technology .....</b>	<b>22</b>
<b>GEOG 7: Introduction to Geographic Information Systems (GIS).....</b>	<b>22</b>
<b>PSYC 60: Introduction to Statistics in Psychology .....</b>	<b>22</b>

## Graduate Courses: UBC

Course Title	Textbooks/References	Course Materials	Grades
--------------	----------------------	------------------	--------

<p><b>DSCI 591:</b> Capstone Project - NLP Product Knowledge Graph - UBC</p>	<p>Aumüller, Martin, Erik Bernhardsson, and Alexander Faithfull. 2018. “ANN-Benchmarks: A Benchmarking Tool for Approximate Nearest Neighbor Algorithms.” <a href="https://arxiv.org/abs/1807.05614">https://arxiv.org/abs/1807.05614</a>.</p> <p>Bernhardsson, Erik. 2017. “ANNOY Spotify Repository.” GitHub repository. <a href="https://github.com/spotify/annoy">https://github.com/spotify/annoy</a>.</p>	<ul style="list-style-type: none"> <li>• Named-entity Recognition</li> <li>• LLMs Fine-tuning (BERT, RoBERTa and T5)</li> <li>• Data Annotation</li> <li>• Inter-annotator Agreement</li> <li>• ANNOY Model for Similarity Search</li> <li>• Similarity Algorithms</li> <li>• neo4j Graph Local Database Deployment</li> <li>• Dashboard Deployment on Render</li> <li>• AWS Neptune Graph Database Deployment</li> </ul>	A+
<p><b>DSCI 575:</b> Advanced Machine Learning - UBC</p>	<p>Bird, S., Klein, E., &amp; Loper, E. (2009). <i>Natural Language Processing with Python</i>. O'Reilly Media, Inc.</p> <p>Eisenstein, J. (2018). <i>Natural Language Processing</i>. MIT Press.</p> <p>Goldberg, Y. (2017). <i>Neural Network Methods for Natural Language Processing</i> (Vol. 37). San Rafael, CA: Morgan &amp; Claypool.</p> <p>Goodfellow, I., Bengio, Y., &amp; Courville, A. (2016). <i>Deep Learning</i>. MIT Press.</p> <p>Jurafsky, D. and Martin, J.H. (2019) <i>Speech and Language Processing</i>. Pearson Education International.</p>	<ul style="list-style-type: none"> <li>• Markov Models (n-gram)</li> <li>• PageRank Algorithm</li> <li>• spaCy &amp; nltk NLP libraries</li> <li>• Hidden Markov Models</li> <li>• Viterbi Algorithm</li> <li>• Automatic Speech Recognition (ASR) System</li> <li>• Noisy Channel Model</li> <li>• Part-of-speech Tag Sequence</li> <li>• Baum-Welch Algorithm</li> <li>• Advanced Topic Modeling</li> <li>• Latent Dirichlet Allocation</li> <li>• <a href="#">Topic2Vec</a> Algorithm</li> <li>• <a href="#">BERTopic</a> Algorithm</li> <li>• Recurrent Neural Networks (Stacked and Bidirectional)</li> <li>• Self-attention Networks &amp; Positional Encodings</li> <li>• Advanced Transformers Block Architectures</li> <li>• GPT Models</li> <li>• Bidirectional Transformer Models (BERT, RoBERTa)</li> <li>• Next Sentence Prediction</li> <li>• Transfer Learning via Fine-Tuning</li> <li>• Sentiment Analysis</li> <li>• Zero-shot Classification via NLI Models</li> </ul>	A+

<p><b>DSCI 574:</b> Spatial and Temporal Models - UBC</p>	<p>Hyndman, Rob J. &amp; Athanasopoulos, George. &amp; OTexts.com, issuing body. (2021). <i>Forecasting: principles and practice</i>.</p> <p>Hyndman, R. J., Koehler, A. B., Ord, J. K., &amp; Snyder, R. D. (2008). <i>Forecasting with Exponential Smoothing: The State Space Approach</i>. Springer.</p> <p>John., R., Sergio. Arribas-Bel, Dani. Wolf, Levi. (2023). <i>Geographic Data Science With Python</i>. Routledge.</p> <p>Pebesma, E.; Bivand, R. (2023). <i>Spatial Data Science: With Applications in R</i>. Chapman and Hall/CRC.</p> <p>Shumway, R. H., Stoffer, D. S. (2017). <i>Time Series Analysis and Its Applications</i>. Springer.</p>	<ul style="list-style-type: none"> <li>• Time Series Seasonal Trend-cycle Decomposition</li> <li>• STL Decomposition</li> <li>• Holt-Winter's Model</li> <li>• Exponential Smoothing Algorithms</li> <li>• ETS (Error, Trend, Seasonal) Models</li> <li>• Correlograms ACF &amp; PACF</li> <li>• SARIMAX Models</li> <li>• Forecasting with Machine Learning Models</li> <li>• Multivariate Time Series</li> <li>• Quantile Regression</li> <li>• Anomaly Detection</li> <li>• NA Imputation</li> <li>• Time Series with Deep Learning (2D CNNs, RNN, LSTM)</li> <li>• Spatial Analysis with Vector and Raster Data</li> <li>• Spatial Modelling with Spatial and Areal Interpolation</li> <li>• Dijkstra Algorithm</li> </ul>	A+
<p><b>DSCI 572:</b> Supervised Learning II - UBC</p>	<p>Bishop, Christopher M. (2006). <i>Pattern recognition and machine learning</i>. New York: Springer.</p> <p>Goodfellow, I., Bengio, Y., &amp; Courville, A. (2016). <i>Deep learning</i>. MIT Press.</p> <p>Poole, D., Mackworth, A. (2023). <i>Artificial Intelligence: Foundations of Computational Agents</i>. Cambridge, UK: Cambridge University Press.</p> <p>Russell, S., Norvig, P. (2020). <i>Artificial Intelligence: A Modern Approach</i>. Prentice Hall.</p>	<ul style="list-style-type: none"> <li>• Floating-point Numbers</li> <li>• Overflow &amp; Underflow Errors</li> <li>• Loss Functions</li> <li>• Optimization and Gradient Descent Algorithms</li> <li>• Stochastic Gradient Descent and Mini-batch Gradient Descent Algorithms</li> <li>• Neural Networks</li> <li>• Convolutional Neural Networks (CNNs)</li> <li>• Hyperparameter Tuning</li> <li>• Transfer Learning</li> <li>• Siamese Networks for Few-Shot Learning</li> <li>• Generative Adversarial Networks (GANs)</li> </ul>	A+

<p><b>DSCI 563:</b> Unsupervised Learning - UBC</p>	<p>Abu-Mostafa, Y. S., Magdon-Ismail, M., &amp; Lin, H. (2012). <i>Learning from data: a short course</i>. [United States] :MLBook.com</p> <p>Daume, H. III (2017) <i>A Course in Machine Learning</i>. The Perceptron.</p> <p>Russell, S., Norvig, P. (2020). <i>Artificial Intelligence: A Modern Approach</i>. Prentice Hall.</p>	<ul style="list-style-type: none"> <li>• K-Means Clustering</li> <li>• Gaussian Mixture Models</li> <li>• DBSCAN Clustering</li> <li>• Hierarchical Clustering</li> <li>• Principal Component Analysis (PCA)</li> <li>• Latent Semantic Analysis (LSA)</li> <li>• Non-negative Matrix Factorization (NMF)</li> <li>• Term-term Co-occurrence Matrix</li> <li>• word2vec Skip-gram Model</li> <li>• fastText algorithm</li> <li>• GloVe algorithm</li> <li>• t-SNE for Manifold Learning</li> <li>• Collaborative Filtering</li> <li>• Content-based Filtering</li> </ul>	A+
<p><b>DSCI 562:</b> Regression II - UBC</p>	<p>Fahrmeir, L. (2013). <i>Regression Models, Methods and Applications</i>. Springer Berlin Heidelberg.</p> <p>Faraway, Julian J. (2005). <i>Extending the Linear Model with R: Generalized Linear, Mixed Effects and Nonparametric Regression Models</i>, CRC Press LLC.</p> <p>Gelman, A. and Hill, J. (2007). <i>Data Analysis Using Regression and Multilevel/Hierarchical Models</i>. Analytical Methods for Social Research. Cambridge University Press.</p> <p>Kleinbaum, D. G. and Klein, M. (2005). <i>Survival analysis: A Self-Learning Text</i>. Springer.</p> <p>Rubin, D. B. (1987). <i>Multiple Imputation for Nonresponse in Surveys</i>. Wiley.</p>	<ul style="list-style-type: none"> <li>• Binary Logistic Regression for GLMs</li> <li>• Analysis of Deviance, AIC, and BIC</li> <li>• Poisson Regression</li> <li>• Quasi-Poisson Regression</li> <li>• Negative Binomial Regression</li> <li>• Multinomial Logistic Regression</li> <li>• Ordinal Logistic Regression</li> <li>• Linear Mixed-Effects Models</li> <li>• Kaplan-Meier Non-parametric Estimation for Survival Analysis</li> <li>• Parametric Estimation for Survival Analysis</li> <li>• Cox Proportional Hazards Model</li> <li>• Data Imputation for Various Missing Data</li> <li>• Quantile Regression</li> <li>• LOWESS Regression</li> <li>• Robust Regression</li> </ul>	A+

<p><b>DSCI 554:</b> Experimentation and Causal Inference - UBC</p>	<p>Cetinkaya-Rundel, M., Diez, D., &amp; Barr, C. (2019). <i>OpenIntro Statistics</i>. (Fourth Edition ed.) OpenIntro, Inc.</p> <p>Oehlert, G.W. (2010). <i>A First Course in Design and Analysis of Experiments</i>. Library of Congress Cataloging-in-Publication Data, USA.</p> <p>Seltman, H. J. (2015). <i>Experimental Design and Analysis</i>. Pittsburgh: Carnegie Mellon University.</p>	<ul style="list-style-type: none"> <li>• Bonferroni Correction</li> <li>• False Discovery Rate</li> <li>• Simpson's Paradox</li> <li>• Randomization</li> <li>• A/B/n Testing</li> <li>• Blocked Designs</li> <li>• Power Analysis</li> <li>• Early Stopping and Peeking</li> <li>• Confounding</li> <li>• Stratified Analysis for Causality</li> <li>• Sampling Schemes (Case-control, Cross-sectional and Cohort)</li> <li>• Simulations</li> </ul>	A+
<p><b>DSCI 553:</b> Statistical Inference and Computation II - UBC</p>	<p>Downey, A. B. (2022). <i>Think Bayes</i>. Sebastopol, California: O'Reilly Media.</p> <p>Johnson, A.A., Ott, M.Q., &amp; Dogucu, M. (2022). <i>Bayes Rules!: An Introduction to Applied Bayesian Modeling</i> (1st ed.). Chapman and Hall/CRC.</p>	<ul style="list-style-type: none"> <li>• Bayesian Statistics</li> <li>• Maximum a Posteriori Estimation (MAP)</li> <li>• Beta-Binomial Model</li> <li>• Gamma-Poisson Model</li> <li>• Normal-Normal Model</li> <li>• Markov Chain Monte Carlo (MCMC) Algorithm</li> <li>• Metropolis-Hastings Algorithm</li> <li>• Bayesian Normal Linear Regression</li> <li>• Bayesian Binary Logistic Regression</li> <li>• Bayesian Hierarchical Models</li> <li>• Gelman-Rubin Diagnostic</li> </ul>	A+
<p><b>DSCI 542:</b> Communication and Argumentation: - UBC</p>	<p>Levitin, D. J. (2020). <i>A Field Guide to Lies: Critical Thinking with Statistics and the Scientific Method</i>. Dutton.</p> <p>Savage, S. A. (2012). <i>The Flaw of Averages: Why We Underestimate Risk in the Face of Uncertainty</i>. John Wiley &amp; Sons.</p>	<ul style="list-style-type: none"> <li>• P-hacking</li> <li>• Correlation and Causation</li> <li>• ML Fairness</li> <li>• Inverse Fallacy</li> <li>• Explanation with Analogy</li> <li>• Writing Technical Reports</li> </ul>	A+

<b>DSCI 541:</b> Privacy, Ethics, and Security - UBC	<p>DiResta, R. 2018, April 13. <i>How Do We Know What's True Anymore</i> [Video]. YouTube.  <a href="https://www.youtube.com/watch?v=vIkJoUeoY-o">https://www.youtube.com/watch?v=vIkJoUeoY-o</a></p> <p>Murgia, M. 2017, May 23. <i>How data brokers sold my identity</i> [Video]. YouTube.  <a href="https://www.youtube.com/watch?v=AU66C6HePfg&amp;t=433s">https://www.youtube.com/watch?v=AU66C6HePfg&amp;t=433s</a></p>	<ul style="list-style-type: none"> <li>• Disinformation</li> <li>• Goodhart's/Campbell's Law</li> <li>• Deepfakes</li> <li>• Metadata</li> <li>• Data K-anonymization</li> <li>• L-diversity</li> <li>• Differential Privacy</li> <li>• Types of Bias in DS</li> <li>• Confusion Matrix &amp; Fairness Metrics</li> </ul>	A+
<b>DSCI 532:</b> Data Visualization II - UBC	Wickham, H. (2021). <i>Mastering Shiny</i> . O'Reilly Media, Inc.	<ul style="list-style-type: none"> <li>• Shiny App Development</li> <li>• Dash App Development</li> <li>• Interactive Visualization</li> </ul>	A+
<b>DSCI 525:</b> Web and Cloud Computing - UBC	Amazon Web Services. (2023). AWS Documentation. Amazon Web Services. <a href="https://docs.aws.amazon.com/index.html">https://docs.aws.amazon.com/index.html</a>	<ul style="list-style-type: none"> <li>• Big Data</li> <li>• Parquet Data</li> <li>• Serialization/Deserialization</li> <li>• Apache Arrow</li> <li>• DuckDB, Polars and Ibis</li> <li>• Amazon VPC</li> <li>• Amazon EC2 Instance</li> <li>• AWS Lambda</li> <li>• Amazon S3 Bucket</li> <li>• Amazon RDS</li> <li>• Amazon DynamoDB</li> <li>• Hadoop Framework</li> <li>• MapReduce and Spark</li> <li>• Amazon EMR</li> </ul>	A+
<b>DSCI 524:</b> Collaborative Software Development - UBC	<p>Beuzen, T., Timbers, T. (2022). <i>Python Packages</i>. Chapman &amp; Hall/CRC.</p> <p>Wickham, H., Bryan, J. (2023). <i>R Packages</i>. O'Reilly Media, Inc.</p>	<ul style="list-style-type: none"> <li>• Cookiecutter and Poetry</li> <li>• Git Flow</li> <li>• pytest &amp; testthat for Unit Testing, Integration Testing and Regression Testing</li> <li>• Code Coverage Metrics</li> <li>• Continuous Integration (CI)</li> <li>• Continuous Development (CD)</li> <li>• GitHub Actions</li> <li>• Deployment on CRAN and PyPI</li> </ul>	A+

DSCI 573: Feature and Model Selection - UBC	<p>Bishop, Christopher M. (2006). <i>Pattern recognition and machine learning</i>. New York: Springer.</p> <p>Goodfellow, I., Bengio, Y., &amp; Courville, A. (2016). <i>Deep learning</i>. MIT Press.</p> <p>Poole, D., Mackworth, A. (2023). <i>Artificial Intelligence: Foundations of Computational Agents</i>. Cambridge, UK: Cambridge University Press.</p> <p>Russell, S., Norvig, P. (2020). <i>Artificial Intelligence: A Modern Approach</i>. Prentice Hall.</p>	<ul style="list-style-type: none"> <li>• Confusion Matrix</li> <li>• Precision-recall (PR) Curve and Receiver Operating Characteristic (ROC) Curve</li> <li>• Class Imbalance</li> <li>• Hyperparameter Tuning</li> <li>• MSE, RMSE, MAPE, R^2</li> <li>• Linear and Non-linear Feature Engineering</li> <li>• Radial Basis Function(RBF)</li> <li>• nltk and spaCy</li> <li>• TF-IDF Term Weighting</li> <li>• Part-of-speech (POS)</li> <li>• Feature Importance</li> <li>• Recursive Feature Elimination Algorithm</li> <li>• Loss Functions</li> <li>• L0, L1, L2 Regularization</li> <li>• Sparsity</li> <li>• Collinearity</li> <li>• Ensembles</li> <li>• Random Forests (RF)</li> <li>• Gradient Boosted Trees</li> <li>• eli5</li> <li>• SHapley Additive exPlanations (SHAP)</li> </ul>	A+
DSCI 571: Supervised Learning I - UBC	<p>Bishop, Christopher M. (2006). <i>Pattern recognition and machine learning</i>. New York: Springer.</p> <p>Goodfellow, I., Bengio, Y., &amp; Courville, A. (2016). <i>Deep learning</i>. MIT Press.</p> <p>Poole, D., Mackworth, A. (2023). <i>Artificial Intelligence: Foundations of Computational Agents</i>. Cambridge, UK: Cambridge University Press.</p> <p>Russell, S., Norvig, P. (2020). <i>Artificial Intelligence: A Modern Approach</i>. Prentice Hall.</p>	<ul style="list-style-type: none"> <li>• Decision Trees</li> <li>• Cross Validations</li> <li>• Overfitting and Underfitting</li> <li>• k-Nearest Neighbors (KNN)</li> <li>• Support Vector Machines (SVMs)</li> <li>• Imputation and Scaling</li> <li>• One-hot encoding (OHE)</li> <li>• Bag of Words (BOW)</li> <li>• TF-IDF</li> <li>• Hyperparameter Optimization</li> <li>• Optimization Bias</li> <li>• Naive Bayes</li> <li>• Laplace Smoothing</li> <li>• Gaussian Naive Bayes</li> <li>• Linear Regression</li> <li>• Logistic Regression</li> </ul>	A+

<b>DSCI 561:</b> Regression I - UBC	Faraway, J.J. (2014). <i>Linear Models with R</i> . Chapman and Hall/CRC.  Irizarry, R. (2019). <i>Introduction to Data Science</i> . CRC Press.	<ul style="list-style-type: none"> <li>• Linear Regression</li> <li>• F-test for nested models</li> <li>• Evaluation Metrics</li> <li>• Prediction Intervals VS Confidence Intervals</li> <li>• Multicollinearity</li> <li>• Confounding</li> </ul>	A+
<b>DSCI 552:</b> Statistical Inference and Computation I - UBC	Ismay, C., & Kim, A.Y. (2019). <i>Statistical Inference via Data Science: A ModernDive into R and the Tidyverse</i> . Chapman and Hall/CRC.	<ul style="list-style-type: none"> <li>• Sampling Distributions</li> <li>• Bootstrapping</li> <li>• Simulation</li> <li>• Randomization</li> <li>• Central Limit Theorem</li> <li>• Normal Distributions</li> <li>• t-Distributions</li> <li>• Hypothesis Testing</li> <li>• Pearson's Chi-squared Test</li> <li>• Type I and II Errors</li> <li>• p-value</li> <li>• Analysis of Variance (ANOVA)</li> <li>• Maximum Likelihood Estimation</li> </ul>	A+
<b>DSCI 551:</b> Descriptive Statistics and Probability for Data Science - UBC	Blitzstein, J.K., & Hwang, J. (2019). <i>Introduction to Probability</i> . Chapman and Hall/CRC.  Tsun, A. (2020). <i>Probability &amp; Statistics with Applications to Computing</i> .	<ul style="list-style-type: none"> <li>• Discrete Distributions</li> <li>• Joint Probability</li> <li>• Independence</li> <li>• Conditional Probability</li> <li>• Continuous Distributions</li> <li>• Contour Plots</li> <li>• Marginal Distributions</li> <li>• Simulation</li> </ul>	A+
<b>DSCI 531:</b> Data Visualization I - UBC	Munzner, T. (2014). <i>Visualization Analysis and Design</i> . CRC Press.  Wilke, C. (2019). <i>Fundamentals of Data Visualization</i> . O'Reilly Media Inc.	<ul style="list-style-type: none"> <li>• ggplot2</li> <li>• Altair</li> <li>• Exploratory Data Analysis</li> <li>• Color Theory</li> <li>• Interactive Plots</li> <li>• Pairwise Comparisons</li> </ul>	A+

<b>DSCI 523:</b> Programming for Data Manipulation - UBC	Irizarry, R. (2019). <i>Introduction to Data Science</i> . CRC Press.	<ul style="list-style-type: none"> <li>• R Programming</li> <li>• Data Wrangling</li> <li>• Packages (tidyverse, gapminder, purr and so on)</li> <li>• Factor Data</li> <li>• roxygen2 Documentation</li> <li>• Anonymous Functions in R</li> <li>• Mapping</li> <li>• Nested Data Frames</li> <li>• Tidy Evaluation</li> </ul>	A+
<b>DSCI 522: Data Science Workflows</b> - UBC	Irizarry, R. (2019). <i>Introduction to Data Science</i> . CRC Press.  Peng, R., Matsui, E. (2017). <i>The Art of Data Science</i> .	<ul style="list-style-type: none"> <li>• Read-eval-print-loop (REPL) Framework</li> <li>• Shell Scripts</li> <li>• Advanced Reporting in R Markdown</li> <li>• Makefile for Pipelines</li> <li>• Docker</li> </ul>	A+
<b>DSCI 521:</b> Computing Platforms for Data Science - UBC	Adhikari, A., DeNero, J., Wagner, D. (2022). <i>Computational and Inferential Thinking: The Foundations of Data Science</i> .	<ul style="list-style-type: none"> <li>• LaTex</li> <li>• Bash Shell</li> <li>• Wildcards</li> <li>• Regular Expressions</li> <li>• Git and GitHub for Version Control</li> <li>• Merge Conflicts</li> <li>• Virtual Environments</li> <li>• Quarto Documents</li> </ul>	A+
<b>DSCI 513:</b> Databases and Data Retrieval - UBC	Karwin, B. (2010). <i>SQL antipatterns: Avoiding the Pitfalls of Database Programming</i> . The Pragmatic Programmers, LLC.  Ramakrishnan, R., & Gehrke, J. (2002). <i>Database Management Systems</i> . McGraw-Hill.  Celko, J. (2011). <i>Joe celko's SQL for Smarties: Advanced SQL programming</i> . Morgan Kaufmann.	<ul style="list-style-type: none"> <li>• Relational Databases</li> <li>• SQL</li> <li>• Data Retrieval</li> <li>• Data Manipulation</li> <li>• ACID Properties</li> <li>• Sub-queries</li> <li>• Non-relational Databases</li> <li>• MongoDB</li> <li>• Advanced Queries</li> </ul>	A+

<b>DSCI 512:</b> <b>Algorithms and</b> <b>Data Structures</b> - UBC	Dasgupta, S., Papadimitriou, C., Vazirani, U. (2006). <i>Algorithms</i> . McGraw Hill.  Goodrich, M. T., Tamassia, R., Goldwasser, M. H. (2013). <i>Data Structures and Algorithms in Python</i> . Wiley.	<ul style="list-style-type: none"> <li>• Big O Time and Space Complexity</li> <li>• Binary Search Trees</li> <li>• Sorting Algorithms</li> <li>• Hash Tables &amp; Functions</li> <li>• Recursion &amp; Recursive Data Structures</li> <li>• Nearest Neighbors</li> <li>• k-d Trees</li> <li>• Amortization</li> <li>• Graphs</li> <li>• Graph Search: Recursive &amp; Iterative Implementation</li> <li>• Stacks and Queues</li> <li>• Centrality and PageRank</li> <li>• Matrices</li> <li>• Sparsity</li> <li>• Discrete &amp; Continuous Optimization</li> <li>• Linear Programming</li> <li>• Caching &amp; Memorization</li> <li>• Dynamic Programming</li> <li>• Vectorization</li> <li>• Backtracking</li> <li>• Linked List</li> </ul>	A+
<b>DSCI 511:</b> <b>Programming for</b> <b>Data Science</b> -UBC	McKinney, W. (2012). <i>Python for Data Analysis</i> . O'Reilly Media.	<ul style="list-style-type: none"> <li>• Loops &amp; Functions</li> <li>• Unit Tests &amp; Classes</li> <li>• NumPy</li> <li>• Pandas</li> <li>• Advanced Data Wrangling</li> </ul>	A+

## Undergraduate Courses: UCLA & UCSD

<b>ECON 171:</b> <b>Decisions Under</b> <b>Uncertainty</b> - UCSD	Lindley, D. V. (1991). <i>Making Decisions</i> . Wiley.	<ul style="list-style-type: none"> <li>• Game Theories</li> <li>• Decision Theories</li> <li>• Conditional Probability and Bayes' Law</li> <li>• Expected Utility Theories</li> <li>• Theory of Stochastic Dominance</li> <li>• Comparative Statics of Risk</li> <li>• Prospect Theory</li> <li>• Subjective Uncertainty</li> </ul>	A
--	---	---	---

<b>ECON 147:</b> Economics of Education - UCSD	Assigned Journal Publications, Articles and Readings by Professor Julian Betts.	<ul style="list-style-type: none"> <li>• Correlation &amp; Causality</li> <li>• Omitted Variable Bias and Endogenous Regressors</li> <li>• Regression Discontinuity</li> <li>• Instrumental Variables</li> <li>• Difference-in-Difference Models</li> <li>• Human Capital Theory</li> <li>• Signaling Theory</li> <li>• School Choice</li> <li>• School Finance</li> <li>• Postsecondary Education</li> </ul>	A+
<b>ECON 135 /</b> USP 102: Urban Economics - UCSD	O'Sullivan, A. (2009). <i>Urban Economics</i> . McGraw Hill.	<ul style="list-style-type: none"> <li>• Suburbanization</li> <li>• Zoning</li> <li>• Urban Transportation</li> <li>• Segregation</li> <li>• Local Government</li> </ul>	A-
<b>ECON 130:</b> Public Policy - UCSD	Miller, R. (2017). <i>The Economics of Public Issues</i> . Pearson.	<ul style="list-style-type: none"> <li>• Pareto Efficiency</li> <li>• Taxes and Subsidies</li> <li>• Coase Theorem</li> <li>• Market Failure</li> <li>• Monopoly</li> <li>• Pigouvian Taxes</li> <li>• Tradeable Permits</li> <li>• Imperfect Information</li> </ul>	A+
<b>ECON 120C:</b> Econometrics C - UCSD (with Stata coding)	Stock, J., Watson, M. (2007) <i>Introduction to Econometrics</i> . Pearson.	<ul style="list-style-type: none"> <li>• Discrete Choice Model</li> <li>• Instrumental Variable Regression</li> <li>• Panel Data</li> <li>• Experiments and Quasi-Experiments</li> <li>• Time Series Data</li> <li>• Simultaneous Equations</li> <li>• Non-parametric Statistics</li> </ul>	A+
<b>ECON 120B:</b> Econometrics B - UCSD (with Stata coding)	Stock, J., Watson, M. (2007) <i>Introduction to Econometrics</i> . Pearson.	<ul style="list-style-type: none"> <li>• Ordinary Least Squares</li> <li>• Confidence Interval and Hypothesis Testing</li> <li>• Dummy Variables</li> <li>• Multiple Regression</li> <li>• Simultaneity</li> <li>• Sample Selection</li> <li>• Measurement Error</li> <li>• Non-linear Models</li> </ul>	A+

<b>ECON 120A:</b> <b>Econometrics A</b> - UCSD	Stock, J., Watson, M. (2007) <i>Introduction to Econometrics.</i> Pearson.	<ul style="list-style-type: none"> <li>• Descriptive Statistics</li> <li>• Probability Distributions</li> <li>• Sampling</li> <li>• Point Estimation</li> <li>• Interval Estimation</li> <li>• Hypothesis Tests</li> </ul>	A+
<b>ECON 110B:</b> <b>Macroeconomics B</b> - UCSD	Jones, C. I. (2014). <i>Macroeconomics.</i>	<ul style="list-style-type: none"> <li>• IS Curve</li> <li>• Monetary Policy &amp; Philips Curve</li> <li>• AS/AD Model</li> <li>• The Great Recession</li> <li>• Short-Run Model</li> <li>• Exchange Rate and International Trading</li> </ul>	A
<b>ECON 102:</b> <b>Globalization</b> - UCSD	Assigned Journal Publications, Articles, and readings by Professor Marc-Andreas Muendler.	<ul style="list-style-type: none"> <li>• Comparative Advantage</li> <li>• Ricardo Model and HOS Model</li> <li>• Distribution of Gains</li> <li>• Heckscher-Ohlin Model</li> <li>• Production Possibility Frontier</li> <li>• Offshoring</li> <li>• Fischer model of International Capital Flows</li> </ul>	A+
<b>ECON 101:</b> <b>Microeconomic Theory</b> - UCLA	Nicholson, W., Snyder, C. (2016) <i>Microeconomic Theory: Basic Principles and Extensions.</i> South-Western College.	<ul style="list-style-type: none"> <li>• Game Theories</li> <li>• Mixed Strategies and Sequential Games</li> <li>• Monopoly</li> <li>• Oligopoly</li> <li>• Economics of Information</li> </ul>	A+
<b>ECON 100B:</b> <b>Microeconomics B</b> - UCSD	Perloff, J. (2016). <i>Microeconomics: Theory and Applications with Calculus.</i> Pearson.	<ul style="list-style-type: none"> <li>• Production Functions</li> <li>• Law of Diminishing Marginal Returns</li> <li>• Isoquants</li> <li>• Costs</li> <li>• Profit Maximization</li> <li>• Competitive Model</li> <li>• General Equilibrium</li> </ul>	A+

ECON 100A: Microeconomics <b>A</b> - UCSD	Perloff, J. (2016). <i>Microeconomics: Theory and Applications with Calculus.</i> Pearson.	<ul style="list-style-type: none"> <li>• Level Curves and Utility Functions</li> <li>• Optimization and Comparative Statics of Demand</li> <li>• Slutsky Equation and Elasticity</li> <li>• Decision Theories for Supply of Labor and Saving</li> </ul>	A
ECON 3: Economics - <b>Macro</b> - Advanced Placement Credit - UCSD	Bernanke, B., Frank, R., Antonovics, K., Heffetz, O. (2015). <i>Principles of Economics.</i> McGraw-Hill Irwin.	<ul style="list-style-type: none"> <li>• Long Run Growth</li> <li>• Savings &amp; Investment</li> <li>• Money &amp; Banking System</li> <li>• Short Run Business Cycles</li> <li>• Macroeconomic Models</li> <li>• Fiscal &amp; Monetary Policy</li> <li>• Exchange Rates</li> </ul>	AP 5
ECON 1: Economics - <b>Micro</b> - Advanced Placement Credit - UCSD	Bernanke, B., Frank, R., Antonovics, K., Heffetz, O. (2015). <i>Principles of Economics.</i> McGraw-Hill Irwin.	<ul style="list-style-type: none"> <li>• Supply &amp; Demand Curves</li> <li>• Elasticity</li> <li>• Production, Cost &amp; Profit</li> <li>• Perfect Competition</li> <li>• Monopoly</li> </ul>	AP 5

<b>MATH 181A:</b> Introduction to Mathematical Statistics I - UCSD (with statistical R coding)	Rice, J. (2006). <i>Mathematical Statistics and Data Analysis</i> . Duxbury Press.  Wackerly, D., Mendenhall, W., Scheaffer, R. (2007). <i>Mathematical Statistics with Applications</i> . Duxbury Press.	<ul style="list-style-type: none"> <li>• Method of Moments Estimator (MME)</li> <li>• Maximum Likelihood Estimator (MLE)</li> <li>• Interval Estimation</li> <li>• Unbiasedness &amp; Efficiency</li> <li>• Monte Carlo Simulation</li> <li>• Parametric Bootstrap Algorithm</li> <li>• Cramer-Rao Lower Bound</li> <li>• Fisher Information</li> <li>• Convergence in Probability</li> <li>• Asymptotic Normality</li> <li>• Hypothesis Test &amp; p-value</li> <li>• Exact Test &amp; t-test</li> <li>• Type I and II errors</li> <li>• Wald Testing</li> <li>• Likelihood Ratio Test</li> <li>• Neyman-Pearson Lemma</li> <li>• Uniformly Most Powerful Test</li> <li>• Normal Distributions</li> <li>• Bayesian Inference</li> </ul>	A
<b>MATH 180A:</b> Introduction to Probability - UCSD (with Python coding)	Adhikari, A. and Pitman, J. (2016). <i>Probability for Data Science</i> .  Anderson, D., Seppäläinen, T. and Valkó, B. (2017). <i>Introduction to Probability</i> . Cambridge University Press.	<ul style="list-style-type: none"> <li>• Conditional Probability &amp; Bayes' Rule</li> <li>• Independence</li> <li>• Probability Distributions</li> <li>• Sampling</li> <li>• Normal and Poisson Approximation</li> <li>• Poisson Process</li> <li>• Moment Generating Function</li> <li>• Joint Distributions</li> <li>• Law of Large Numbers</li> <li>• Central Limit Theorem</li> </ul>	A+
<b>MATH 174E:</b> Mathematics of Finance - UCLA	Hull, J. C. (2017). <i>Options, Futures and Other Derivatives</i> . Pearson.	<ul style="list-style-type: none"> <li>• Options &amp; Futures Markets</li> <li>• Binomial Trees</li> <li>• Wiener Processes</li> <li>• Ito's lemma</li> <li>• Black-Scholes-Merton Model</li> <li>• Numerical Procedures</li> <li>• Volatility &amp; Value at Risk</li> </ul>	B

<b>MATH 170B:</b> Probability Theory - UCLA	<p>Durrett, R. (2010). <i>Probability: Theory and Examples</i>. Cambridge University Press.</p> <p>Grimmett, G. R., &amp; Welsh, D. J. (2014). <i>Probability: An Introduction</i>. Oxford: Oxford University Press.</p>	<ul style="list-style-type: none"> <li>• Multinomial Distributions</li> <li>• Hypergeometric Distribution</li> <li>• Negative Binomial Distribution</li> <li>• Conditional Distribution</li> <li>• Convergence in Distribution</li> <li>• Central Limit Theorem</li> <li>• Markov and Chebyshev Inequalities</li> <li>• Convergence in Probability</li> <li>• Weak and Strong Law of Large Numbers</li> <li>• Almost-sure Convergence</li> <li>• Bernoulli Process</li> <li>• Poisson Process</li> <li>• Random Walks</li> </ul>	A-
<b>MATH 170A:</b> Introduction to Numerical Analysis - UCSD (with MATLAB coding)	<p>Watkins, D. S. (2010). <i>Fundamentals of Matrix Computation</i>. Wiley.</p>	<ul style="list-style-type: none"> <li>• Floating-Point Numbers</li> <li>• Linear Systems of Equations</li> <li>• Triangular Systems of Equations</li> <li>• LU Decomposition</li> <li>• Gauss Elimination</li> <li>• Cholesky Factorization</li> <li>• Sensitivity of Linear Systems</li> <li>• Gram-Schmidt Process</li> <li>• QR Factorization</li> <li>• Iterative Methods</li> </ul>	A+
<b>MATH 164:</b> Optimization - UCLA	<p>Chong, E. K.P. and Zak, S. H. (2013). <i>An Introduction to Optimization</i>. Wiley.</p>	<ul style="list-style-type: none"> <li>• Convex Geometry</li> <li>• Lagrange Multipliers</li> <li>• Second Derivative Test</li> <li>• Gradient Ascent/Descent</li> <li>• Newton's Method</li> <li>• Conjugate Gradient Methods</li> <li>• Least Squares</li> <li>• Linear Programming (the Simplex, Ellipsoid and Interior Point Methods)</li> <li>• Dual Linear Program</li> <li>• Semidefinite Programming</li> <li>• Shortest Paths in 3D</li> <li>• Curves of Quickest Descent</li> </ul>	A-

<b>MATH 142B:</b> Introduction to Analysis II - UCSD	Fitzpatrick, P. M. (2009). <i>Advanced Calculus</i> . American Mathematical Society.  Ross, K. A. (2013). <i>Elementary Analysis: The Theory of Calculus</i> . Springer.	<ul style="list-style-type: none"> <li>• Uniform Convergence</li> <li>• Cauchy Sequences</li> <li>• Power Series</li> <li>• Abel's Theorem</li> <li>• Weierstrass Approximation</li> <li>• Differentiability</li> <li>• Mean Value Theorem</li> <li>• L'Hopital's Rule</li> <li>• Taylor's Theorem</li> <li>• Riemann Integrals</li> <li>• Continuity</li> <li>• Fundamental Theorems of Calculus</li> </ul>	A
<b>MATH 142A:</b> Introduction to Analysis I - UCSD	Ross, K. A. (2013). <i>Elementary Analysis: The Theory of Calculus</i> . Springer.	<ul style="list-style-type: none"> <li>• Limit Theorems for Sequences</li> <li>• Monotone Sequences</li> <li>• Cauchy Sequences</li> <li>• Subsequences and Series</li> <li>• Uniform Continuity</li> <li>• Limits of Functions</li> <li>• Mean Value Theorem</li> <li>• L'Hopital's Rule</li> <li>• Taylor's Theorem</li> </ul>	A
<b>MATH 135:</b> Ordinary Differential Equations - UCLA	Simmons, G. F. (2016). <i>Differential Equations with Applications and Historical Notes</i> . Chapman and Hall/CRC.	<ul style="list-style-type: none"> <li>• Laplace Transforms</li> <li>• Existence and Uniqueness Theorems</li> <li>• Lipschitz Condition</li> <li>• Convolution Theorem</li> <li>• Heaviside Expansion Theorem</li> <li>• Dirac Distribution</li> <li>• Weierstrauss M-test</li> <li>• Picard Iteration</li> <li>• Fourier Series</li> <li>• Periodic Functions</li> <li>• Convergence Theorems</li> <li>• Heat Equations</li> <li>• Sturm-Liouville Theory</li> <li>• Euler's Differential Equation</li> <li>• Green's Functions</li> <li>• Isoperimetric Problems</li> <li>• Separation of Variable Solutions to Partial Differential Equations</li> </ul>	A

<b>MATH 115A: Linear Algebra - UCLA</b>	Friedberg, S., Insel, A. and Spence, L. (2018). <i>Linear Algebra</i> . Pearson.	<ul style="list-style-type: none"> <li>• Vector Spaces &amp; Span</li> <li>• Lagrange Interpolation</li> <li>• Linear Transformations</li> <li>• Matrix Multiplication</li> <li>• Invertibility &amp; Isomorphisms</li> <li>• Eigenvalues &amp; Eigenvectors</li> <li>• Diagonalization</li> <li>• Inner Products</li> <li>• Orthogonalization &amp; Norms</li> <li>• Adjoint</li> </ul>	A
<b>MATH 109: Mathematical Reasoning - UCSD</b>	Eccles, P. J. (1997). <i>An Introduction to Mathematical Reasoning</i> . Cambridge University Press.	<ul style="list-style-type: none"> <li>• Boolean Algebra</li> <li>• Proof by Contradiction</li> <li>• Induction Principle</li> <li>• Set Theories</li> <li>• Limits of Sequences</li> <li>• Injections, Surjections and Bijections</li> <li>• Finite Sets</li> <li>• Division Algorithms</li> <li>• Euclidean Algorithm</li> <li>• Congruence Classes</li> <li>• Limits of Functions</li> <li>• Continuity &amp; Derivatives</li> </ul>	A+
<b>MATH 61: Discrete Structures - UCLA</b>	Johnsonbaugh, R. (2017). <i>Discrete Mathematics</i> . Prentice-Hall.	<ul style="list-style-type: none"> <li>• Mathematical Induction</li> <li>• Sets &amp; Functions</li> <li>• Sequences &amp; Strings</li> <li>• Matrices of Relations</li> <li>• Counting Principles</li> <li>• Generalized Permutations and Combinations</li> <li>• Pigeonhole Principle</li> <li>• Recurrence Relations</li> <li>• Graphs, Paths, and Cycles</li> <li>• Shortest-path Algorithm</li> <li>• Planar Graphs</li> <li>• Minimal Spanning Trees</li> <li>• Binary Trees and Sorting</li> <li>• Isomorphic Trees</li> </ul>	A-

<b>MATH 20F:</b> <b>Linear Algebra</b> - UCSD (with MATLAB coding)	Lay, D. C. (2011). <i>Linear Algebra and Its Applications</i> . Pearson.	<ul style="list-style-type: none"> <li>• Systems of Linear Equations</li> <li>• Vector Equations</li> <li>• Linear Independence</li> <li>• Linear Transformations</li> <li>• Invertible Matrices</li> <li>• Vector Spaces and Subspaces</li> <li>• Determinants</li> <li>• Cramer's Rule</li> <li>• Eigenvectors and Eigenvalues</li> <li>• Diagonalization</li> <li>• Orthogonality</li> <li>• Gram-Schmidt Process</li> </ul>	A
<b>MATH 20E:</b> <b>Vector Calculus</b> - UCSD (with MATLAB coding)	Marsden, J. E. and Tromba, A. J. (2011). <i>Vector Calculus</i> . W. H. Freeman.	<ul style="list-style-type: none"> <li>• Mean Value Inequality</li> <li>• Triple Integrals</li> <li>• Change of Variable Theorem</li> <li>• Vector Fields</li> <li>• Path and Line Integrals</li> <li>• Surface Integrals</li> <li>• Green's Theorem</li> <li>• Stokes' Theorem</li> <li>• Gauss Divergence Theorem</li> <li>• Gauss Law</li> </ul>	A
<b>MATH 20D:</b> Introduction to Differential Equations - UCSD	Nagle, R., Saff, E. and Snider, A. (2017). <i>Fundamentals of Differential Equations</i> . Pearson.	<ul style="list-style-type: none"> <li>• Homogeneous Linear Equations</li> <li>• Exact Equations</li> <li>• Undetermined Coefficients</li> <li>• Superposition Principle</li> <li>• Nonhomogeneous Equations</li> <li>• Variable-Coefficient Equations</li> <li>• Laplace Transforms</li> <li>• Power Series</li> <li>• Linear Systems</li> </ul>	A+

<b>MATH 20C:</b> Calculus & Analytic Geometry for Science and Engineering - UCSD	Marsden, J. E. and Tromba, A. J. (2011). <i>Vector Calculus</i> . W. H. Freeman.	<ul style="list-style-type: none"> <li>• Vectors</li> <li>• Inner Product and Cross Product</li> <li>• Differentiation</li> <li>• Paths and Curves</li> <li>• Gradients and Directional Derivatives</li> <li>• Iterated Partial Derivatives</li> <li>• Lagrange Multiplier</li> <li>• Acceleration &amp; Arc Length</li> <li>• Double Integrals</li> </ul>	A+
<b>MATH 20B:</b> Calculus for Science and Engineering - Advanced Placement Credit - UCSD	Rogawski, J., Adams, C. and Franzosa, R. (2018). <i>Calculus: Early Transcendentals</i> . W. H. Freeman.	<ul style="list-style-type: none"> <li>• Integration</li> <li>• Polar Coordinates</li> <li>• Complex Exponential</li> <li>• Trigonometric and Hyperbolic Integrals</li> <li>• Power Series</li> <li>• Taylor Series</li> </ul>	AP 5
<b>MATH 20A:</b> Calculus - Advanced Placement Credit - UCSD	Rogawski, J., Adams, C. and Franzosa, R. (2018). <i>Calculus: Early Transcendentals</i> . W. H. Freeman.	<ul style="list-style-type: none"> <li>• Limits and Continuity</li> <li>• Squeeze Theorem</li> <li>• Intermediate Value Theorem</li> <li>• Mean Value Theorem</li> <li>• L'Hopital's Rule</li> </ul>	AP 5
<b>COMPTNG 10A:</b> Introduction to Programming - UCLA (with C++ coding)	Horstmann, C. & Budd, T. (2008). <i>Big C++</i> . John Wiley & Sons.	<ul style="list-style-type: none"> <li>• Fundamental Types</li> <li>• Casting</li> <li>• Overflow &amp; Underflow</li> <li>• Loops</li> <li>• Functions</li> <li>• References</li> <li>• Vectors and Arrays in 2D</li> <li>• Classes</li> <li>• Points &amp; Pointer Arithmetic</li> <li>• Array Pointer Duality Law</li> <li>• Memory Leak</li> <li>• File and String Streams</li> </ul>	A

<p><b>CSE 3: Fluency in Information Technology</b>  - UCSD  (with C++, Java, HTML coding)</p>	<p>Lehnert, W. (2002). <i>The Web Wizard's Guide to HTML</i>. Addison-Wesley.</p> <p>Snyder, L. (2008). <i>Fluency with Information Technology</i>. Addison-Wesley.</p>	<ul style="list-style-type: none"> <li>• Path and Directory</li> <li>• ASCII</li> <li>• List and Strings</li> <li>• Loops</li> <li>• Functions</li> <li>• TCP/IP and Domain Names</li> <li>• Cookies</li> <li>• JavaScript Basics</li> <li>• HTML Basics</li> <li>• Relational Database &amp; SQL</li> <li>• Data Queries</li> <li>• Adobe Photoshop</li> <li>• Bitmap &amp; Vector Graphics</li> <li>• Encryption Schemes</li> <li>• Internet Security</li> <li>• Turing Test</li> <li>• NP-complete Problems</li> </ul>	A
<p><b>GEOG 7: Introduction to Geographic Information Systems (GIS)</b>  - UCLA  (with QGIS coding)</p>	<p>Shin, M., Campbell, J. and Burkhardt, N. (2015). <i>Essentials of Geographic Information Systems</i>. Washington, D.C. Flat World Knowledge.</p>	<ul style="list-style-type: none"> <li>• Reference Mapping</li> <li>• Cartographic Principles</li> <li>• Thematic Mapping</li> <li>• Coordinate Systems</li> <li>• Raster Data Analysis</li> <li>• Spatial Analysis</li> <li>• Map Projections</li> <li>• Data Queries</li> <li>• Vector Operations</li> </ul>	P
<p><b>PSYC 60: Introduction to Statistics in Psychology</b>  - UCSD  (with statistical R coding)</p>	<p>Howell, D. (2016). <i>Fundamental Statistics for Behavioral Sciences</i>. Wadsworth Publishing.</p>	<ul style="list-style-type: none"> <li>• Normal Distribution</li> <li>• Z scores</li> <li>• Correlation</li> <li>• Regression</li> <li>• t-tests</li> <li>• ANOVA</li> <li>• Chi-squared Test</li> </ul>	A-