# Data Analyst Nanodegree Syllabus

Discover Insights from Data with Python and SQL



#### Before You Start

Prerequisites: In order to succeed in this program, we recommend having experience working with data in Python (Numpy and Pandas) and SQL.

#### **Contact Info**

While going through the program, if you have questions about anything, you can reach us at dataanalyst-support@udacity.com. For help from Udacity Mentors and your peers, join the community discussion on Slack or visit the Udacity Classroom.

## Nanodegree Program Info

This program prepares you for a career as a data analyst by helping you learn to organize data, uncover patterns and insights, draw meaningful conclusions, and clearly communicate critical findings. You'll develop proficiency in Python and its data analysis libraries (Numpy, pandas, Matplotlib) and SQL as you build a portfolio of projects to showcase in your job search.

Depending on how quickly you work through the material, the amount of time required is variable. We have included an hourly estimation for each section of the program. The program covers one term of three month (approx. 13 weeks). If you spend about 10 hours per week working through the program, you should finish the term within 13 weeks. Students will have an additional four weeks beyond the end of the term to complete all projects.

**Length of Program\***: 1 term, 13 weeks and approximately 200 hours

**Frequency of Classes**: Self-paced within the 13-week term

Textbooks required: None

**Instructional Tools Available**: Video lectures, Text instructions, Quizzes, In-classroom mentorship

\*This is a self-paced program and the length is an estimation of total hours the average student may take to complete all required coursework, including lecture and project time. Actual hours may vary.



## Intro Project: Explore Weather Trends (5 hrs)

This project will introduce you to the SQL and how to download data from a database. You'll analyze local and global temperature data and compare the temperature trends where you live to overall global temperature trends.

## Project 1: Investigate a Dataset (40 hrs)

In this project, you'll choose one of Udacity's curated datasets and investigate it using NumPy and pandas. You'll complete the entire data analysis process, starting by posing a question and finishing by sharing your findings.

#### Supporting Lesson Content: Introduction to Data Analysis

Lesson Title	Learning Outcomes
ANACONDA	→ Learn to use Anaconda to manage packages and environments for use with Python
JUPYTER NOTEBOOKS	→ Learn to use this open-source web application to combine explanatory text, math equations, code, and visualizations in one sharable document
DATA ANALYSIS PROCESS	<ul> <li>→ Learn about the keys steps of the data analysis process</li> <li>→ Investigate multiple datasets using Python and Pandas</li> </ul>
PANDAS AND NUMPY: CASE STUDY 1	<ul> <li>→ Perform the entire data analysis process on a dataset</li> <li>→ Learn to use NumPy and Pandas to wrangle, explore, analyze, and visualize data</li> </ul>
PANDAS AND NUMPY: CASE STUDY 2	<ul> <li>→ Perform the entire data analysis process on a dataset</li> <li>→ Learn more about NumPy and Pandas to wrangle, explore, analyze, and visualize data</li> </ul>
PROGRAMMING WORKFLOW FOR DATA ANALYSIS	→ Learn about how to carry out analysis outside Jupyter notebook using IPython or the command line interface

## Project 2: Analyze Experiment Results (45 hrs)

In this project, you will be provided a dataset reflecting data collected from an experiment. You'll use statistical techniques to answer questions about the data and report your conclusions and recommendations in a report.



## Supporting Lesson Content: Practical Statistics

Lesson Title	Learning Outcomes
SIMPSON'S PARADOX	→ Examine a case study to learn about Simpson's Paradox
PROBABILITY	→ Learn the fundamental rules of probability
BINOMIAL DISTRIBUTION	<ul> <li>→ Learn about binomial distribution where each observation represents one of two outcomes</li> <li>→ Derive the probability of a binomial distribution</li> </ul>
CONDITIONAL PROBABILITY	→ Learn about conditional probability, i.e., when events are not independent
BAYES RULE	<ul> <li>→ Build on conditional probability principles to understand the Bayes rule</li> <li>→ Derive the Bayes theorem</li> </ul>
STANDARDIZING	<ul> <li>→ Convert distributions into the standard normal distribution using the Z-score</li> <li>→ Compute proportions using standardized distributions</li> </ul>
SAMPLING DISTRIBUTIONS AND CENTRAL LIMIT THEOREM	<ul> <li>→ Use normal distributions to compute probabilities</li> <li>→ Use the Z-table to look up the proportions of observations above, below, or in between values</li> </ul>
CONFIDENCE INTERVALS	→ Estimate population parameters from sample statistics using confidence intervals
HYPOTHESIS TESTING	→ Use critical values to make decisions on whether or not a treatment has changed the value of a population parameter
T-TESTS AND A/B TESTS	→ Test the effect of a treatment or compare the difference in means for two groups when we have small sample sizes
REGRESSION	<ul> <li>→ Build a linear regression model to understand the relationship between independent and dependent variables</li> <li>→ Use linear regression results to make a prediction</li> </ul>
MULTIPLE LINEAR REGRESSION	→ Use multiple linear regression results to interpret coefficients for several predictors
LOGISTIC REGRESSION	→ Use logistic regression results to make a prediction about the relationship between categorical dependent variables and predictors



## Project 3: Wrangle and Analyze Data (50 hrs)

Real-world data rarely comes clean. Using Python, you'll gather data from a variety of sources, assess its quality and tidiness, then clean it. You'll document your wrangling efforts in a Jupyter Notebook, plus showcase them through analyses and visualizations using Python and SQL.

#### Supporting Lesson Content: Data Wrangling

Lesson Title	Learning Outcomes
INTRO TO DATA WRANGLING	<ul> <li>→ Identify each step of the data wrangling process (gathering, assessing, and cleaning)</li> <li>→ Wrangle a CSV file downloaded from Kaggle using fundamental gathering, assessing, and cleaning code</li> </ul>
GATHERING DATA	<ul> <li>→ Gather data from multiple sources, including gathering files, programmatically downloading files, web-scraping data, and accessing data from APIs</li> <li>→ Import data of various file formats into pandas, including flat files (e.g. TSV), HTML files, TXT files, and JSON files</li> <li>→ Store gathered data in a PostgreSQL database</li> </ul>
ASSESSING DATA	<ul> <li>→ Assess data visually and programmatically using pandas</li> <li>→ Distinguish between dirty data (content or "quality" issues) and messy data (structural or "tidiness" issues)</li> <li>→ Identify data quality issues and categorize them using metrics: validity, accuracy, completeness, consistency, and uniformity</li> </ul>
CLEANING DATA	<ul> <li>→ Identify each step of the data cleaning process (defining, coding, and testing)</li> <li>→ Clean data using Python and pandas</li> <li>→ Test cleaning code visually and programmatically using Python</li> </ul>

## Project 4: Communicate Data Findings (50 hrs)

In this project, you will use Python's data visualization tools to systematically explore a selected dataset for its properties and relationships between variables. Then, you will create a presentation that communicates your findings to others.

Supporting Lesson Content: Data Visualization with Python

Lesson Title Learning Outcomes



DATA VISUALIZATION IN DATA ANALYSIS	<ul> <li>→ Understand why visualization is important in the practice of data analysis</li> <li>→ Know what distinguishes exploratory analysis from explanatory analysis, and the role of data visualization in each</li> </ul>
DESIGN OF VISUALIZATIONS	<ul> <li>→ Interpret features in terms of level of measurement</li> <li>→ Know different encodings that can be used to depict data in visualizations</li> <li>→ Understand various pitfalls that can affect the effectiveness and truthfulness of visualizations</li> </ul>
UNIVARIATE EXPLORATION OF DATA	<ul> <li>→ Use bar charts to depict distributions of categorical variables</li> <li>→ Use histograms to depict distributions of numeric variables</li> <li>→ Use axis limits and different scales to change how your data is interpreted</li> </ul>
BIVARIATE EXPLORATION OF DATA	<ul> <li>→ Use scatterplots to depict relationships between numeric variables</li> <li>→ Use clustered bar charts to depict relationships between categorical variables</li> <li>→ Use violin and bar charts to depict relationships between categorical and numeric variables</li> <li>→ Use faceting to create plots across different subsets of the data</li> </ul>
MULTIVARIATE EXPLORATION OF DATA	<ul> <li>→ Use encodings like size, shape, and color to encode values of a third variable in a visualization</li> <li>→ Use plot matrices to explore relationships between multiple variables at the same time</li> <li>→ Use feature engineering to capture relationships between variables</li> </ul>
EXPLANATORY VISUALIZATIONS	<ul> <li>→ Understand what it means to tell a compelling story with data</li> <li>→ Choose the best plot type, encodings, and annotations to polish your plots</li> <li>→ Create a slide deck using a Jupyter Notebook to convey your findings</li> </ul>
VISUALIZATION CASE STUDY	→ Apply your knowledge of data visualization to a dataset involving the characteristics of diamonds and their prices

