**CCT College Dublin**

**Assessment Cover Page**

*To be provided separately as a word doc for students to include with every submission*

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| **Module Title:** | **Programming for DA**  **Statistics for Data Analytics**  **Machine Learning for Data Analysis**  **Data Preparation & Visualisation** |
| **Assessment Title:** | **MSC\_DA\_CA1** |
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**Muhammad Asif 12-11-2023**

**Declaration**

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| By submitting this assessment, I confirm that I have read the CCT policy on Academic Misconduct and understand the implications of submitting work that is not my own or does not appropriately reference material taken from a third party or other source. I declare it to be my own work and that all material from third parties has been appropriately referenced. I further confirm that this work has not previously been submitted for assessment by myself or someone else in CCT College Dublin or any other higher education institution. |

**Abstract**

*Data analytics plays a critical and increasingly important role in today's data-driven world. Its significance can be seen in various aspects of business, science, and society. In this report, we utilized data analytics techniques to gain useful insights from the Annual Population change in Ireland. Data is imported and checked thoroughly to answer questions. During the process, it went through different stages from preparation for analysis to graphical representation, using statistics to seek trends and finally creating a Machine Learning model to seek the output parameters from validation set. Whole work is enclosed in Jupyter Notebook which used Python framework to carry out the programming requirements*

**Introduction**

When it comes to examining annual population changes, data analytics plays a crucial role in the field of population studies. In order to uncover trends, patterns, and invaluable insights, it entails carefully analyzing data. Its significance stems from the fact that it facilitates an understanding of the intricacies of population dynamics, including migration trends, birth and death rates, and other crucial demographic factors that significantly influence a region's evolutionary processes. By analyzing and interpreting massive datasets, data analytics enables academics and decision-makers to make well-informed decisions in areas like resource allocation, urban planning, and the development of successful policies. In this case, the data analysis pipeline entails several crucial procedures. These include collecting, sanitizing, and organizing data; doing exploratory analysis to look for hidden patterns; and, finally, developing predictive models. Through these procedures, data analytics contributes significantly to the transformation of raw data into insightful knowledge, improving comprehension of population changes, and enabling more informed, data-driven decision-making in the ever-changing field of demographics.

**Data Preprocessing**

Data preprocessing is a vital stage in the data analysis pipeline that involves a number of crucial tasks related to preparing raw data for machine learning and statistical analysis. Data cleansing, which addresses outliers, inconsistent values, and missing values in the dataset, is one of the initial processes. Standardization or normalization is frequently required to bring conflicting data to a common scale, outliers can be discovered and handled to stop them from influencing the model, and imputation techniques can be used to address missing data. Data transformation, which includes feature engineering to create new, informative attributes that might enhance model performance, scaling of numerical features, and categorical variable encoding, is another essential element. In order to decrease dimensionality and maybe enhance model precision and computational overhead, feature selection strategies, and data reduction techniques like Principal Component Analysis (PCA) are used. In essence, data preparation is a crucial step that ensures the data is in the best possible shape so that machine learning models can generate accurate predictions.

**Data Visualization**

Data visualization, which involves displaying data visually to detect trends, correlations, and other patterns, is an essential phase in the data analysis process. Effective data visualization requires not only the creation of aesthetically pleasing charts and graphs, but also the selection of the right visualization techniques based on the objectives and type of data. Histograms for data distribution, scatter plots for variable relationships, bar charts for comparisons, line charts for trends over time, and heat maps for pattern detection are just a few examples of the many various kinds of visualizations that can be employed. Because its ultimate purpose is to enable data-driven decision-making and effective distribution of insights to stakeholders, data visualization is an essential component of the data analysis process.

**Machine Learning**

Algorithms and mathematical structures known as machine learning models enable computers to learn from data and make predictions or judgments without the need for explicit programming. These models embrace a wide range of methods: unsupervised learning works with unlabeled data to find innate patterns and structures, whereas supervised learning concentrates on labeled data, where the model is trained to generate predictions based on input features. Several methods, including support vector machines, decision trees, deep neural networks, and linear regression, are used in supervised learning to tackle tasks like regression and classification. For tasks like data segmentation and feature reduction, unsupervised learning includes clustering algorithms like K-means and dimensionality reduction techniques like Principal Component Analysis (PCA). Furthermore, reinforcement learning is a subfield of machine learning that is used in robotics and autonomous systems to teach agents how to maximize cumulative rewards by making a series of decisions in dynamic situations. The problem at hand determines which machine learning model is best, and in order to guarantee optimal model performance, careful processes in model selection, hyper parameter tuning, and model evaluation are necessary. It's critical to recognize that the overall performance of machine learning models in real-world applications is strongly influenced by the caliber of the data as well as the efficiency of data preparation and feature engineering.

**Annual Population Ireland**

One of the primary functions of data analytics in the study of annual population changes is to identify trends. By scrutinizing historical data, statisticians and demographers can determine whether the population is expanding, declining, or remaining stable. This information is of utmost importance to governments, urban planners, and businesses, as it allows them to foresee shifts in resource requirements, infrastructure needs, and market opportunities. Moreover, data analytics enables the identification of the factors driving these changes, such as birth rates, death rates, and migration patterns, offering a more profound understanding of their root causes.

This report adheres to the same approach and utilizes Annual Population Change Data for Ireland to provide the government and citizens with an overview of evolving population trends and its components, including annual births and deaths, immigrants and emigrants, natural increase, net migration derived from formal variables, total population, and population distribution. Data has been collected from 1951 to 2023, with values measured in thousands.

In the next stages, we will thoroughly discuss the methods from which we drive the statistics and useful insights using various data analytics techniques. We will be discussing the important python libraries that enabled us to carry the required operation and how each analytics plays its fundamental role in identifying the patterns and trends for our dataset.

**Methodology**

**Data Preprocessing**

We utilized the important python packages at each step to carry out specific task. First, is the use of Python Pandas to import and manipulate the dataset. The first thing which is needed to be done is to look for any ambiguity in our dataset to avoid any noise or wrong information. Therefore, we used head () and sine () method with sum () to look for any missing values and we found 144 of these in the VALUE field. From the VALUE distribution plot, we found out that there is low saturation of data in the beginning rows and it gets denser afterwards which leads to the fact that data for year 1986 and below were missing. Since, we got enough years to carry out our analysis so we discarded that row and cleaned our dataset.

**Statistics**

The normal distribution, often referred to as the Gaussian distribution or the bell curve, holds paramount importance in statistics and various scientific disciplines due to its versatility and prevalence. It serves as a fundamental model for characterizing the distribution of data in natural and man-made phenomena. The central limit theorem, a cornerstone of statistics, underscores its significance, as it dictates that the means of repeated random samples from any population tend to follow a normal distribution, even when the underlying data is not normally distributed. This crucial characteristic allows researchers and statisticians to conduct hypothesis testing, estimate parameters accurately, and draw strong conclusions about populations. The normal distribution offers a common framework for comprehending and modeling variability, which helps to simplify complicated real-world issues and aids in forecasting and decision-making. The normal distribution is a universal and essential concept in the field of statistics and empirical research, with applications spanning from physics and engineering to economics and biology. It is a potent tool for analyzing and forecasting data. It is the cornerstone of probability and statistics due to its symmetrical, well-defined properties and broad applicability, which enhances our comprehension of the world and informs innumerable practical applications.

We plotted the normal distribution's probability density function for our dataset using the spicy package. This information was crucial because it allowed us to understand the variance and average estimates of each parameter along the bell curve, as depicted in figure 01.

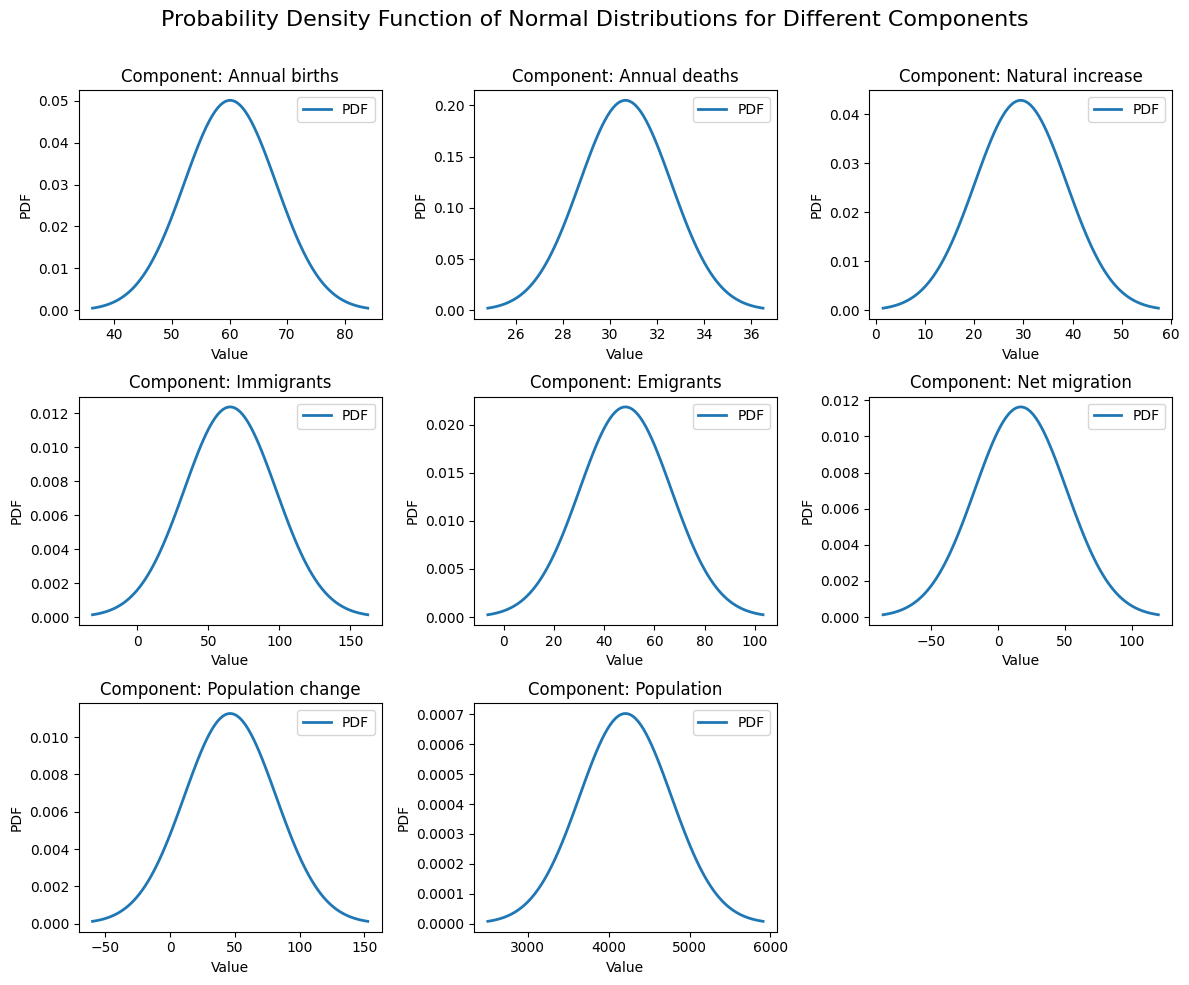


Figure 01. PDF of Normal Distribution for each Component

**Data Preparation and Visualization**

Bar graphs are an essential part of this visual communication toolkit, which is used in data visualization, a potent technique for communicating complex information. Bar graphs offer a simple and straightforward method for comparing values between various categories and representing categorical data. Usually, they are made up of rectangular bars that can be vertical or horizontal, and each bar's length corresponds to the quantity it represents. Bar graphs are a vital option for many applications, ranging from science and education to business and finance, due to their simplicity and efficacy. They are the perfect option for displaying discrete data, such as sales figures for different products, student performance across subjects, or population distribution across regions, because they allow viewers to quickly understand trends, variations, and relationships within the data. Bar graphs can be further enhanced by incorporating color coding, grouping, and stacking to highlight specific insights within the data. Their adaptability, accessibility, and straightforward design make them a staple in the world of data visualization, helping analysts, researchers, and decision-makers present their findings in a manner that is both visually engaging and easy to interpret, thereby facilitating more informed decisions and a deeper understanding of the underlying informationTop of Form. After statistical analysis, we prepared our data for the exploratory analysis by eliminating the STATISTIC Label and UNIT column as they didn’t serve any purpose. Our data required reshaping since the categories in component field needed to be align in columns not rows so we used pivot () method from pandas to reshaped our data. We started with the describe () method that gave the additional statistical information with 25%, 50% and 75% quantiles which are important to draw projects at different intervals of data.