Dataset Description and PCA Analysis

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1. Dataset Description

1.1 Overview

The Online Shoppers Purchasing Intention Dataset is sourced from the UCI Machine Learning Repository and provides valuable insights into online shopping behavior. This dataset tracks user interactions on an e-commerce website, with the primary goal of predicting whether a shopper will make a purchase during their visit.

It includes features such as the number of pages visited, time spent on various types of pages, bounce rates, and exit rates. Additionally, the dataset captures demographic information about the shoppers, such as whether they are new or returning visitors. The target variable indicates whether the shopper made a purchase, making the dataset a valuable resource for understanding and modeling purchasing behavior.

The original dataset contains 12,330 records and 18 features, including both numerical and categorical variables. For our analysis, we've taken a subset of the original dataset which contains 2,000 records and 18 features.

1.2 Features

Below is a table representation with all the key features:

Feature Name	Type	Description
Administrative	Quantitative	Number of administrative pages visited
AdministrativeDuration	Quantitative	Time spent on administrative pages in seconds
Informational	Quantitative	Number of informational pages visited
InformationalDuration	Quantitative	Time spent on informational pages in seconds
ProductRelated	Quantitative	Number of product-related pages visited
${\bf Product Related Duration}$	Quantitative	Time spent on product-related pages in seconds
BounceRates	Quantitative	Percentage of visitors leaving after one page
ExitRates	Quantitative	Percentage of sessions exiting from each page
PageValues	Quantitative	Average value attributed to a page
SpecialDay	Quantitative	Metric indicating proximity to significant holidays
Weekend	Binary	Indicates if the session occurred on a weekend (1 = Yes, $0 = No$)
Revenue	Binary	Indicates if the session resulted in a purchase $(1 = Yes, 0 = No)$
VisitorType	Categorical	Visitor category (e.g., Returning, New, Other)
Month	Categorical	Month of the visit (e.g., Jan, Feb)
OperatingSystems	Categorical	Visitor operating system (e.g., Windows, MacOS)
Browser	Categorical	Browser used by the visitor (e.g., Safari, Chrome)
Region	Categorical	Visitor geographical region
TrafficType	Categorical	Type of traffic source leading to the visit

To obtain this structure a preprocessing step was performed, which included the following transformations: - The categorical variables were converted to factors. - The target variable (0 for no purchase, 1 for purchase) and the binary variables were converted to a binary format - Name changes were made to some variables for better readability and naming consistency. - A random subset of 2,000 observations was selected from the original dataset.

The code to preprocess the dataset is as follows:

```
set.seed(100)
data <- read_csv(</pre>
  paste('../../data/online+shoppers+purchasing+intention+dataset/',
        'online_shoppers_intention.csv', sep = '')
)
data <- data %>%
  mutate(
   Month = as.factor(Month),
   Region = as.factor(Region),
   TrafficType = as.factor(TrafficType),
   VisitorType = as.factor(VisitorType),
   OperatingSystems = as.factor(OperatingSystems),
   Browser = as.factor(Browser),
   Weekend = as.factor(as.numeric(Weekend == 'TRUE')),
   Revenue = as.factor(as.numeric(Revenue == 'TRUE')),
    SpecialDay = as.factor(SpecialDay)
  ) %>%
  rename(
    AdministrativeDuration = Administrative_Duration,
    InformationalDuration = Informational_Duration,
   ProductRelatedDuration = ProductRelated_Duration
  ) %>%
  sample_n(2000)
```

1.3 Data Integrity

We begin by assessing the integrity of the dataset by analyzing the presence of missing data across its features. The table below presents the count of missing values for each feature. As shown, there are **no missing values** in any of the features, which indicates that the dataset is complete and does not require any imputation for missing data.

Additionally, we check for duplicate records in the dataset. The analysis revealed **6 duplicate rows**, which may be due to several factors. These duplicates could represent unintentional repetitions of the same individual's data, or they could arise from issues during data collection, such as individuals' data being recorded more than once under slightly different conditions. As the amount is relatively small, we will proceed with the analysis without removing these duplicates.

Given the absence of missing values and the limited number of duplicate rows, the dataset does not require any further cleaning or preprocessing steps before proceeding with our initial analysis.

Table 2: Missing Values Count for Each Feature

	Х
Administrative	C
AdministrativeDuration	C
Informational	0
InformationalDuration	0
ProductRelated	0
${\bf Product Related Duration}$	0
BounceRates	C
ExitRates	0
PageValues	C

	X
SpecialDay	0
Month	0
OperatingSystems	0
Browser	0
Region	0
TrafficType	0
VisitorType	0
Weekend	0
Revenue	0

#> Number of duplicate rows: 6

2. Univariate Analysis

2.1 Numeric Variables

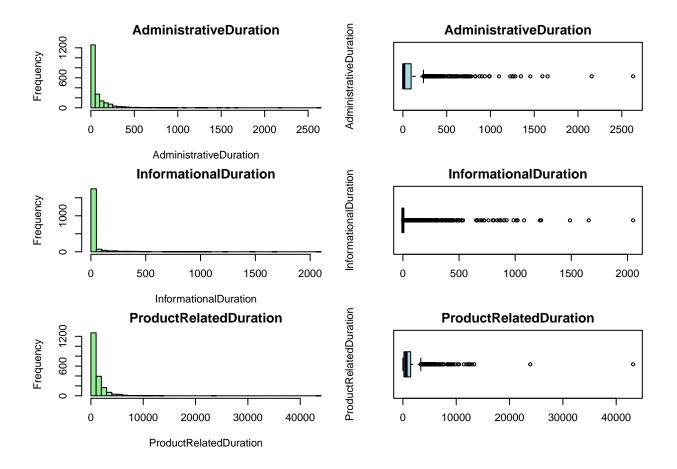
2.1.1 Time Related Variables

The duration variables (e.g., AdministrativeDuration, InformationalDuration, ProductRelatedDuration) exhibit values for the mean significantly higher than the median, indicating a heavily right-skewed distribution. This suggests that there are a few sessions with very high durations that are pulling the mean upwards. The range of values for these variables is quite large, with the maximum values being several times higher than the 75th percentile.

Table 3: Summary of Numeric Time Related Variables

AdministrativeDuration	In formation al Duration	${\bf Product Related Duration}$
Min.: 0.00	Min.: 0.00	Min.: 0.0
1st Qu.: 0.00	1st Qu.: 0.00	1st Qu.: 187.7
Median: 10.05	Median: 0.00	Median: 602.0
Mean: 79.24	Mean: 36.28	Mean: 1161.3
3rd Qu.: 94.00	3rd Qu.: 0.00	3rd Qu.: 1459.5
Max. :2629.25	Max. $:2050.43$	Max. :43171.2

The boxplots and histograms provide clear evidence of the right-skewed nature of these variables, characterized by a high concentration of values near zero and a long tail of extreme values. This distribution highlights the presence of numerous outliers, which are not isolated anomalies but rather a significant portion of the data. These extreme values likely represent important behavioral patterns or user interactions that could offer valuable insights. Therefore, instead of removing or treating them as noise, we choose to retain these outliers in our analysis to ensure a comprehensive understanding of the dataset and its implications.



2.1.2 Page Interaction Variables

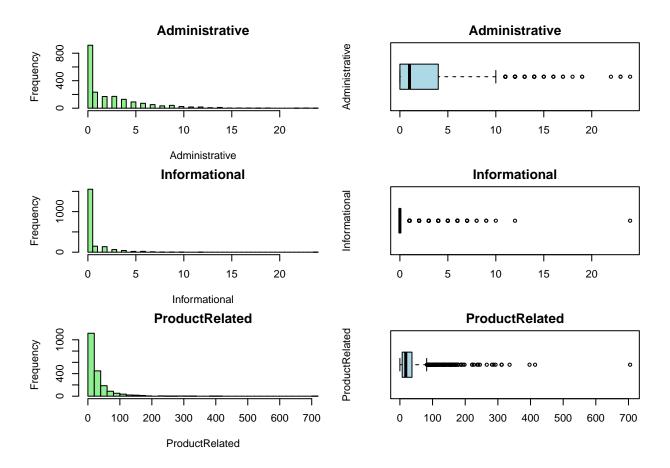
The variables related to page interactions (e.g., Administrative, Informational, ProductRelated) represent the number of pages visited by the user during the session. From all 3 variables, ProductRelated has the highest mean and median values while Informational and Administrative have lower but similar values. The summary statistics also suggest a right-skewed distribution for these variables, with the mean exceeding the median.

Table 4: Summary of Page Interaction Variables

Administrative	Informational	${\bf ProductRelated}$
Min.: 0.000	Min.: 0.000	Min.: 0.00
1st Qu.: 0.000	1st Qu.: 0.000	1st Qu.: 7.00
Median: 1.000	Median: 0.000	Median: 18.00
Mean: 2.321	Mean: 0.577	Mean: 30.53
3rd Qu.: 4.000	3rd Qu.: 0.000	3rd Qu.: 37.00
Max. :24.000	Max. $:24.000$	Max. $:705.00$

The boxplots and histograms for these variables confirm the right-skewed distribution, with a large number of sessions having low page interaction counts and a few sessions with very high counts. While all three variables exhibit similar patterns, the Informational variable has a more pronounced skewness, followed by ProductRelated and Administrative. The presence of outliers in these variables is expected, as user behavior can vary significantly, with some users exploring multiple pages while others may exit quickly after

viewing a few pages. We believe this outlier behavior is essential for understanding user engagement and purchase intent and should be retained in the analysis.



2.1.3 Other Numeric Variables

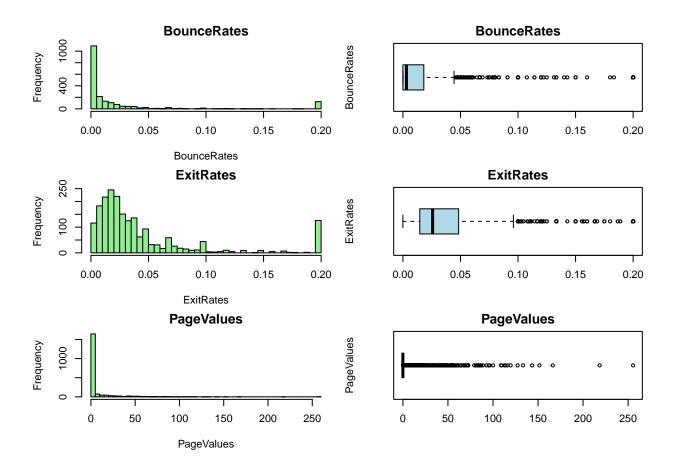
The following variables are derived from Google Analytics, providing insights into user behavior on the website:

- BounceRates: Represents the percentage of visitors who leave the site after viewing only one page. Summary statistics show values ranging from 0 to 0.2, with a mean of 0.023 and a median of 0.003. The distribution is right-skewed, as the mean is higher than the median, indicating a concentration of low values and a few higher values.
- ExitRates: Represents the percentage of visitors who exit the site from a specific page. While the dataset doesn't specify which page this refers to, it may be a critical page or one leading to a purchase. The values range from 0 to 0.2, with a mean of 0.043 and a median of 0.025. Like BounceRates, this variable is also right-skewed although not as extreme, with a concentration of low values and a few higher values.
- PageValues: Represents the average value of a page that a user visits before completing a transaction. The values range from 0 to 255, with a mean of 5.89 and a median of 0.0. This variable is heavily right-skewed, indicating that while most pages have little to no assigned value. There are a few pages with high values that significantly impact the mean.

Table 5: Summary of Page Interaction Variables

BounceRates	ExitRates	PageValues
Min. :0.000000	Min. :0.00000	Min.: 0.000
1st Qu.:0.000000	1st Qu.:0.01467	1st Qu.: 0.000
Median $:0.003089$	Median: 0.02581	Median: 0.000
Mean $:0.023104$	Mean $:0.04380$	Mean: 5.899
3rd Qu.:0.018182	3rd Qu.:0.04839	3rd Qu.: 0.000
Max. $:0.200000$	Max. $:0.20000$	Max. :255.569

The visualizations for these variables further confirm the right-skewed distribution, with a concentration of low values and a few high values. The histograms and boxplots provide a clear representation of the distribution of these variables, highlighting the presence of outliers and the need to consider these extreme values in the analysis. We've observed outliers in all variables which may suggest specific user behaviors or interactions that could be crucial for understanding purchasing intent and user engagement on the website. As done with other variables, we choose to retain these outliers in our analysis to ensure a comprehensive understanding of the dataset.



2.2 Categorical Variables

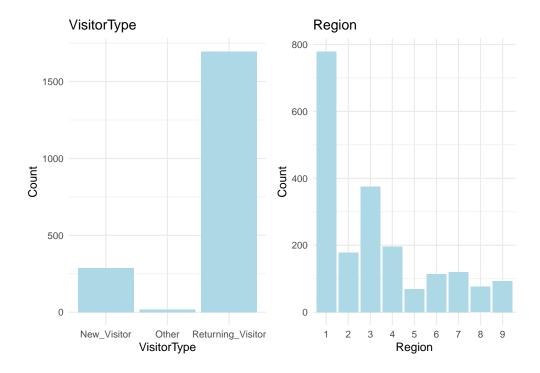
2.2.1 Visitor Demographics/Identity

• VisitorType:

The majority of visitors are **Returning Visitors**, indicating a strong base of repeat users. **New Visitors** form a smaller proportion, and the **Other** category is negligible, suggesting minimal contribution from other visitor types.

• Region:

Region 1 has the highest count of visitors, followed by Region 3. Traffic is highly concentrated in these specific regions, with other regions showing a relatively lower number of visitors.



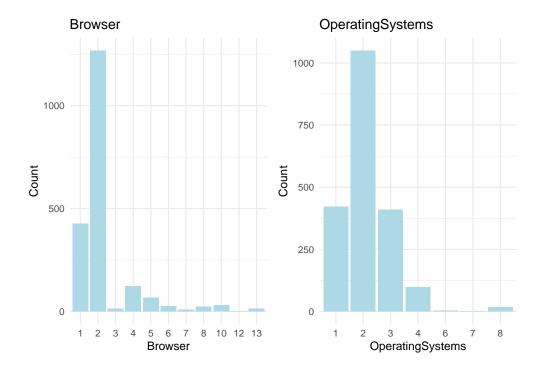
2.2.2 Technical Attributes

Browser:

Browser 2 is the most commonly used browser, significantly surpassing others in popularity. Browser 1 has a notable user base, while the other browsers are used by only a small fraction of users.

• OperatingSystems:

Operating System 2 dominates usage, followed by Operating System 1. Other operating systems have a minor presence, highlighting user preference for a few dominant operating systems.



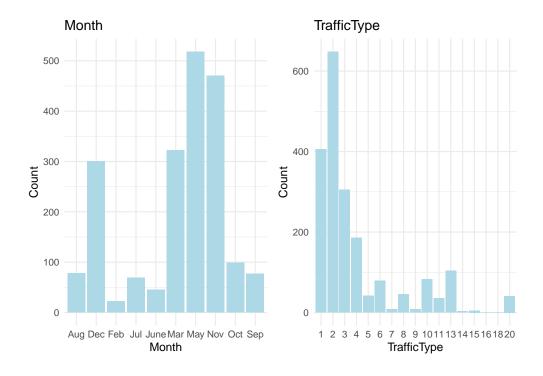
2.2.3 Traffic and Source

• Month:

Traffic peaks in March, May, and November, with the highest activity in May. Lower traffic is observed in June, July, and February, potentially reflecting seasonal trends or business cycles.

• TrafficType:

TrafficType 2 is the primary source of traffic, with a significant lead over TrafficType 1. Other TrafficTypes contribute marginally, indicating that a few referral sources or marketing channels drive the majority of traffic.



2.3 Binary Variables

• Weekend:

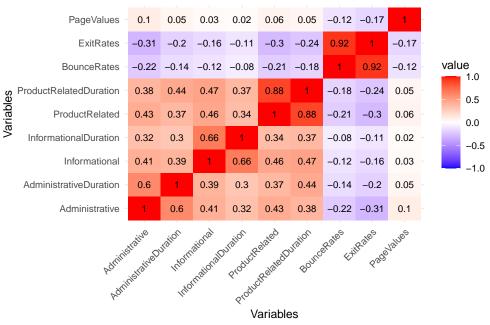
The binary variable Weekend indicates whether the session occurred on a weekend. We observe that around $\sim 23\%$ of the sessions occurred on weekends, while the majority ($\sim 77\%$) took place on weekdays. This distribution suggests that the website receives a higher volume of traffic on weekdays compared to weekends (which also holds if we took the daily average).

• Revenue: The target variable Revenue indicates whether a session resulted in a purchase. The dataset is imbalanced, with a higher number of sessions where no purchase (~85%) was made compared to sessions resulting in a purchase (~15%). This imbalance is expected in e-commerce datasets, where the conversion rate is typically lower than the non-conversion rate.



3. Multivariate Analysis





#>		PC1	PC2	PC3	PC4
#>	Administrative	0.37105796	0.0329912	0.21953028	-0.05220820
#>	${\tt AdministrativeDuration}$	0.34714509	0.1235379	0.19485757	0.01541194
#>	Informational	0.37346416	0.2609986	0.04525560	-0.39448582

```
#> InformationalDuration
                       #> ProductRelated
#> ProductRelatedDuration 0.41707403 0.1563865 -0.21780465
                                                      0.45935918
#> BounceRates
                       -0.25010949 0.6305022 0.16507323
                                                      0.16667502
#> ExitRates
                       -0.29019922 0.6081152 0.11151647
                                                      0.11549280
#> PageValues
                       0.07613712 -0.2007321 0.87761893 0.26375189
                              PC5
#>
                                         PC6
                                                     PC7
                       #> Administrative
#> AdministrativeDuration 0.60278060 -0.662670738 -0.049419726
                                                         0.1448946839
#> Informational
                       -0.23809702 -0.024825486 0.759799087
                                                         0.0007661561
#> InformationalDuration -0.33796380 -0.026948406 -0.641251513 0.0299799561
#> ProductRelated
                       -0.20213008 0.153751548 -0.040885215
                                                         0.6700840381
#> ProductRelatedDuration -0.18765289 -0.123524046 -0.071149018 -0.6888790978
                       0.04411895 0.098547988 -0.010363055 0.1339391720
#> BounceRates
#> ExitRates
                       #> PageValues
                       -0.33345823 -0.039138054 0.010958484 -0.0100583735
#>
                               PC9
#> Administrative
                       0.044425841
#> AdministrativeDuration 0.010389639
#> Informational
                       -0.011634553
#> InformationalDuration
                       0.004633944
#> ProductRelated
                       0.155956055
#> ProductRelatedDuration -0.114255943
#> BounceRates
                       -0.674668205
#> ExitRates
                       0.709947449
#> PageValues
                       0.034384375
#> [1] 3.57835089 1.72376992 0.96781746 0.93079294 0.88913071 0.40830014 0.32216982
#> [8] 0.11170116 0.06796695
#>
              [,1]
                        [,2]
                                  [,3]
                                             [,4]
                                                        [,5]
                                                                   [,6]
#>
  [1,] -0.37105796 -0.0329912 0.21953028 -0.05220820 0.52507075 0.713679807
   [2,] -0.34714509 -0.1235379 0.19485757 0.01541194 0.60278060 -0.662670738
   [4,] -0.31872058 -0.2662497 0.07498729 -0.54298773 -0.33796380 -0.026948406
  [5,] -0.41740872 -0.1120424 -0.22096803 0.47422249 -0.20213008 0.153751548
#>
   [6,] -0.41707403 -0.1563865 -0.21780465 0.45935918 -0.18765289 -0.123524046
  [7,] 0.25010949 -0.6305022 0.16507323 0.16667502 0.04411895 0.098547988
#>
#>
  [8,] 0.29019922 -0.6081152 0.11151647 0.11549280 0.02890485 0.006699396
#>
  [9,] -0.07613712  0.2007321  0.87761893  0.26375189 -0.33345823 -0.039138054
#>
               [,7]
                           [,8]
                                       [,9]
#>
  [1,] 0.045214693 -0.1457811805
                               0.044425841
   [2,] 0.049419726 0.1448946839 0.010389639
   [3,] -0.759799087  0.0007661561 -0.011634553
#>
#>
   [4,] 0.641251513 0.0299799561 0.004633944
#> [5,] 0.040885215 0.6700840381 0.155956055
#> [6,] 0.071149018 -0.6888790978 -0.114255943
#> [7,] 0.010363055 0.1339391720 -0.674668205
#> [8,] -0.007341775 -0.1234750628 0.709947449
#> [9,] -0.010958484 -0.0100583735 0.034384375
```

5. References

• UCI Machine Learning Repository. (n.d.). Online Shoppers Purchasing Intention Dataset. Retrieved from https://archive.ics.uci.edu/dataset/468/online+shoppers+purchasing+intention+dataset.