TIL Programming Project Group

Group Members:

- Alex van Wijngaarden 4874277
- Eka Devi Wulandari 5912822
- Nadhira Zahrani Widiafina 5835429
- · Simon, van der Horst 5125227
- Wiet Kalf 4498445

Research Objective

Main question: "How does the weather influence traffic accidents in urban and rural areas in the Netherlands during the period of 2021-2022?".

Sub questions:

- 1. Do urban and rural areas have significantly different amount of traffic accidents?
- 2. Does weather type influence the amount of traffic accidents in urban and rural areas?
- 3. Does weather type influence the severity of the traffic accidents?

Research Report

1. Introduction

In the intricate tapestry of our daily lives, transport plays an important role. Going from A to B, people use many different modes to travel, whether by car, bus, bike or foot. Yet, while traveling, the potential for accidents lurks, influenced by an array of factors. One of these factors is the weather. To further understand how and why the weather plays a potential pivotal role in the amount of accidents in the Netherlands, a research question has been set:

How does the weather influence traffic accidents in urban and rural areas in the Netherlands during the period of 2021-2022?

To ensure that the research question can be answered, subquestions have also been formulated to support the validity:

- 1. Are road accidents more common in urban areas compared to rural areas?
- 2. Does the type of weather influence the amount of accidents?
- 3. Doest the type of weather influence the severity of accidents?

2. Data Processing

```
In [1]: M
    import pandas as pd
    import requests
    from time import sleep
    from requests.packages.urllib3.util.retry import Retry
    import plotly.express as px
    import numpy as np
    from sklearn.cluster import KMeans
    import matplotlib.pyplot as plt
    from scipy import stats
    from scipy.stats import f_oneway
    from scipy.stats import chi2_contingency
```

Dataset 1: Road accidents data in Netherlands

```
df_accidents = df_accidents[['AP3_CODE', 'WGD_CODE_1', 'GME_NAAM', 'PVE_NAAM']]
            df_accidents_renamed = df_accidents.rename(columns={'AP3_CODE': 'Severity', 'WGD_CODE_1': 'Weather_Type', 'GME_NAAM': 'Mu
            df accidents renamed
   Out[7]:
                    Severity Weather Type
                                            Municipality
                                                      Province Name
                       DOD
                                      D
                                                        Noord-Holland
                                             Amsterdam
                15
                       DOD
                                     D
                                                Tilburg
                                                        Noord-Brabant
                16
                      DOD
                                     D
                                        's-Hertogenbosch
                                                        Noord-Brabant
                30
                       UMS
                                     D
                                         Súdwest-Fryslân
                                                            Friesland
                31
                       UMS
                                     D
                                          Smallingerland
                                                            Friesland
                                     D
                                                Voorst
             113136
                      UMS
                                     D
                                                Niikerk
                                                          Gelderland
             113140
                      UMS
                                     D
                                                         Zuid-Holland
                                             Rotterdam
             113149
                       UMS
                                     D
                                        Haarlemmermee
                                                        Noord-Holland
             113154
                      UMS
                                     D
                                                             Drenthe
                                             Hoogeveen
```

Dataset 2: Municipality's density and road length data

```
▶ # Adding data reference from CBS for clustering analysis
In [2]:
             clustering_raw_data = r"C:\Users\LENOVO YOGA\TIL6022 Phython Programming\TIL 6022 TIL Python Programming\00 Final Project
             df_cluster_data = pd.read_csv(clustering_raw_data, encoding = 'latin1', low_memory=False)
             # Display only the relevant / needed columns
             df cluster data = df cluster data [['Municipality final','Population','Area','Density','Road Length']]
             df cluster data = df cluster data.rename(columns={'Municipality final': 'Municipality Name'})
             df cluster data
              4
    Out[2]:
                  Municipality_Name
                                                     Density
                                    Population
                                                             Road_Length
                0
                        Aa en Hunze
                                        25579 276.66
                                                          93
                                                                      639
                1
                                               20.00
                                                        1615
                                                                      157
                             Almere
                                        32452
                2
                        Alblasserdam
                                        20087
                                                8.87
                                                        2290
                                                                       82
                3
                       Albrandswaard
                                        25934
                                               21.17
                                                        1197
                                                                      151
                                       110783
                                               110.01
                                                        1004
                                                                      594
                            Alkmaar
              229
                         Zoetermeer
                                       125767
                                               34.43
                                                        3653
                                                                      406
              230
                                         9302
                                               21.16
                                                         440
                                                                       63
                        Zoeterwoude
              231
                            Zuidplas
                                        45794
                                               57.76
                                                         790
                                                                      316
              232
                                        44789
                                               20.01
                                                        2206
                         Zwijndrecht
                                                                      209
                                       130668 110.07
                                                                      778
              233
                                                        1181
```

234 rows × 5 columns

7704 rows × 4 columns

3. Analysis

3.1 Traffic Accident Heatmap Visualization

Section 3.1 focuses on the synthesis of accident data with geospatial analysis to create an intuitive heatmap visualization. This method graphically represents the density and distribution of traffic accidents, offering a clear visual summary of areas with higher and lower incident frequencies. Heatmaps are particularly effective in revealing hotspots and patterns that might not be immediately apparent from raw data, thereby providing valuable insights for further investigation and targeted safety measures.

3.1.1 Display Municipality Information

The visualization process begins by identifying and enumerating the unique municipalities involved in the traffic accident dataset. Utilizing the GME_NAAM column, the code snippet isolates each distinct municipality and computes the total count of such unique entities. This preliminary step is instrumental in the preparation for subsequent visual mapping. The output, which includes both a count and a detailed list of these municipalities, serves as a cornerstone for the accurate plotting of incidents in the following stages of analysis. It ensures that the data fed into the heatmap visualization is well-structured and comprehensive, reflecting the true breadth of the dataset's geographic coverage.

```
In [5]: # Extract unique municipalities from the GME_NAAM column
unique_municipalities = df_accidents['GME_NAAM'].unique()

# Count of unique municipalities
num_unique_municipalities = df_accidents['GME_NAAM'].nunique()

print(f"Number of unique municipalities (GME_NAAM): {num_unique_municipalities}")
print("List of unique municipalities:")
for municipality in unique_municipalities[:5]:
    print(municipality)

Number of unique municipalities:
Amsterdam
Tilburg
's-Hertogenbosch
Súdwest-Fryslân
Smallingerland
```

3.1.2 Coordinates for Municipalities in the Netherlands: Longitude and Latitude Data

Prior to delving into the visualization of accident locations on the map, it is essential to gather accurate geographical coordinates for each municipality in the Netherlands. The dataset provided contains precise longitudinal and latitudinal points, which are crucial for plotting each location on the map. These coordinates serve as the foundational data that will enable the creation of an informative and detailed map, illustrating the spatial distribution of traffic incidents across the municipalities. With these coordinates, each municipality can be accurately represented on the map, providing a clear visual representation of the number of accidents in relation to their geographical context.

```
In [8]: ▶ # this piece of code takes a while to run, be patient
                                                  def get_coordinates(municipality_name, country="Netherlands", max_retries=3, backoff_factor=0.3):
                                                                  base_url = "https://nominatim.openstreetmap.org/search"
                                                                   params = {
                                                                                    'q': f'{municipality_name}, {country}',
                                                                                    'format': 'json',
                                                                                    'limit': 1
                                                                   }
                                                                   session = requests.Session()
                                                                   retry = Retry(total=max_retries, backoff_factor=backoff_factor)
                                                                   adapter = HTTPAdapter(max_retries=retry)
                                                                   session.mount('http://', adapter)
session.mount('https://', adapter)
session.headers.update({'User-Agent': 'YourAppName/1.0'}) # Add your User-Agent here
                                                                                   response = session.get(base_url, params=params, timeout=10)
                                                                                    response.raise_for_status()
                                                                                    data = response.json()
                                                                                    if data:
                                                                                                   lat = float(data[0]['lat'])
                                                                                                    lon = float(data[0]['lon'])
                                                                                                   return lat, lon
                                                                   except requests.exceptions.RequestException as e:
                                                                                   print(f"Error fetching data for {municipality_name}: {e}")
                                                                   return None
                                                  # Sample municipalities
                                                  municipalities = [
                                                                   Capalities = [
"Amsterdam", "Tilburg", "'s-Hertogenbosch", "Súdwest-Fryslân", "Smallingerland",
"Westerkwartier", "Oost Gelre", "Roerdalen", "Horst aan de Maas", "Venlo",
"Breda", "Terneuzen", "Best", "Waddinxveen", "Weststellingwerf",
"De Fryske Marren", "Leeuwarden", "Apeldoorn", "Barneveld", "Hattem",
"Gilze en Rijen", "Rotterdam", "Schiedam", "Oosterhout", "Geertruidenberg",
"Weststell", "Weststell", "Deurstell", "Voorbettell", "Geertruidenberg",
                                                                   "Venray", "Neder-Betuwe", "Deventer", "Overbetuwe", "Nijmegen", "Purmerend", "Zuidplas", "'s-Gravenhage", "Leidschendam-Voorburg", "Weert"
                                                                  "Purmerend", "Zuidplas", "'s-Gravenhage", "Leidschendam-Voorburg", "Weert",
"Echt-Susteren", "Haarlemmermeer", "Rijswijk", "Koggenland", "Kaag en Braassem",
"Moerdijk", "Vlissingen", "Oirschot", "Vught", "Maashorst",
"Opsterland", "Ammersfoort", "Hollands Kroon", "Papendrecht", "Staphorst",
"Waadhoeke", "Arnhem", "Leiden", "Wierden", "Noordoostpolder",
"Eindhoven", "Veldhoven", "Beuningen", "West Betuwe", "Land van Cuijk",
"Heumen", "Utrecht", "Houten", "Utrechtse Heuvelrug", "Oostzaan",
"Beverwijk", "Zaanstad", "Alkmaar", "Lansingerland", "Vlaardingen",
"Midden_Delfland" "Heemskerk" "Doodrecht" "Hoeksche Waard" "Zwiindrecht"
                                                                 "Beverwijk", "Zaanstad", "Alkmaar", "Lansingerland", "Vlaandingen",
"Midden-Delfland", "Heemskerk", "Dordrecht", "Hoeksche Waard", "Zuijndrecht",
"Zoetermeer", "Heerenveen", "Sittard-Geleen", "Heerlen", "Ouder-Amstel",
"Peel en Maas", "Almelo", "Lelystad", "Almere", "Tynaarlo",
"Twenterand", "Wijchen", "Vijfheerenlanden", "Ridderkerk", "Barendrecht",
"Wormerland", "Alblasserdam", "Etten-Leur", "Harderwijk", "Diemen",
"Wassenaar", "Albrandswaard", "Maasgouw", "Maastricht", "Beekdaelen",
"Meerssen", "Leudal", "Raalte", "Eemnes", "De Bilt",
"Woorden", "Zaltbommel", "Ommen", "Zwolle", "Den Helder",
"Borne", "Castricum", "Waalwijk", "Son en Breugel", "Sluis",
"Oss", "Bodegraven-Reeuwijk", "Gouda", "Gooise Meren", "Amstelveen",
                                                               "Woerden", "Zaltbommel", "Ommen", "Zwoile", Den neige, "Sluis", "Borne", "Castricum", "Waalwijk", "Son en Breugel", "Sluis", "Oss", "Bodegraven-Reeuwijk", "Gouda", "Gooise Meren", "Amstelveen", "Hellevoetsluis", "Velsen", "Kampen", "Steenwijkerland", "Rucphen", "Bergen op Zoom", "Enschede", "Oude IJsselstreek", "Alphen aan den Rijn", "Meierijstad", "Geldrop-Mierlo", "Uitgeest", "Cranendonck", "De Ronde Venen", "Sliedrecht", "Tiel", "Culemborg", "Losser", "Veenendaal", "Waalre", "Stichtse Vecht", "Beek", "Bernheze", "Groningen", "Goeree-Overflakkee", "Heusden", "Gennep", "Laren", "Zeist", "Leiderdorp", "Maasdriel", "Valkenburg aan de Geul", "Reimerswaal", "Boxtel", "Westland", "Delft", "Emmen", "Leusden", "Soest", "Gorinchem", "Midden-Groningen", "Hoogeveen", "Kapelle", "Altena", "Halderberge", "Roosendaal", "Woensdrecht", "Eersel", "Aa en Hunze", "Meppel", "Schouwen-Duiveland", "Hardinxveld-Giessendam", "Ede", "Putten", "Hengelo", "Bergen (NH.)", "Zoeterwoude", "Beesel", "Oldambt", "Nunspeet", "Veendam", "Lingewaard", "Blaricum", "Medemblik", "Assen", "Middelburg", "Haaksbergen", "Haarlem", "De Wolden", "Goes", "The steen of the st
                                                                 "Veendam", "Lingewaard", "Blaricum", "Medemblik", "Assen",
"Middelburg", "Haaksbergen", "Haarlem", "De Wolden", "Goes",
"Someren", "Roermond", "Westerveld", "Stein", "Voerendaal",
"Nieuwegein", "Hilvarenbeek", "Schagen", "Epe", "Doetinchem",
"Hendrik-Ido-Ambacht", "Duiven", "Alphen-Chaam", "Heeze-Leende", "Nijkerk",
"Renkum", "Hellendoorn", "Rijssen-Holten", "Buren", "Oldebroek",
"Eemsdelta", "Steenbergen", "Midden-Drenthe", "Zevenaar", "Heerde",
"Hilversum", "Berkelland", "Bladel", "Harlingen", "Baarn",
"Montferland", "Teylingen", "Dinkelland", "Deurne", "Oldenzaal",
"Voorst", "Capelle aan den IJssel", "Coevorden", "Heiloo", "Bunnik",
"Veere", "Rheden", "Ermelo", "Lisse", "Bunschoten",
"Oegstgeest", "Zeewolde", "Brielle", "Elburg", "Edam-Volendam",
"Drimmelen", "Dronten", "Huizen", "Tytsjerksteradiel", "Valkenswaard",
"Oisterwijk", "Hardenberg", "Noord-Beveland", "Woudenberg", "Molenlanden",
"Dalfsen", "Hulst", "Zandvoort", "Tholen", "IJsselstein",
                                                                  "Dalfsen", "Hulst", "Zandvoort", "Tholen", "IJsselstein",
"Katwijk", "Voorburg", "Mook en Middelaar", "Scherpenzeel", "Brunssum",
"Weesp", "Borger-Odoorn", "Goirle", "Grave", "Reusel-De Mierden",
"Noordenveld", "Vaals", "Achtkarspelen", "Loppersum", "Gulpen-Wittem",
"Bronckhorst", "Hoorn", "Eijsden-Margraten"
                                                  1
                                                  # Replace this list with all 253 municipalities
                                                  municipality_coords = {}
                                                  for municipality in municipalities:
```

```
sleep(1) # To comply with rate limiting
coords = get_coordinates(municipality)
if coords:
    municipality_coords[municipality] = coords
print(municipality_coords)
```

{'Amsterdam': (52.3730796, 4.8924534), 'Tilburg': (51.58561845, 5.0660616131367675), "'s-Hertogenbosch": (51.6889387, 5.303116), 'Súdwest-Fryslân': (52.974832649999996, 5.436073892467922), 'Smallingerland': (53.11525265, 6.05073891558769 9), 'Westerkwartier': (53.21425645, 6.3459813379440115), 'Oost Gelre': (52.0153031, 6.561010327461534), 'Roerdalen': (5 1.1454691, 6.034390914431709), 'Horst aan de Maas': (51.4515953, 6.030898893223135), 'Venlo': (51.39244885, 6.151172414122955), 'Breda': (51.5887845, 4.7760237), 'Terneuzen': (51.29884295, 3.8422043956030683), 'Best': (51.5033671, 5.3938 665), 'Waddinxveen': (52.0443437, 4.641596120380745), 'Weststellingwerf': (52.87191835, 6.008674359652279), 'De Fryske Marren': (52.9118448, 5.713876378916552), 'Leeuwarden': (53.2005936, 5.7918548), 'Apeldoorn': (52.17957375, 5.930077460 807249), 'Barneveld': (52.17150665, 5.645439299954028), 'Hattem': (52.476711699999996, 6.061913261141655), 'Gilze en Ri jen': (51.56323905, 4.913039391005126), 'Rotterdam': (51.9244424, 4.47775), 'Schiedam': (51.9300454, 4.38645105261573
1), 'Oosterhout': (51.63223759999996, 4.863042417836681), 'Geertruidenberg': (51.6961662, 4.877843805275408), 'Venra
y': (51.5082754, 5.958901096612693), 'Neder-Betuwe': (51.92150995, 5.616123265710504), 'Deventer': (52.2695736, 6.23633 9604766698), 'Overbetuwe': (51.921961249999995, 5.786081524088322), 'Nijmegen': (51.842574850000005, 5.83896062874822 9), 'Purmerend': (52.5024789, 4.9614827), 'Zuidplas': (51.9995787, 4.613450887425165), "'s-Gravenhage": (52.07494555, 4.2696802205364515), 'Leidschendam-Voorburg': (52.09260105, 4.4321249720751625), 'Weert': (51.235582949999994, 5.705079 711646853), 'Echt-Susteren': (51.08213519999996, 5.899125547533698), 'Haarlemmermeer': (52.3238421, 4.71526124887600 9), 'Rijswijk': (52.0446265, 4.3188215), 'Koggenland': (52.64758125, 4.952321716593174), 'Kaag en Braassem': (52.193046 249999995, 4.621636649034843), 'Moerdijk': (51.66277225, 4.525592262905145), 'Vlissingen': (51.49433865, 3.415005767601 362), 'Oirschot': (51.488961599999996, 5.300594979407693), 'Vught': (51.6511806, 5.23680014410728), 'Maashorst': (51.69 5027705, 5.657493808428191), 'Opsterland': (53.04680105, 6.095135723829035), 'Amersfoort': (52.1562499, 5.3896944), 'Holo lands Kroon': (52.875946799999994, 4.965504712047365), 'Papendrecht': (51.834666999999996, 4.704688255674374), 'Staphorst': (52.6368387, 6.208129727382328), 'Waadhoeke': (53.2460054, 5.610797779708593), 'Arnhem': (52.0056642, 5.8762345077641305), 'Leiden': (52.1594747, 4.4908843), 'Wierden': (52.35195215, 6.567448420567358), 'Noordoostpolder': (52.7252045 5, 5.705372865399351), 'Eindhoven': (51.4392648, 5.478633), 'Veldhoven': (51.4190691, 5.4045572), 'Beuningen': (51.8586 5, 5.705372865399351), 'Eindhoven': (51.4392648, 5.478633), 'Veldhoven': (51.4190691, 5.4045572), 'Beuningen': (51.8586 611, 5.767357), 'West Betuwe': (51.8697052, 5.1941371524390245), 'Land van Cuijk': (51.6635613, 5.851042081979072), 'He umen': (51.78111135, 5.824896568483027), 'Utrecht': (52.080985600000005, 5.12768396945229), 'Houten': (52.03061555, 5.1 61216639426677), 'Utrechtse Heuvelrug': (52.028653649999995, 5.394884269227865), 'Oostzaan': (52.45218065, 4.8685487325 95015), 'Beverwijk': (52.4876371, 4.6603502), 'Zaanstad': (52.4689235, 4.773783522451712), 'Alkmaar': (52.600853799999 96, 4.817099446443963), 'Lansingerland': (52.01101685, 4.507645445233463), 'Vlaardingen': (51.9227863, 4.32956337887085 1), 'Midden-Delfland': (51.97058165, 4.322980951300874), 'Heemskerk': (52.51011775, 4.635060887073173), 'Dordrecht': (51.7958812, 4.6779351), 'Hoeksche Waard': (51.76743725, 4.435615258877212), 'Zwijndrecht': (51.8162227, 4.6405166), 'Zoe termeer': (52.0622531, 4.4901218), 'Heerenveen': (52.998474200000004, 5.923148795038062), 'Sittard-Geleen': (51.009404 3, 5.8376057867167965), 'Heerlen': (50.8775239, 5.981506585454879), 'Ouder-Amstel': (52.2946678, 4.91646890749465), 'Pe el en Maas': (51.33064175, 5.979363705773345), 'Almelo': (52.3567956, 6.66725277634109), 'Lelystad': (52.53668095000000 4, 5.361043525412799), 'Almere': (52.44175125, 5.2092943104327585), 'Tynaarlo': (53.11201305, 6.5598426150845), 'Twente rand': (52.439304, 6.635427575044659), 'Wijchen': (51.82025285, 5.696090348571646), 'Vijfheerenlanden': (51.9302055, 5. rand': (52.439304, 6.635427575044659), 'Wijchen': (51.82025285, 5.696090348571646), 'Vijfheerenlanden': (51.3302055, 5.0495233241373185), 'Ridderkerk': (51.8714812, 4.5999778), 'Barendrecht': (51.85193795, 4.52938352520747), 'Wormerland': (52.5056112000000004, 4.859412783877223), 'Alblasserdam': (51.8599524, 4.666202275442476), 'Etten-Leur': (51.5692065, 4. 6360813), 'Harderwijk': (52.33599915, 5.640271521392062), 'Diemen': (52.3350493, 4.984282926187014), 'Wassenaar': (52.1 4361135, 4.370966818678455), 'Albrandswaard': (51.8524269, 4.435285362646326), 'Maasgouw': (51.1577134, 5.8945473488255 03), 'Maastricht': (50.85798545, 5.6969881818221095), 'Beekdaelen': (50.93449595, 5.88869051100548), 'Meerssen': (50.90 470915, 5.735207828646779), 'Leudal': (51.2339148, 5.895853139985457), 'Raalte': (52.3874327, 6.307848238791079), 'Eemn es': (52.2547415, 5.286070995513654), 'De Bilt': (52.1445592, 5.173777333788463), 'Woerden': (52.08701595, 4.8766871986 73499), 'Zaltbommel': (51.782164, 5.189826737158967), 'Ommen': (52.510346999999996, 6.452424670840507), 'Zwolle': (52.5 14565250000004, 6.097720359677693), 'Den Helder': (52.9529894, 4.8267592494488), 'Borne': (52.3004342, 6.75492254872685 9), 'Castricum': (52.558830549999996, 4.639675526200153), 'Waalwijk': (51.687337150000005, 5.017307939950269), 'Son en Breugel': (51.5173136, 5.500838), 'Sluis': (51.32566285, 3.5166308518864264), 'Oss': (51.778354199999995, 5.53208361093 2192), 'Bodegraven-Reeuwijk': (52.0677553, 4.7575800066386025), 'Gouda': (52.018119350000006, 4.711122134697802), 'Gooi se Meren': (52.322224000000006, 5.104556785563771), 'Amstelveen': (52.2862177, 4.852649259746028), 'Hellevoetsluis': (51.833921, 4.1436739), 'Velsen': (52.451910299999994, 4.603891145874141), 'Kampen': (52.5559484, 5.9033303), 'Steenwijke rland': (52.747874100000004, 6.023206628871991), 'Rucphen': (51.5209462, 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3.1.3 Visual (Graphical) Representation of Traffic Incidents by Municipality in the Netherlands

This analysis explores the distribution and frequency of accidents within The Netherland's municipalities. Data visualization techniques are employed to offer a lucid and informative perspective on traffic safety. The ensuing charts and heatmaps provide insight into the areas with the most reported accidents, pinpointing regions that might require enhanced traffic safety interventions. Such visual data representation not only clarifies the present road safety conditions in the Netherlands but also acts as a valuable resource for policymakers and public safety officials striving to address and reduce traffic-related issues.

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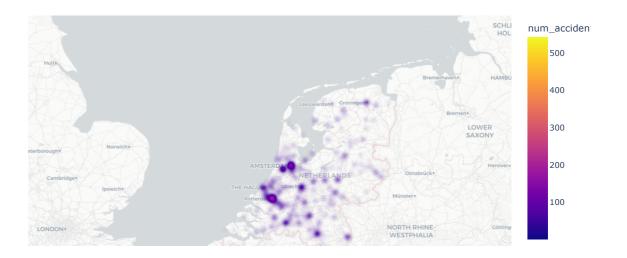
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'Nunspeet': (52.3372909, 5.795636429005031),
```

```
'Veendam': (53.0807517, 6.877303483300453),
'Lingewaard': (51.9074337, 5.9439721314706215),
'Blaricum': (52.28113374999995, 5.29371090566366), 'Medemblik': (52.7907137, 5.16862281678412),
'Assen': (52.9952273, 6.560498),
'Middelburg': (51.4996784, 3.6137371), 'Haaksbergen': (52.1564581, 6.740239),
'Haarlem': (52.3885317, 4.6388048),
'De Wolden': (52.705585049999996, 6.338952403673677),
'Goes': (51.5031033, 3.8897596),
'Someren': (51.38632235, 5.688884477579495),
'Roermond': (51.1933903, 5.9882649),
'Westerveld': (52.828469850000005, 6.311221629926694),
'Stein': (50.9732145, 5.760681940061414),
'Voerendaal': (50.8722592, 5.913755707785212),
'Nieuwegein': (52.0298704, 5.0929333789192395),
'Hilvarenbeek': (51.48698395, 5.146972817481462),
'Schagen': (52.78822125, 4.741356804254988),
'Epe': (52.3270371, 5.9631945020748915),
'Doetinchem': (51.958534799999995, 6.281455126241391),
'Hendrik-Ido-Ambacht': (51.84664995, 4.634280416662918),
'Duiven': (51.9470218, 6.02303247329986),
'Alphen-Chaam': (51.50689565, 4.871359861218567), 'Heeze-Leende': (51.35863595, 5.5779773009269675),
'Nijkerk': (52.214853649999995, 5.478146323965094),
'Renkum': (51.99168495, 5.791692252659134),
'Hellendoorn': (52.39046275, 6.451511548705841),
'Rijssen-Holten': (52.284324850000004, 6.440777380657181),
'Buren': (51.92963705, 5.420000238944224),
'Oldebroek': (52.45673635, 5.951864370961877),
'Eemsdelta': (53.34781255, 6.795656528819859),
'Steenbergen': (51.603292100000004, 4.309251716161476),
'Midden-Drenthe': (52.87061935, 6.568055858657804),
'Zevenaar': (51.927908099999996, 6.077526279085048).
'Heerde': (52.4056303, 6.0413984003658525),
'Hilversum': (52.2241375, 5.1719396),
'Berkelland': (52.10152, 6.618404626883939),
'Bladel': (51.37121075, 5.234630644566941),
'Harlingen': (53.1752103, 5.4141998),
'Baarn': (52.2002549, 5.269143393313942)
'Montferland': (51.92400575, 6.205514601615217),
'Teylingen': (52.21739375, 4.511393222538171),
'Dinkelland': (52.37144879999996, 6.915066619316923),
'Deurne': (51.4281834, 5.828106871124028),
'Oldenzaal': (52.30853685, 6.914603743122237),
'Voorst': (52.2374112, 6.086274393475191),
'Capelle aan den IJssel': (51.9312552, 4.5883596),
'Coevorden': (52.6612281, 6.7407359),
'Heiloo': (52.60192475, 4.70503175175436),
'Bunnik': (52.039151849999996, 5.220319799249577),
'Veere': (51.55640385, 3.577269355747127),
'Rheden': (52.03136695, 6.060504296329905),
'Ermelo': (52.2858356, 5.654938589920705),
'Lisse': (52.2540183, 4.541928029675811),
'Bunschoten': (52.2391743, 5.3734101),
'Oegstgeest': (52.1833669, 4.472129402976295),
'Zeewolde': (52.3451194, 5.4580775473608565),
'Brielle': (51.901662, 4.1626292),
'Elburg': (52.415659149999996, 5.841321708557087), 'Edam-Volendam': (52.54861165, 5.0266156034337826),
'Drimmelen': (51.69768865, 4.743934177567655),
'Dronten': (52.531389649999994, 5.750631211201343), 'Huizen': (52.296484449999994, 5.240582040690262),
'Tytsjerksteradiel': (53.20512135, 5.973447146507526),
'Valkenswaard': (51.3202409, 5.458181897733282),
'Oisterwijk': (51.56610775, 5.198644959051684),
'Hardenberg': (52.56982145, 6.633166913313982),
'Noord-Beveland': (51.56520675, 3.8009095733783114), 'Woudenberg': (52.08025955, 5.4122265761644925),
'Molenlanden': (51.891504350000005, 4.836502624912571),
'Dalfsen': (52.51178365, 6.256636721854065),
'Hulst': (51.33154005, 4.087702848747496),
'Zandvoort': (52.3713394, 4.5307624),
'Tholen': (51.57085015, 4.076142821673509)
'IJsselstein': (52.026205250000004, 5.028685923229064),
'Katwijk': (52.2005078, 4.4146372)
'Voorburg': (52.0703201, 4.3558561)
'Mook en Middelaar': (51.74830635, 5.906406373368311),
'Scherpenzeel': (52.0795901, 5.4904134),
'Brunssum': (50.9431108, 5.9705675),
'Weesp': (52.3072593, 5.0420822),
'Borger-Odoorn': (52.9000925, 6.8821614819282875),
'Goirle': (51.5056266, 5.033774070783131),
'Grave': (51.7596504, 5.7411125),
'Reusel-De Mierden': (51.37099705, 5.1468666691668705),
'Noordenveld': (53.0945705, 6.4209724500257455),
'Vaals': (50.7781016, 5.981067967268041),
'Achtkarspelen': (53.207516749999996, 6.150763705175438),
'Loppersum': (53.3320144, 6.7469502),
'Gulpen-Wittem': (50.79962605, 5.890529675273097),
'Bronckhorst': (52.0529087, 6.269274781349422),
'Hoorn': (52.653272, 5.0735802),
```

```
'Eijsden-Margraten': (50.80499415, 5.768281669696572)
}
# Group by GME NAAM and count the number of accidents
df_grouped = df_accidents.groupby('GME_NAAM').size().reset_index(name='num_accidents')
# Merge with coordinates
 \texttt{df\_grouped['latitude']} = \texttt{df\_grouped['GME\_NAAM']}. \\ \texttt{man}(lambda \ x: \ \texttt{municipality\_coords.get(x, (np.nan, np.nan))[0])} 
df_grouped['longitude'] = df_grouped['GME_NAAM'].map(lambda x: municipality_coords.get(x, (np.nan, np.nan))[1])
# Remove rows with NaN coordinates (municipalities not in the dictionary)
df_grouped.dropna(subset=['latitude', 'longitude'], inplace=True)
# Define a threshold for displaying municipality names
threshold = 100
fig = px.density_mapbox(df_grouped,
                         lat='latitude',
                         lon='longitude'
                         z='num_accidents',
                         radius=10, # Reduced radius for less blurriness
                         center=dict(lat=52.3, lon=5.5), # Centering on Netherlands
                         hover_name='GME_NAAM',
                         title="Number of Accidents by Municipality"
                         color_continuous_scale=px.colors.sequential.Plasma,
                         template="plotly")
fig.update layout(mapbox style="carto-positron")
fig.show()
```

Number of Accidents by Municipality



The image depicts a heatmap of the number of accidents by municipality in the Netherlands. The heatmap is color-coded based on the number of accidents, with the color scale shown on the right side indicating that darker shades of purple represent areas with more accidents, scaling from 1 to over 500.

The map allows for the visual identification of these municipalities through prominent purple spots. Large cities in the Netherlands, such as Rotterdam and Amsterdam, are depicted with large, dark purple bubbles, signifying a high concentration of accidents. Other municipalities are also represented with bubbles, which may appear slightly smaller or lighter in color, corresponding to a fewer number of accidents.

Moving from the visual cues on the heatmap to the empirical data, this table presents a quantitative snapshot of traffic accidents across various municipalities in the Netherlands, offering a focused look at the locations with the highest incidence rates. The data is organized to highlight the number of accidents in descending order, starting with Rotterdam at the forefront, followed by Amsterdam, and trailing down to Apeldoorn. Each entry correlates to a unique municipality, represented by its respective name under the column 'GME_NAAM', alongside the total count of recorded accidents in the 'num_accidents' column. The figures reflect a range of accident frequencies, providing a basis for understanding the distribution of traffic-related incidents in these urban areas.

3
3
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3
4

The table lists the number of traffic accidents in various Dutch municipalities, providing a snapshot of areas with the highest incidence of events. Rotterdam tops the table with 543 reported incidents, marking it as the municipality with the most accidents. Amsterdam is next with 419 accidents, underscoring its status as another focal point for traffic incidents. Further down, Haarlemmermeer, Utrecht, and Breda also report significant numbers, though not as high as the leading cities. Additional municipalities like Eindhoven, 's-Gravenhage, and 's-Hertogenbosch are noted, each with accidents totaling over a hundred, ending with Venlo and Apeldoorn. This information is key to identifying and understanding the patterns of traffic-related incidents in these populated areas.

3.1.4 Visual (Graphical) Representation of Traffic Incidents by Province in the Netherlands

Building on the analysis of traffic incidents at the municipal level, the scope now extends to the examination of accident patterns across the provinces of the Netherlands. The data aggregated at this higher administrative level reveals regional trends and identifies provinces with higher frequencies of traffic-related events. This comparative view highlights the volume of accidents within each province and offers insights into potential underlying factors contributing to regional differences. The ensuing discussion aims to deliver a comprehensive understanding of the distribution of traffic accidents across the Dutch provinces.

```
In [11]: ▶ import pandas as pd
                                         import plotly.express as px
                                                      'North Holland': ['Amsterdam', 'Haarlem', 'Hilversum', 'Purmerend', 'Zandvoort'],
'South Holland': ['Rotterdam', 's-Gravenhage', 'Leiden', 'Delft', 'Schiedam', 's-Gravenzande', 'Zoetermeer', 'Rijswij
'Zeeland': ['Vlissingen', 'Kapelle', 'Schouwen-Duiveland', 'Borsele', 'Noord-Beveland', 'Tholen', 'Veere'],
                                                      'Drenthe':
                                                                  'Aa en Hunze', 'Assen', 'Borger-Odoorn', 'Coevorden', 'De Wolden', 'Emmen', 'Hoogeveen', 'Meppel', 'Midden-Drenthe', 'Noordenveld',
                                                                  'Tynaarlo', 'Westerveld'
                                                      'Flevoland': [
'Almere', 'Dronten', 'Lelystad', 'Noordoostpolder', 'Urk', 'Zeewolde'
                                                        Friesland': [
                                                                  'De Fryske Marren', 'Heerenveen', 'Leeuwarden', 'Opsterland',
'Smallingerland', 'Súdwest-Fryslân', 'Tytsjerksteradiel', 'Waadhoeke'
                                                                 Iderland': [
'Aalten', 'Apeldoorn', 'Arnhem', 'Beuningen', 'Berg en Dal', 'Berkelland',
'Beuningen', 'Buren', 'Culemborg', 'Doesburg', 'Doetinchem', 'Druten',
'Duiven', 'Ede', 'Elburg', 'Epe', 'Ermelo', 'Geldrop-Mierlo', 'Harderwijk',
'Hattem', 'Heerde', 'Heumen', 'Lingewaard', 'Lochem', 'Maasdriel', 'Montferland',
'Nijkerk', 'Nijmegen', 'Nunspeet', 'Oldebroek', 'Oost Gelre', 'Ooststellingwerf',
'Oude IJsselstreek', 'Overbetuwe', 'Putten', 'Renkum', 'Rheden', 'Rijssen-Holten',
'Rozendaal', 'Scherpenzeel', 'Tiel', 'Voorst', 'Wageningen', 'West Betuwe',
'West Maas en Waal', 'Westervoort', 'Wijchen', 'Winterswijk', 'Zaltbommel',
'Zevenaar', 'Gennep', 'Buren', 'Lingewaard', 'Overbetuwe', 'West Betuwe',
'Montferland', 'Oude IJsselstreek', 'Winterswijk', 'Voorst', 'Doetinchem',
'West Maas en Waal', 'Nijkerk', 'Heerde', 'Epe', 'Berkelland', 'Bronckhorst',
'Lochem', 'Oost Gelre', 'Aalten', 'Zutphen'
                                                         Gelderland': [
                                                        'Groningen': [
                                                                  'Eemsdelta', 'Groningen', 'Het Hogeland', 'Midden-Groningen', 'Oldambt',
'Pekela', 'Stadskanaal', 'Veendam', 'Westerkwartier', 'Westerwolde'
                                                      'Limburg': [
    'Beek', 'Beekdaelen', 'Beesel', 'Bergen', 'Brunssum', 'Echt-Susteren',
    'Beek', 'Beekdaelen', 'Beesel', 'Bergen', 'Brunssum', 'Brunssum'
                                                                  'Gulpen-Wittem', 'Heerlen', 'Horst aan de Maas', 'Kerkrade', 'Landgraaf', 'Leudal', 'Maastricht', 'Maasgouw', 'Meerssen', 'Mook en Middelaar', 'Nederweert', 'Peel en Maas', 'Roermond', 'Simpelveld', 'Sittard-Geleen', 'Stein', 'Vaals', 'Valkenburg aan de Geul', 'Venlo', 'Venray', 'Voerendaal',
                                                                  'Weert'
                                                      ],
                                                        Noord-Brabant': [
                                                                  'Alphen-Chaam', 'Altena', 'Baarle-Nassau', 'Hilvarenbeek', 'Goirle', 'Reusel-De Mierden', 'Eersel', 'Bladel', 'Oirschot', 'Best', 'Veldhoven',
                                                                 'Eindhoven', 'Oisterwijk', 'Haaren', 'Tilburg', "'s-Hertogenbosch",
'Meierijstad', 'Sint-Michielsgestel', 'Boxtel', 'Haaren', 'Vught', 'Dongen',
'Loon op Zand', 'Gilze en Rijen', 'Tilburg', 'Gilze en Rijen',
'Geertruidenberg', 'Oosterhout', 'Waalwijk', 'Heusden', 'Altena'
                                                      ٦,
                                                                  'Eemnes', 'De Bilt', 'Stichtse Vecht', 'Baarn', 'Nieuwegein', 'Woudenberg', 'Utrecht', 'Houten', 'Lopik', 'Vijfheerenlanden', 'Woerden', 'Zeist'
                                                     ]
                                         }
                                         # Data
                                                      'Province': ['North Holland', 'South Holland', 'Zeeland', 'Drenthe', 'Flevoland',
                                                      'Friesland', 'Gelderland', 'Groningen', 'Limburg', 'Noord-Brabant', 'Total accidents': [465, 1198, 40, 171, 133, 259, 823, 198, 478, 757, 386]
                                         }
                                         coordinates = {
                                                      'North Holland': (52.6323814, 4.7533755),
'South Holland': (51.9225, 4.47917),
                                                      'Zeeland': (51.4991747, 3.6151451),
                                                      'Drenthe': (52.8775717, 6.6178052),
'Flevoland': (52.5246256, 5.7400598),
'Friesland': (53.1641641, 5.7763946),
                                                      'Gelderland': (52.0583672, 5.5363016),
                                                      'Groningen': (53.2190652, 6.5680077),
                                                      'Limburg': (51.2091958, 5.9925004),
                                                      'Noord-Brabant': (51.5287863, 5.5592763),
                                                      'Utrecht': (52.0809856, 5.1276839)
                                         df = pd.DataFrame(data)
                                         df['Latitude'] = df['Province'].apply(lambda x: coordinates[x][0])
                                         df['Longitude'] = df['Province'].apply(lambda x: coordinates[x][1])
                                          # Create a density_mapbox figure
                                         fig = px.density_mapbox(df,
                                                                                                                   lat='Latitude'.
                                                                                                                   lon='Longitude',
                                                                                                                   z='Total accidents',
                                                                                                                   radius=20, # Adjust radius as needed
                                                                                                                   center=dict(lat=52.3, lon=5.5), # Centering on Netherlands
```

```
zoom=6,
hover_name='Province',
title="Number of Accidents by Province",
color_continuous_scale=px.colors.sequential.Plasma,
template="plotly")

# Update the layout to use a mapbox style
fig.update_layout(mapbox_style="carto-positron")

# Display the figure
fig.show()
```

Number of Accidents by Province



The image displays a heatmap on a map of the Netherlands, showing the distribution of traffic accidents by province. The intensity of the accidents is visualized by color-coded circles, with the color scale on the right indicating the number of accidents. Darker shades of purple signify a higher number of accidents, scaling from 200 to over 1000, as indicated by the color legend labeled "Total accidents."

The largest and darkest bubble is centered over South Holland, reflecting its status as the province with the highest number of accidents. Other provinces with notable accident counts, like North Holland and Gelderland, are marked by slightly smaller, yet still intensely colored bubbles. Provinces with fewer accidents are represented by even smaller and lighter-colored circles. The map provides a clear visual representation of how traffic accidents are distributed across the provinces, highlighting areas with higher and lower frequencies of traffic incidents.

```
In [12]: ▶ # Load accident data from the .pkl file
                         df_accidents = pd.read_pickle("accidents.pkl")
                         # Define the provinces and their municipalities
                                'North Holland': ['Amsterdam', 'Haarlem', 'Hilversum', 'Purmerend', 'Zandvoort'],
'South Holland': ['Rotterdam', 's-Gravenhage', 'Leiden', 'Delft', 'Schiedam', 's-Gravenzande', 'Zoetermeer', 'Rijswij
'Zeeland': ['Vlissingen', 'Kapelle', 'Schouwen-Duiveland', 'Borsele', 'Noord-Beveland', 'Tholen', 'Veere'],
                                        'Aa en Hunze', 'Assen', 'Borger-Odoorn', 'Coevorden', 'De Wolden', 'Emmen', 'Hoogeveen', 'Meppel', 'Midden-Drenthe', 'Noordenveld',
                                        'Tynaarlo', 'Westerveld'
                                |;
'Flevoland': [
                                        'Almere', 'Dronten', 'Lelystad', 'Noordoostpolder', 'Urk', 'Zeewolde'
                                  Friesland': [
                                        'De Fryske Marren', 'Heerenveen', 'Leeuwarden', 'Opsterland',
'Smallingerland', 'Súdwest-Fryslân', 'Tytsjerksteradiel', 'Waadhoeke'
                                       Iderland': [
'Aalten', 'Apeldoorn', 'Arnhem', 'Beuningen', 'Berg en Dal', 'Berkelland',
'Beuningen', 'Buren', 'Culemborg', 'Doesburg', 'Doetinchem', 'Druten',
'Duiven', 'Ede', 'Elburg', 'Epe', 'Ermelo', 'Geldrop-Mierlo', 'Harderwijk',
'Hattem', 'Heerde', 'Heumen', 'Lingewaard', 'Lochem', 'Maasdriel', 'Montferland',
'Nijkerk', 'Nijmegen', 'Nunspeet', 'Oldebroek', 'Oost Gelre', 'Ooststellingwerf',
'Oude IJsselstreek', 'Overbetuwe', 'Putten', 'Renkum', 'Rheden', 'Rijssen-Holten',
'Rozendaal', 'Scherpenzeel', 'Tiel', 'Voorst', 'Wageningen', 'West Betuwe',
'West Maas en Waal', 'Westervoort', 'Wijchen', 'Winterswijk', 'Zaltbommel',
'Zevenaar', 'Gennep', 'Buren', 'Lingewaard', 'Overbetuwe', 'West Betuwe',
'Montferland', 'Oude IJsselstreek', 'Winterswijk', 'Voorst', 'Doetinchem',
'West Maas en Waal', 'Nijkerk', 'Heerde', 'Epe', 'Berkelland', 'Bronckhorst',
'Lochem', 'Oost Gelre', 'Aalten', 'Zutphen'
                                  Gelderland': [
                                 'Groningen': [
                                        'Eemsdelta', 'Groningen', 'Het Hogeland', 'Midden-Groningen', 'Oldambt',
'Pekela', 'Stadskanaal', 'Veendam', 'Westerkwartier', 'Westerwolde'
                                 'Limburg': [
                                        'Beek', 'Beekdaelen', 'Beesel', 'Bergen', 'Brunssum', 'Echt-Susteren',
                                        'Gulpen-Wittem', 'Heerlen', 'Horst aan de Maas', 'Kerkrade', 'Landgraaf',
'Leudal', 'Maastricht', 'Maasgouw', 'Meerssen', 'Mook en Middelaar',
                                        'Nederweert', 'Peel en Maas', 'Roermond', 'Simpelveld', 'Sittard-Geleen', 'Stein', 'Vaals', 'Valkenburg aan de Geul', 'Venlo', 'Venray', 'Voerendaal',
                                        'Weert'
                                  Noord-Brabant': [
                                        'Alphen-Chaam', 'Altena', 'Baarle-Nassau', 'Hilvarenbeek', 'Goirle', 'Reusel-De Mierden', 'Eersel', 'Bladel', 'Oirschot', 'Best', 'Veldhoven',
                                       'Eindhoven', 'Oisterwijk', 'Haaren', 'Tilburg', "'s-Hertogenbosch", 'Meierijstad', 'Sint-Michielsgestel', 'Boxtel', 'Haaren', 'Vught', 'Dongen', 'Loon op Zand', 'Gilze en Rijen', 'Tilburg', 'Gilze en Rijen', 'Geertruidenberg', 'Oosterhout', 'Waalwijk', 'Heusden', 'Altena'
                                  Utrecht': [
                                        'Eemnes', 'De Bilt', 'Stichtse Vecht', 'Baarn', 'Nieuwegein', 'Woudenberg', 'Utrecht', 'Houten', 'Lopik', 'Vijfheerenlanden', 'Woerden', 'Zeist'
                                ]
                         }
                         # Create a dictionary to store the total accidents per province
                         total_accidents_by_province = {}
                         # Iterate over provinces and calculate the total accidents
                         for province_name, municipalities in provinces.items():
                                filtered_data = df_accidents[df_accidents['GME_NAAM'].isin(municipalities)]
                                total_accidents = filtered_data.shape[0]
                                total_accidents_by_province[province_name] = total_accidents
                         # Sort the total accidents by province in descending order
                         sorted_accidents_by_province = sorted(total_accidents_by_province.items(), key=lambda item: item[1], reverse=True)
                         # Print the sorted total accidents for each province
                         for province, accidents in sorted_accidents_by_province:
                                print(f"Total accidents in {province}: {accidents}")
```

```
Total accidents in South Holland: 1198
Total accidents in Gelderland: 823
Total accidents in Noord-Brabant: 757
Total accidents in Limburg: 478
Total accidents in North Holland: 465
Total accidents in Utrecht: 386
Total accidents in Friesland: 259
Total accidents in Groningen: 198
Total accidents in Drenthe: 171
Total accidents in Flevoland: 133
Total accidents in Zeeland: 40
```

The data provided lists the number of traffic accidents recorded across various provinces in the Netherlands, ordered from the highest to the lowest number of incidents. South Holland tops the list with 1198 accidents, indicating it has the highest occurrence of traffic incidents among the provinces. Gelderland follows with 823 accidents, and Noord-Brabant is close behind with 757, suggesting these are also high-incident areas.

Limburg and North Holland are in the middle range, with 478 and 465 accidents respectively, while Utrecht, Friesland, and Groningen show fewer incidents, ranging from 386 to 198. Drenthe and Flevoland have even lower numbers, with 171 and 133 accidents, suggesting a lower rate of traffic-related issues. Zeeland has the fewest reported accidents, with only 40, which could indicate it is the safest among the listed provinces in terms of traffic incidents or it has lower traffic density.

3.2 Analysis for Subquestion 1: Comparison of Traffic Accidents in Rural and Urban Area

3.2.1 Clustering Analysis

Clustering analysis is performed to classify the cities / municipality into two groups, urban and rural area. We use two variables as the clustering factors, they are population density and road length, because urban area is usually identified with higher population density and better infrastructure development (represented by the road length variable) compared to rural area. This analysis is needed because higher density area with better infrastructure will have more traffic movements, and eventually resulted in higher number of accidents.

Out[13]:

	Municipality_Name	Population	Area	Density	Road_Length	Accident_Count
0	Aa en Hunze	25579	276.66	93	639	21
1	Almere	32452	20.00	1615	157	63
2	Alblasserdam	20087	8.87	2290	82	20
3	Albrandswaard	25934	21.17	1197	151	3
4	Alkmaar	110783	110.01	1004	594	28
229	Zoetermeer	125767	34.43	3653	406	36
230	Zoeterwoude	9302	21.16	440	63	8
231	Zuidplas	45794	57.76	790	316	59
232	Zwijndrecht	44789	20.01	2206	209	13
233	Zwolle	130668	110.07	1181	778	54

234 rows × 6 columns

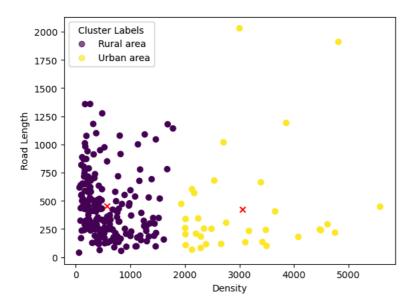
The clustering process is conducted by plotting all cities in two axis chart, with Density as X-axis and Road Length as Y-axis. Two groups of city were formed, the purple ones with lower population density is named **Rural area** and the yellow ones with higher density is labeled as **Urban area**. Meanwhile, there is no remarkable difference in the number of road length between the two groups. Then, the cluster name is added to the existing dataframe.

```
In [14]: k = 2 # The number of clusters
              kmeans = KMeans(n_clusters=k)
              # Select numeric columns for clustering
              numeric_data = df_merged_data[['Density', 'Road_Length']].values
              kmeans.fit(numeric_data)
              labels = kmeans.labels_ # This assigns each data point to a cluster
              centers = kmeans.cluster_centers_ # These are the cluster centroids
              # Map cluster labels to desired names
              label_names = {0: 'Rural area', 1: 'Urban area'}
              label_names_list = [label_names[label] for label in labels]
              cluster_names = [label_names[label] for label in labels]
              # Plotting
              scatter = plt.scatter(numeric_data[:, 0], numeric_data[:, 1], c=labels, cmap='viridis', label=label_names_list)
              plt.scatter(centers[:, 0], centers[:, 1], c='red', marker='x', label='Cluster Centers')
              # Add labels to the clusters in the legend
              handles, labels = scatter.legend_elements(prop="colors", alpha=0.6)
              legend1 = plt.legend(handles, [label_names[0], label_names[1]], title='Cluster Labels')
plt.gca().add_artist(legend1) # Add the Legend to the existing plot
              plt.xlabel('Density')
              plt.ylabel('Road Length')
              plt.show()
              # Add the 'Cluster_Name' column to the DataFrame
              df_merged_data['Cluster_Name'] = cluster_names
              df merged data
```

C:\Users\LENOVO YOGA\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1412: FutureWarning:

The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning

KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can a void it by setting the environment variable OMP_NUM_THREADS=1.



Out[14]:

	Municipality_Name	Population	Area	Density	Road_Length	Accident_Count	Cluster_Name
0	Aa en Hunze	25579	276.66	93	639	21	Rural area
1	Almere	32452	20.00	1615	157	63	Rural area
2	Alblasserdam	20087	8.87	2290	82	20	Urban area
3	Albrandswaard	25934	21.17	1197	151	3	Rural area
4	Alkmaar	110783	110.01	1004	594	28	Rural area
229	Zoetermeer	125767	34.43	3653	406	36	Urban area
230	Zoeterwoude	9302	21.16	440	63	8	Rural area
231	Zuidplas	45794	57.76	790	316	59	Rural area
232	Zwijndrecht	44789	20.01	2206	209	13	Urban area
233	Zwolle	130668	110.07	1181	778	54	Rural area
 229 230 231 232	Zoetermeer Zoeterwoude Zuidplas Zwijndrecht	125767 9302 45794 44789	34.43 21.16 57.76 20.01	3653 440 790 2206	 406 63 316 209	 36 8 59	Urban area Rural area Rural area Urban area

234 rows × 7 columns

3.2.2 T-test analysis

Rural area consists of 199 out of 234 cities in the data, while the rest is classified as urban area. The descriptive statistics data shows that urban area has higher average number of accidents compared to the rural area. To test whether this difference is statistically significant or not, we will perform a T-test analysis. T-test method is commonly used to check the significance of difference between the average of two populations. In this case, the two populations are urban and rural area.

The T-test results show that the higher number of accidents owned by urban area is statistically significant in comparison to rural area. Hence, in order to answer the research question, the identification of weather's impact to the number of accidents have to be performed separately for each city group.

```
In [15]: ▶ # Give the name for the table
                                  print('\033[1mDescriptive Statistics Data for each City Group created\033[0m')
                                  # Descriptive Data: ACCIDENT COUNT
                                  accident_descriptive = df_merged_data.groupby('Cluster_Name')['Accident_Count'].describe()
                                  accident_descriptive = accident_descriptive.rename(columns={'count': 'Count', 'mean': 'Mean', 'std': 'Std', 'min': 'Min',
                                  # Descriptive Data: POPULATION DENSITY
                                  density_descriptive = df_merged_data.groupby('Cluster_Name')['Density'].describe()
                                  density_descriptive = density_descriptive.rename(columns={'count': 'Count', 'mean': 'Mean', 'std': 'Std', 'min': 'Min',
                                  # Descriptive Data: ROAD LENGTH
                                  roadlength_descriptive = df_merged_data.groupby('Cluster_Name')['Road_Length'].describe()
                                  roadlength_descriptive = roadlength_descriptive.rename(columns={'count': 'Count', 'mean': 'Mean', 'std': 'Std', 'min': 'Nean', 'mean': 'Mean', 'std': 'Std', 'min': 'Nean', 'std': 'Std': 'Std', 'min': 'Nean', 'std': 'Std'
                                  combined_descriptive = pd.concat([accident_descriptive, density_descriptive, roadlength_descriptive], keys=['ACCIDENT COL
                                  # Format 'Mean' and 'Std' columns to display one digit after the decimal point
                                 combined_descriptive['Mean'] = combined_descriptive['Mean'].round(1)
                                  combined_descriptive['Std'] = combined_descriptive['Std'].round(1)
                                  # Display the combined DataFrame
                                  combined_descriptive
```

Descriptive Statistics Data for each City Group created

Out[15]:

		Count	Mean	Std	Min	25%	50% (Med)	75%	Max
	Cluster_Name								
ACCIDENT COUNT	Rural area	199.0	25.3	28.1	1.0	9.0	18.0	32.0	246.0
	Urban area	35.0	62.3	111.7	1.0	13.5	31.0	55.5	543.0
POPULATION DENSITY	Rural area	199.0	575.4	417.3	63.0	244.0	432.0	790.5	1783.0
	Urban area	35.0	3049.3	1000.6	1934.0	2226.0	2708.0	3574.5	5576.0
ROAD LENGTH	Rural area	199.0	451.2	283.8	40.0	245.0	374.0	589.0	1359.0
	Urban area	35.0	424.3	461.4	67.0	181.5	252.0	460.5	2030.0

```
In [16]: ► from scipy import stats
             # Print hypothesis to test and confidence level used
             print('\033[1mHYPOTHESIS\033[0m')
             print('HO: Average road accident in Urban area = Average road accident in Rural area')
             print('H1: Average road accident in Urban area # Average road accident in Rural area')
             print('\n')
             print('\033[1mCONFIDENCE LEVEL\033[0m')
             print('Alpha = 5%')
             print('\n')
             # Creating two population of data: Urban and Rural area
             urban_data = merged_data[df_merged_data['Cluster_Name'] == 'Urban area']
             rural_data = merged_data[df_merged_data['Cluster_Name'] == 'Rural area']
             # Extract the 'Accident Count' data for each group
             urban accident count = urban data['Accident Count']
             rural_accident_count = rural_data['Accident_Count']
             # Calculate average and standard deviation for each group
             urban_avg = urban accident count.mean()
             rural_avg = rural_accident_count.mean()
             # Perform an independent two-sample t-test
             t stat, p value = stats.ttest ind(urban accident count, rural accident count)
             # Print results
             print('\033[1mT-TEST RESULTS\033[0m')
             print(f"T-statistic: {t_stat:.2f}")
             print(f"P-value: {p_value:.5f}")
             print('\n')
             # Determine whether the difference is statistically significant (using confidence Level = 0.05)
             print('\033[1mCONCLUSION\033[0m')
             if p_value < 0.05:</pre>
                 print("P_value < alpha, then H0 is rejected.")</pre>
             else:
                 print("P_value > alpha, then H0 is accepted.")
             if p value < 0.05:
                 print("This hypothesis: 'Average road accident in Urban area ≠ Average road accident in Rural area' is accepted.")
                 print("This hypothesis: 'Average road accident in Urban area = Average road accident in Rural area' is accepted.")
             if urban_avg > rural_avg:
                 print("Then, it's concluded that \033[1mUrban area has statistically higher average of road accident than Rural area.
                 print("Then, it's concluded that \033[1mRural area has statistically higher average of road accident than Urban area.
             4
             HYPOTHESIS
             HO: Average road accident in Urban area = Average road accident in Rural area
             H1: Average road accident in Urban area ≠ Average road accident in Rural area
             CONFIDENCE LEVEL
             Alpha = 5%
             T-TEST RESULTS
             T-statistic: 4.04
             P-value: 0.00007
             CONCLUSION
             P_value < alpha, then H0 is rejected.
             This hypothesis: 'Average road accident in Urban area ≠ Average road accident in Rural area' is accepted.
             Then, it's concluded that Urban area has statistically higher average of road accident than Rural area.
```

3.3 Analysis for Subquestion 2: Influence of Weather Type to the Amount of Accidents

The impact of weather type to the amount of accidents is identified by comparing the number of accidents for each weather type. If one or more weather type have different number of accidents (either lower or higher) compared to other weather type, then it's statistically safe to say that the weather type does have impact to the accidents.

The analysis is performed by using ANOVA method. This statistical test is commonly used to identify is there any significant difference between the average of more than two populations. And in this case, we have five populations that represent all types of weather listed in the data (Dry, Rain, Snow, Wind, Fog and Unknown). The ANOVA test is done twice, for urban and rural area, because they have significantly different average number of accidents (based on the result of T-test analysis).

For the analysis, we need a new dataset that contains municipality, weather type, accident count and cluster name. So, we have to create an aggregated accident dataset per municipality per weather type, and then merge it with the area clustering dataset (output of clustering analysis).

The ANOVA test will compare the average of **accident percentage** between each weather type instead of the average of **absolute accident number**. Hence, we need to form an additional column named *Accident_Percentage* for the analysis.

```
In [17]: ▶ # Creating new dataset from df_accidents which aggregating tne number of accident by weather_type
             df_accident_weather_agg = df_accidents.groupby(['Municipality','Weather_Type']).size().reset_index(name='Accident_Count')
             df_accident_weather_agg = df_accident_weather_agg.sort_values(by=['Municipality','Weather_Type'])
             # Get the filtered cluster name data per municipality
             df_clustered_area = df_merged_data[['Municipality_Name','Cluster_Name']]
             # Merge the two DataFrames on the 'Municipality' and 'Municipality_Name' columns
             merged_data_2 = df_accident_weather_agg.merge(df_clustered_area, left_on='Municipality', right_on='Municipality_Name', ho
             # Drop the redundant 'Municipality_Name' column
             df_cluster_weather_data = merged_data_2.drop(columns='Municipality_Name')
             df_cluster_weather_data = df_cluster_weather_data.sort_values(by=["Municipality", 'Weather_Type'])
             df_cluster_weather_cleaned = df_cluster_weather_data.dropna()
             # Create new column for accident percentage
             total accident counts = df cluster weather cleaned.groupby('Municipality')['Accident Count'].transform('sum')
             # Calculate the accident percentage and add it to the dataframe
             df_cluster_weather_cleaned['Accident_Percentage'] = (df_cluster_weather_cleaned['Accident_Count'] / total_accident_counts
             df_cluster_weather_cleaned
```

C:\Users\LENOVO YOGA\AppData\Local\Temp\ipykernel_4256\2197853070.py:21: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

Out[17]:

	Municipality	Weather_Type	Accident_Count	Cluster_Name	Accident_Percentage
6	Aa en Hunze	D	18	Rural area	85.714286
7	Aa en Hunze	R	3	Rural area	14.285714
8	Alblasserdam	D	20	Urban area	100.000000
9	Albrandswaard	D	2	Rural area	66.666667
10	Albrandswaard	R	1	Rural area	33.333333
576	Zuidplas	М	1	Rural area	1.694915
577	Zuidplas	R	4	Rural area	6.779661
578	Zwijndrecht	D	13	Urban area	100.000000
579	Zwolle	D	46	Rural area	85.185185
580	Zwolle	R	8	Rural area	14.814815

525 rows × 5 columns

The ANOVA results show that there are significant difference of accident percentage between weather type groups, both in urban area as well as rural area. Dry weather has highest accident percentage, followed by Fog, Rain, Wind and Snow in Rural area. Meanwhile, in Urban area, Dry weather also has the highest accident percentage, but then followed by Rain, Snow, Wind and Fog in the last.

```
In [18]: ▶ # Print hypothesis to test and confidence level used
              print('\033[1mHYPOTHESIS\033[0m')
              print('HO: There is no difference of the average of accident percentage among weather type.')
              print('H1: There is significant difference of the average of accident percentage among weather type.')
              print('\n')
              print('\033[1mCONFIDENCE LEVEL\033[0m')
              print('Alpha = 5%')
              print('\n')
              # Define the dictionary to map old values to new values
              weather_name = {'D': 'Dry', 'R': 'Rain', 'M': 'Fog', 'S': 'Snow', 'H': 'Wind', 'O': 'Unknown'}
              # Apply the mapping to the 'Weather_Type' column
              df_cluster_weather_cleaned['Weather_Type'] = df_cluster_weather_cleaned['Weather_Type'].replace(weather_name)
              # Create subsets of the data for each cluster
              cluster names = df cluster weather cleaned['Cluster Name'].unique()
              print('\033[1mANOVA TEST RESULTS AND CONCLUSION\033[0m')
              for cluster name in cluster names:
                  subset = df_cluster_weather_cleaned[df_cluster_weather_cleaned['Cluster_Name'] == cluster_name]
                 # Perform one-way ANOVA for each cluster
                  unique_weather_types = sorted(subset['Weather_Type'].unique()) # Sort weather types alphabetically
weather_groups = [subset[subset['Weather_Type'] == weather]['Accident_Percentage'] for weather in subset['Weather_Type']
                  # Perform the ANOVA test
                  f_statistic, p_value = f_oneway(*weather_groups)
                  # Calculate and display the average for each weather type
                  averages = [f'{group.mean():.2f}' for group in weather_groups]
                  \# Determine whether the difference is statistically significant (using confidence Level = 0.05)
                  print(f"\033[1mCluster Name: {cluster_name.upper()}\033[0m")
                  print("Average number of accident percentage per weather type:")
                  sorted\_weather\_averages = sorted(zip(unique\_weather\_types, averages), key=lambda \ x: \ x[\emptyset])
                  for weather_type, avg in sorted_weather_averages:
                                 {weather_type} = {avg}")
                  print(f"P-value: {p_value:.5f}")
                  # print t-value as well
                  print(f"\033[1mConclusion:\033[0m")
                  if p_value < 0.05:</pre>
                      print("P_value < alpha, then H0 is rejected.")</pre>
                      print("P_value > alpha, then H0 is accepted.")
                  if p value < 0.05:
                      print(f"Then, it's concluded that there is \033[1msignificant difference of the average of accident percentage am
                  else:
                      print(f"Then, it's concluded that there is \033[1mno significant difference of the average of accident percentage
                  print()
```

```
HYPOTHESIS
HO: There is no difference of the average of accident percentage among weather type.
H1: There is significant difference of the average of accident percentage among weather type.
CONFIDENCE LEVEL
Alpha = 5%
ANOVA TEST RESULTS AND CONCLUSION
Cluster Name: RURAL AREA
Average number of accident percentage per weather type:
  Dry = 84.91
  Fog = 13.75
  Rain = 6.54
  Snow = 3.18
  Wind = 5.78
P-value: 0.00000
Conclusion:
P value < alpha, then H0 is rejected.
Then, it's concluded that there is significant difference of the average of accident percentage among weather type.
Cluster Name: URBAN AREA
Average number of accident percentage per weather type:
  Drv = 89.49
  Fog = 0.75
  Rain = 11.98
  Snow = 2.69
  Wind = 1.30
P-value: 0.00000
Conclusion:
P value < alpha, then H0 is rejected.
Then, it's concluded that there is significant difference of the average of accident percentage among weather type.
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a
-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versus-a-co
```

3.4 Analysis for Subquestion 3: Influence of Weather Type to the Severity of Accidents

We will use chi-square test to examine the impact weather to accident severity. This method is chosen because it allows us to identify the correlation between two categorical groups of data. In this case, both weather type and accident severity data are categorical.

The results show that there is no significant association / correlation between weather type and accident severity. It means the number of accidents is equally spread across all severity level regardless the weather type.

```
In [19]: W # Creating new dataset that aggregating accidents by Weather_Type and Severity
df_weather_severity_agg = df_accidents.groupby(['Weather_Type','Severity']).size().reset_index(name='Accident_Count')
df_weather_severity_agg = df_weather_severity_agg.sort_values(by=['Weather_Type','Severity'])

# Create a dictionary for abbreviation contained in the data
weather_name = {'D': 'Dry', 'R': 'Rain', 'M': 'Fog', 'S': 'Snow', 'H': 'Wind', 'O': 'Unknown'}
severity = {'DDD': 'Deadly', 'LET': 'Injury', 'UMS': 'Material damage'}

# Apply the mapping to the 'Weather_Type' column
df_weather_severity_agg['Weather_Type' af_weather_severity_agg['Weather_Type'].replace(weather_name)
df_weather_severity_agg['Severity'] = df_weather_severity_agg['Severity'].replace(severity)

# Use pivot to create the matrix with 'Weather_Type' as rows and 'Severity' as columns
matrix = df_weather_severity_agg.pivot(index='Weather_Type', columns='Severity', values='Accident_Count').fillna(0)
matrix
```

Out[19]:

py)

Weather_Type						
Dry	44.0	870.0	5771.0			
Fog	0.0	3.0	35.0			
Rain	0.0	85.0	766.0			
Snow	0.0	11.0	100.0			
Wind	0.0	1.0	18.0			

Severity Deadly Injury Material damage

```
In [20]: ▶ # Print hypothesis to test and confidence level used
              print('\033[1mHYPOTHESIS\033[0m')
              print('HO: Accident severity and Weather type have no association.')
              print('H1: Accident severity and Weather type have significant association.')
              print('\n')
              print('\033[1mCONFIDENCE LEVEL\033[0m')
              print('Alpha = 5%')
              print('\n')
              # Creating new dataset that aggregating accidents by Weather_Type and Severity
              df_weather_severity_agg = df_accidents.groupby(['Weather_Type', 'Severity']).size().reset_index(name='Accident_Count')
              df_weather_severity_agg = df_weather_severity_agg.sort_values(by=['Weather_Type','Severity'])
              # Create a dictionary for abbreviation contained in the data
              weather_name = {'D': 'Dry', 'R': 'Rain', 'M': 'Fog', 'S': 'Snow', 'H': 'Wind', 'O': 'Unknown'}
severity = {'DOD': 'Deadly', 'LET': 'Injury', 'UMS': 'Material damage'}
              # Apply the mapping to the 'Weather_Type' column

df_weather_severity_agg['Weather_Type'] = df_weather_severity_agg['Weather_Type'].replace(weather_name)

df_weather_severity_agg['Severity'] = df_weather_severity_agg['Severity'].replace(severity)
              # Create a contingency table for 'Weather_Type' and 'Severity'
              contingency_table = pd.crosstab(df_weather_severity_agg['Weather_Type'], df_weather_severity_agg['Severity'])
              # Perform the chi-square test of independence
              chi2, p_value, _, _ = chi2_contingency(contingency_table)
              # Determine whether the difference is statistically significant (using confidence level = 0.05)
              print('\033[1mCHI-SQUARE TEST RESULTS\033[0m')
              print(f"Chi-square statistic: {chi2:.2f}")
              print(f"P-value: {p_value:.4f}")
              print('\n')
              print('\033[1mCONCLUSION\033[0m')
              if p_value < 0.05:</pre>
                  print("P_value < alpha, then H0 is rejected.")</pre>
              else:
                  print("P_value > alpha, then H0 is accepted.")
              if p value < 0.05:
                  print("Then, it's concluded that there is \033[1msignificant association between Weather Type and Accident Severity.\
                  print("Then, it's concluded that there is \033[1mno significant association between Weather Type and Accident Severit
              4
              HYPOTHESIS
              HO: Accident severity and Weather type have no association.
              H1: Accident severity and Weather type have significant association.
              CONFIDENCE LEVEL
              Alpha = 5%
              CHI-SQUARE TEST RESULTS
              Chi-square statistic: 2.93
              P-value: 0.9385
              CONCLUSION
              P value > alpha, then H0 is accepted.
              Then, it's concluded that there is no significant association between Weather Type and Accident Severity.
```

5. Conclusion

The subquestions have all been answered through different analyses. First, a T-test was performed to provide an answer for subquestion 1. From this, it was concluded that on average more accidents happen in rural areas compared to urban areas. After this, a One-way ANOVA test was conducted to provide an answer for subquestion 2. The conclusion from this test was that there is a significant difference between different weather types influencing the amount of accidents. Finally, subquestion 3 was answered by performing a Chi-Square Analysis. The result from this analysis is that there is no significant association between the weather type and the accident severity. By answering all subquestions, we can conclude that weather does have impact on the amount of traffic accidents, but not on the severity in both type of areas (urban and rural) in Netherlands during the period of 2021-2022.

6. Contribution Statement

Alex van Wijngaarden:

- Writing narrative for introduction part
- Writing narrative for conclusion part

Eka Devi Wulandari:

• Data searching and processing for dataset 2 (density and road length data)

- · Performing Clustering analysis, T-test, ANOVA and Chi-square test
- · Writing narrative for the statistical analysis part

Nadhira Zahrani Widiafina:

- · Municipality longtitude ana latitude data
- Municipality Heatmap Visualization
- Province Heatmap Visualization
- Writing narrative for heatmap visualization (municipality and province)

Simon, van der Horst:

- Data searching and processing for dataset 1 (traffic accident data)
- · Compiling all reports into 1 file

Wiet Kalf:

- · Province and municipality's accident heatmap
- · Report finalization

Everyone:

- · Background research
- · Research conceptualization

7. Data Source

Dataset 2: Additional data (population, area, density and road length) needed for the clustering analysis from CBS and wikipedia. Here are the links:

- density and area data: https://en.wikipedia.org/wiki/Municipalities of the Netherlands
 (https://en.wikipedia.org/wiki/Municipalities of the Netherlands
- length of road data: https://opendata.cbs.nl/statline/#/CBS/nl/dataset/70806ned/table?ts=1699173969758)
 (https://opendata.cbs.nl/statline/#/CBS/nl/dataset/70806ned/table?ts=1699173969758)