Global Plastic Waste Report

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0.1 Executive Summary

This report investigates global plastic waste patterns, focusing on 165 countries with varying recycling practices and environmental risks. We explore how per capita plastic waste and national recycling rates relate to coastal pollution vulnerability. Using figures and tables, we highlight the countries with the most plastic intensity and evaluate systemic differences. The findings offer useful guidance for policy improvements and environmental planning worldwide.

0.2 Introduction

Plastic waste is a growing global concern, particularly for countries near coastlines. Poor waste management not only pollutes land ecosystems but also threatens marine environments. This report aims to analyze patterns in plastic waste production, recycling performance, and coastal pollution risk using a recent global dataset. The dataset includes 165 countries and features metrics such as total plastic waste, per capita generation, and recycling rates.

By comparing countries based on their waste intensity and risk levels, we seek to understand which factors contribute to environmental vulnerability. We use both descriptive and comparative methods, and summarize our findings through visualizations and aggregated statistics. Our focus is on highlighting regional disparities and identifying outliers.

The purpose of this analysis is to provide actionable insights for decision-makers in environmental and public policy sectors. The report contributes to a broader understanding of how waste systems interact with geography, policy, and infrastructure.

0.3 Methodology

```
library(tidyverse)
library(janitor)
library(knitr)
```

The analysis was conducted using the "Plastic Waste Around the World" dataset, which contains country-level data on plastic waste generation, recycling rates, and coastal vulnerability. The dataset was imported from a GitHub source and processed in R using reproducible and version-controlled workflows.

Upon loading the dataset, we conducted several preprocessing steps:

- 1. **Variable Cleaning and Transformation:** All column names were standardized using 'janitor::clean_names()'. Variables with missing values, particularly 'per_capita_waste_kg' and 'coastal_waste_risk', were excluded from analysis. The continuous variable 'recycling_rate' was binned into three categories: "Low", "Medium", and "High" recycling performance, using quantile-based thresholds.
- 2. **Derived Metrics**: We introduced a new variable, 'total_waste_per_person_mt', calculated by dividing 'total_plastic_waste_mt' by an estimated population derived from total waste and per capita rates. This derived variable enables us to assess plastic production intensity across countries more consistently.
- 3. **Risk-based Grouping**: Countries were grouped by 'coastal_waste_risk' (Low, Medium, High) and stratified by binned recycling performance. This allowed for a multifaceted analysis of waste management practices in different geographic and policy contexts.

```
# Load and clean
plastic_data <- read_csv("Data/Plastic Waste Around the World.csv") %>%
  clean_names() %>%
 filter(!is.na(per_capita_waste_kg), !is.na(coastal_waste_risk), !is.na(recycling_rate))
# Bin recycling rates
plastic_data <- plastic_data %>%
 mutate(
   recycling_category = ntile(recycling_rate, 3),
   recycling_category = case_when(
     recycling_category == 1 ~ "Low",
     recycling_category == 2 ~ "Medium",
     recycling_category == 3 ~ "High"
   )
 )
# Derived variable: total waste per person (in MT)
plastic_data <- plastic_data %>%
 mutate(
   est_population_million = total_plastic_waste_mt * 1e6 / per_capita_waste_kg,
   waste_per_person_mt = total_plastic_waste_mt / est_population_million
 )
```

4. Visualization: We created two key outputs:

Figure 1 : A bar chart of the top 10 countries by per capita plastic waste, with reordered factors and enhanced labels.

```
top10_enhanced <- plastic_data %>%
    arrange(desc(per_capita_waste_kg)) %>%
    slice(1:10)

ggplot(top10_enhanced, aes(x = reorder(country, per_capita_waste_kg), y = per_capita_waste_l
    geom_col() +
    coord_flip() +
    scale_fill_manual(
        values = c("Low" = "red", "Medium" = "gold", "High" = "green")
    ) +
    labs(
        title = "Top 10 Countries by Per Capita Plastic Waste",
        x = "Country",
        y = "Plastic Waste (kg per person)",
        fill = "Recycling Category"
    ) +
    theme_minimal()
```

Top 10 Countries by Per Capita Plastic Waste

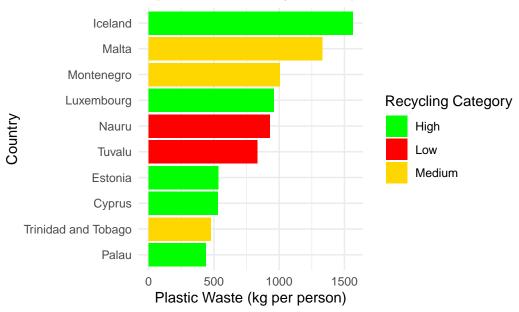
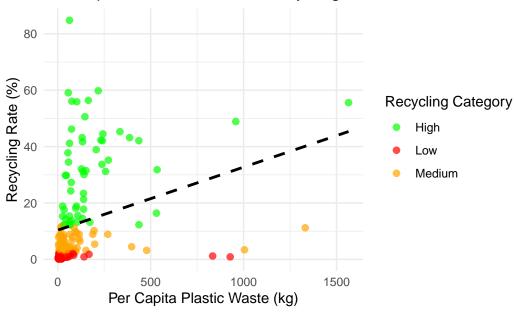


Figure 2: A scatter plot showing the relationship between per capita plastic waste and recycling rate across countries. Points are colored by recycling category (Low, Medium, High), with a linear trend line overlaid. This visual highlights patterns in recycling effectiveness relative to plastic consumption intensity.

```
plastic_data <- plastic_data %>%
  mutate(
    recycling_category = ntile(recycling_rate, 3),
    recycling_category = case_when(
        recycling_category == 1 ~ "Low",
        recycling_category == 2 ~ "Medium",
        recycling_category == 3 ~ "High"
    )
```

```
ggplot(plastic_data, aes(x = per_capita_waste_kg, y = recycling_rate)) +
  geom_point(aes(color = recycling_category), size = 2, alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linetype = "dashed") +
  scale_color_manual(
    values = c("High" = "green", "Medium" = "orange", "Low" = "red")
) +
  labs(
    title = "Per Capita Plastic Waste vs. Recycling Rate",
    x = "Per Capita Plastic Waste (kg)",
    y = "Recycling Rate (%)",
    color = "Recycling Category"
) +
  theme_minimal()
```

Per Capita Plastic Waste vs. Recycling Rate



0.4 Results

```
library(readr)
library(dplyr)
library(janitor)
library(knitr)

# Read the CSV file using a relative path
plastic_data <- read_csv("data/Plastic Waste Around the World.csv")

# Clean column names (convert to lower case with underscores)
plastic_data <- clean_names(plastic_data)

# Group by coastal risk level and calculate summary statistics</pre>
```

Table 1: Plastic Waste Indicators by Coastal Risk Level

coastal_waste_risk	Per_Capita	Total_Capita	SD(Total)	Mean Recycling(%)
Very_High	33.67	2.51	2.45	9.85
High	89.67	1.69	7.44	5.50
Medium	163.46	2.94	7.24	19.78
Low	148.29	0.97	1.30	19.66

As shown in Table 1, plastic waste data from 165 countries is grouped by coastal risk level.

Very High risk countries have the lowest waste per person (33.68 kg), but they still produce the second-largest total waste (2.52 million tonnes), likely due to large populations. Medium risk countries have the highest plastic waste per person (163.46 kg) and also the highest total waste (2.94 million tonnes). This means people in these countries use a lot of plastic, and the overall burden is also high. High risk countries show medium levels of waste per person (89.67 kg), but they have the lowest average recycling rate (5.50%) and the largest difference in total waste between countries (standard deviation of 7.44). Low risk countries produce the smallest amount of total waste (0.97 million tonnes), have more similar waste levels across countries (standard deviation of 1.30), and the highest recycling rate (19.66%).

This shows that coastal waste risk is not simply based on how much waste each person produces. Other factors, like population size and waste system quality, also play an important role.

0.5 Discussion, Conclusion and Recommendations

0.5.0.1 Discussion & Conclusion

This study shows that we can't fully understand coastal plastic pollution just by looking at how much waste people generate or how much gets recycled. We also need to consider consumption habits and how effectively each country manages its waste.

 The Effect of Tourism: Many countries with high plastic waste per person are small islands, such as Iceland and Malta. These places actually recycle quite well. For example, Iceland has the highest plastic waste per person, but also a high recycling rate. This means that the real problem may not be poor waste services, but more waste caused by tourists. In the Maldives, for example, tourists create about five times more waste than local people. Even if the recycling system works well, it is hard to deal with the large increase in waste during busy travel seasons.

- The Role of Population Density: Very High risk countries may have low waste per person, but their total waste is still high. This is because many people live close together, and waste services are often weak. When millions of people each produce a little plastic waste, the total amount becomes very large.
- Differences in Waste Control: Medium and High risk countries show big differences between each other. Some countries pollute much more than others. This makes it harder to plan and manage resources. On the other hand, Low risk countries have more stable systems. They recycle more and have less difference between countries. These countries can be good examples for others to follow.

0.5.0.2 Recommendations

- Improve Waste Management in Tourist Areas: In countries that receive many tourists, it
 is important to have special plans for busy seasons. Governments can charge a small
 environmental fee to visitors to help pay for waste services. Shops should be encouraged
 to use less single-use plastic. Hotels and tourist spots should be required to sort and
 recycle their waste properly.
- Focus on Areas with High Population: In places where many people live close together, it is more helpful to improve waste collection and treatment instead of only focusing on recycling. For example, cities can place smart bins on streets or use mobile waste trucks to reach crowded areas more easily and keep the environment clean.
- Help Countries That Struggle and Share Good Ideas: Some countries in the Medium and High Risk groups have more trouble managing plastic waste. These countries should receive more support through better rules and stronger recycling programs. At the same time, countries that already have good systems can share their experience so that others can learn and use the same methods in their own way.