

## Week-2

## Workshop-2

## AI

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  drive
    sample_data
      README.md
      anscombe.json
      california_housing_test.csv
      california_housing_train.csv
      mnist_test.csv
      mnist_train_small.csv
      product_reviews.csv

#Ananya Dahal
#2408840

import pandas as pd
import numpy as np

# Mount Google Drive
from google.colab import drive
drive.mount('/content/drive')

# Load the CSV file
df_reviews = pd.read_csv('/content/product_reviews.csv')

# Copy original dataframe to preserve it
df_modified = df_reviews.copy()

# Add required columns based on reasonable probabilities
np.random.seed(42) # For reproducibility
n = len(df_modified)

# Add simulated columns
df_modified['Defective'] = np.random.choice([0, 1], size=n, p=[0.95, 0.05])
df_modified['HighReturn'] = np.random.choice([0, 1], size=n, p=[0.9, 0.1])
df_modified['HasComplaint'] = np.random.choice([False, True], size=n, p=[0.8, 0.2])
df_modified['VerifiedPurchase'] = np.random.choice([True, False], size=n, p=[0.6, 0.4])
df_modified['ReviewRating'] = df_modified['rating'] # Copy from original

# Preview modified dataframe
df_modified.head()
```

	review_id	product_category	rating	review_length	helpful_votes	Defective	HighReturn	HasComplaint	VerifiedPurchase	ReviewRating
0	R0000	Home	5	127	7	0	1	False	False	5
1	R0001	Books	4	118	7	1	0	False	True	4
2	R0002	Home	5	113	5	0	1	False	True	5
3	R0003	Books	4	126	3	0	0	False	True	4
4	R0004	Home	5	119	4	0	0	True	False	5

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[13] #Exploratory Analysis

# 1. Prior probability that a product is defective
P_defective = df_modified[df_modified['Defective'] == 1]['ReviewRating'].mean()
print(f"Prior Probability P(Defective): {P_defective:.4f}")

Prior Probability P(Defective): 0.0609

[14] from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

# 2. Compare average review rating for defective vs. non-defective products
avg_rating_defective = df_modified[df_modified['Defective'] == 1]['ReviewRating'].mean()
avg_rating_non_defective = df_modified[df_modified['Defective'] == 0]['ReviewRating'].mean()
print(f"Average Rating (Defective): {avg_rating_defective:.2f}")
print(f"Average Rating (Non-Defective): {avg_rating_non_defective:.2f}")

Average Rating (Defective): 4.33
Average Rating (Non-Defective): 3.72

# 3. Return rate for defective and non-defective products
P_high_return_given_defective = df_modified[df_modified['Defective'] == 1]['HighReturn'].mean()
P_high_return_given_non_defective = df_modified[df_modified['Defective'] == 0]['HighReturn'].mean()
print(f"Return Rate (Defective): {P_high_return_given_defective:.4f}")
print(f"Return Rate (Non-Defective): {P_high_return_given_non_defective:.4f}")

Return Rate (Defective): 0.0000
Return Rate (Non-Defective): 0.0051

[17] #2. Bayesian Inference

# P(HighReturn)
P_high_return = df_modified['HighReturn'].mean()

[18] # Posterior probability: P(Defective | HighReturn)
posterior = (P_high_return_given_defective * P_defective) / P_high_return
print(f"Posterior P(Defective | HighReturn): {posterior:.4f}")

Posterior P(Defective | HighReturn): 0.0000
```

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[19] #Multi-Feature Risk Scoring
# Create 'LowRating' feature
df_modified['LowRating'] = df_modified['ReviewRating'] <= 2

# Calculate conditional probabilities
P_low_rating_given_defective = df_modified[df_modified['Defective'] == 1]['LowRating'].mean()
P_complaint_given_defective = df_modified[df_modified['Defective'] == 1]['HasComplaint'].mean()

[20] # Risk Score Calculation
def risk_score(row):
    score = P_defective
    score += P_high_return_given_defective if row['HighReturn'] == 1 else (1 - P_high_return_given_defective)
    score += P_low_rating_given_defective if row['LowRating'] else (1 - P_low_rating_given_defective)
    score += P_complaint_given_defective if row['HasComplaint'] else (1 - P_complaint_given_defective)
    return score

df_modified['RiskScore'] = df_modified.apply(risk_score, axis=1)

[21] # Identify Top 10 High-Risk Products
top_10_risk = df_modified.sort_values(by='RiskScore', ascending=False).head(10)
print("\nTop 10 High-Risk Products:\n")
print(top_10_risk[['Defective', 'HighReturn', 'ReviewRating', 'HasComplaint', 'RiskScore']])

Top 10 High-Risk Products:
   Defective  HighReturn  ReviewRating  HasComplaint  RiskScore
1          1           0             4           False      0.04
3          0           0             4           False      0.04
14         0           0             5           False      0.04
17         0           0             5           False      0.04
11         1           0             4           False      0.04
20         0           0             4           False      0.04
45         0           0             5           False      0.04
44         0           0             4           False      0.04
43         0           0             5           False      0.04
33         0           0             5           False      0.04
```

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[22] sample_product = {
    'HighReturn': 1,
    'ReviewRating': 1.5,
    'HasComplaint': True,
    'VerifiedPurchase': False,
    'LowRating': 1.5 <= 2 # This is True, since 1.5 <= 2
}

[23] # Calculate risk score for sample product
sample_risk_score = P_defective
sample_risk_score += P_high_return_given_defective if sample_product['HighReturn'] == 1 else (1 - P_high_return_given_defective)
sample_risk_score += P_low_rating_given_defective if sample_product['LowRating'] else (1 - P_low_rating_given_defective)
sample_risk_score += P_complaint_given_defective if sample_product['HasComplaint'] else (1 - P_complaint_given_defective)

print(f"\nRisk Score for sample product: {sample_risk_score:.6f}")

Risk Score for sample product: 0.000000

[24] # Recall recommendation (Threshold Example: 5%)
if sample_risk_score > 0.05:
    print("Recommendation: Recall Suggested.")
else:
    print("Recommendation: No Recall Needed Yet.")

Recommendation: No Recall Needed Yet.

[25] # Suggest Additional Data
print("\nAdditional Data That Would Improve Analysis:")
print("- Supplier Information")
print("- Product Category")
print("- Customer Demographics")
print("- Time-based Return Patterns")

Additional Data That Would Improve Analysis:
- Supplier Information
- Product Category
- Customer Demographics
- Time-based Return Patterns

[25] Start coding or generate with AI.
```