

Microplastic1

October 27, 2024

```
[1]: #!/usr/bin/env python
      # coding: utf-8

      # In[14]:

      import pandas as pd
      # Define the file path provided by the user
      file_path = r'~/Desktop/files/Microplastic.xlsx'

      # Load the Excel file again to check the sheets and prepare for merging
      xls = pd.ExcelFile(file_path)

      # List all sheet names to understand the structure for further operations
      sheet_names = xls.sheet_names
      sheet_names
```

```
[1]: ['Table of Contents',
      'B.1. Chemical list key',
      'B.1. Chemicals list',
      'B.2. Health outcomes list',
      'B.3. Countries list']
```

```
[2]: # In[15]:

      # Read the content of the Excel file to check the sheets available
      xls = pd.ExcelFile(file_path)

      # Display the sheet names
      xls.sheet_names
```

```
[2]: ['Table of Contents',
      'B.1. Chemical list key',
      'B.1. Chemicals list',
      'B.2. Health outcomes list',
      'B.3. Countries list']
```

```
[3]: # In[16]:

# Read the relevant sheets
chemical_list_df = pd.read_excel(file_path, sheet_name='B.1. Chemicals list')
health_outcomes_df = pd.read_excel(file_path, sheet_name='B.2. Health outcomes_
↳list')

# Clean the 'Chemical List' dataframe
chemical_list_cleaned = chemical_list_df[['Chemical class', 'Chemical name', '
↳CAS number', 'Function for SEM inclusion', 'Source', 'Found/Not_
↳found', 'General function*', 'Sector of use*', 'Hazard rating*']].dropna()

# Clean the 'Health Outcomes List' dataframe (after renaming columns_
↳appropriately)
health_outcomes_cleaned = health_outcomes_df.rename(columns={
    'Excel Table B.2. List of health outcome measures showing their ICD_
↳classifications, corresponding search terms, and the health outcome measure_
↳extracted and used for grouping in this systematic evidence map.': 'Health_
↳outcome',
    'Unnamed: 6': 'Group'
})[['Health outcome', 'Group']].dropna()

# Merging the chemical list and health outcomes using 'concat'
merged_data = pd.concat([chemical_list_cleaned, health_outcomes_cleaned],_
↳axis=1)

# Display the first few rows of the merged data
print(merged_data.head())
```

	Chemical class	Chemical name \
0	Bisphenols	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromo-2...
1	Bisphenols	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromopr...
2	Bisphenols	2,6-dibromo-4-[2-(3,5-dibromo-4-hydroxyphenyl)...
26	OPEs	(3-diphenoxyphosphoryloxyphenyl) diphenyl phos...
30	OPEs	2-[diethoxyphosphorylmethyl(2-hydroxyethyl)ami...

	CAS number	Function for SEM inclusion \
0	"97416-84-7"	Flame retardant
1	"21850-44-2"	Flame retardant
2	"79-94-7"	Flame retardant
26	"57583-54-7"	Flame retardant
30	"2781-11-5"	Flame retardant

	Source	Found/Not found \
0	ECHA - EU Chemicals Agency	not found
1	ECHA - EU Chemicals Agency	not found
2	ECHA - EU Chemicals Agency	found

26	Added by expert based on structural similarity...	found
30	Added by expert based on chemical similarity w...	not found

	General function*	\
0	Colorant, Flame Retardant, Intermediates	
1	Biocide, Colorant, Flame Retardant, Lubricant,...	
2	Flame Retardant, Intermediates, Other Processi...	
26	Colorant, Flame Retardant, Lubricant, Plasticizer	
30	Flame Retardant	

	Sector of use*	Hazard rating*	\
0	Building & Construction, Electrical and Electr...	No data available	
1	Building & Construction, Electrical and Electr...	No data available	
2	Automotive, Building & Construction, Electrica...	high	
26	Automotive, Electrical and Electronic Equipmen...	No data available	
30	Textiles	No data available	

	Health outcome	\
0	ICD Level 0	
1	01 Certain infectious or parasitic diseases	
2	01 Certain infectious or parasitic diseases	
26	01 Certain infectious or parasitic diseases	
30	01 Certain infectious or parasitic diseases	

	Group
0	Group
1	Other infectious or parasitic diseases
2	Other infectious or parasitic diseases
26	Viral infectious disease
30	Viral infectious disease

[4]: # In[17]:

```
# Check for any missing, NaN, or empty values in the merged dataset
missing_values = merged_data.isnull().sum()

# Display the count of missing values in each column
print(missing_values)
```

Chemical class	955
Chemical name	955
CAS number	955
Function for SEM inclusion	955
Source	955
Found/Not found	955
General function*	955
Sector of use*	955
Hazard rating*	955

```
Health outcome          36
Group                   36
dtype: int64
```

```
[5]: # In[18]:

# Fill missing values with the mode (most frequent value) for each column
merged_data_filled = merged_data.apply(lambda x: x.fillna(x.mode()[0]))

# Display the first few rows of the filled dataset to verify
print(merged_data_filled.head())
```

	Chemical class	Chemical name \
0	Bisphenols	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromo-2...
1	Bisphenols	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromopr...
2	Bisphenols	2,6-dibromo-4-[2-(3,5-dibromo-4-hydroxyphenyl)...
26	OPEs	(3-diphenoxyphosphoryloxyphenyl) diphenyl phos...
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	CAS number	Function for SEM inclusion \
0	"97416-84-7"	Flame retardant
1	"21850-44-2"	Flame retardant
2	"79-94-7"	Flame retardant
26	"57583-54-7"	Flame retardant
30	"2781-11-5"	Flame retardant

	Source	Found/Not found \
0	ECHA - EU Chemicals Agency	not found
1	ECHA - EU Chemicals Agency	not found
2	ECHA - EU Chemicals Agency	found
26	Added by expert based on structural similarity...	found
30	Added by expert based on chemical similarity w...	not found

	General function* \
0	Colorant, Flame Retardant, Intermediates
1	Biocide, Colorant, Flame Retardant, Lubricant,...
2	Flame Retardant, Intermediates, Other Processi...
26	Colorant, Flame Retardant, Lubricant, Plasticizer
30	Flame Retardant

	Sector of use*	Hazard rating* \
0	Building & Construction, Electrical and Electr...	No data available
1	Building & Construction, Electrical and Electr...	No data available
2	Automotive, Building & Construction, Electrica...	high
26	Automotive, Electrical and Electronic Equipmen...	No data available
30	Textiles	No data available

Health outcome \

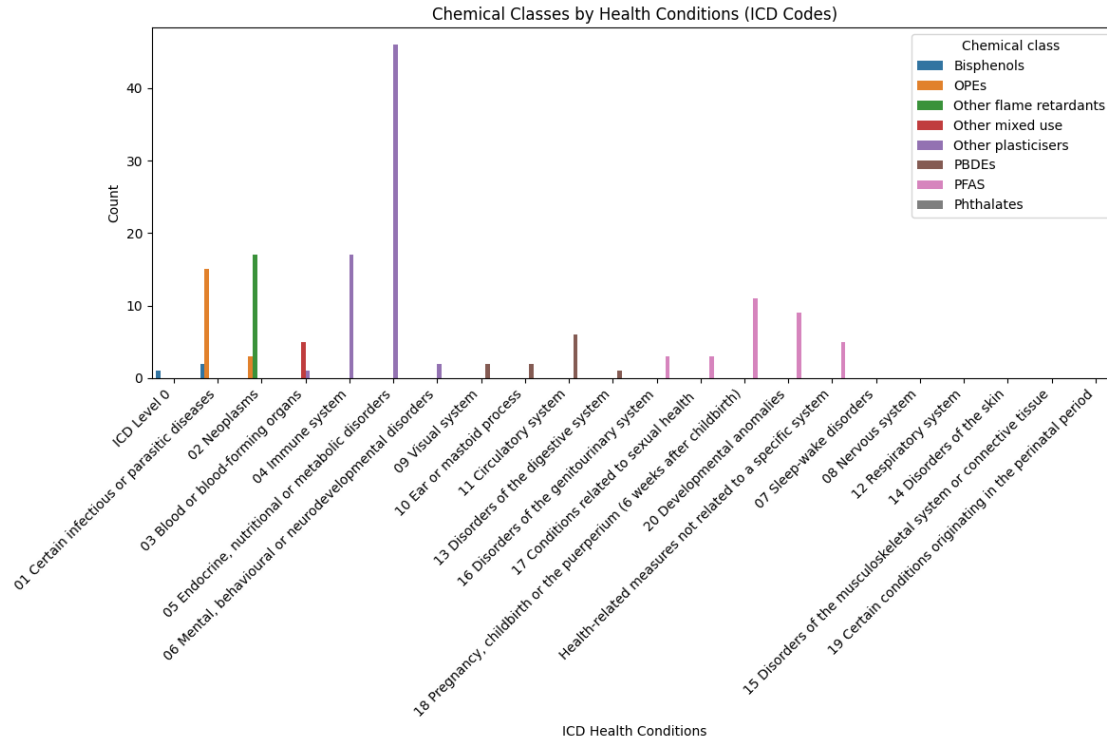
	ICD Level 0
0	
1	01 Certain infectious or parasitic diseases
2	01 Certain infectious or parasitic diseases
26	01 Certain infectious or parasitic diseases
30	01 Certain infectious or parasitic diseases

	Group
0	Group
1	Other infectious or parasitic diseases
2	Other infectious or parasitic diseases
26	Viral infectious disease
30	Viral infectious disease

```
[6]: # In[19]:

import seaborn as sns
import matplotlib.pyplot as plt

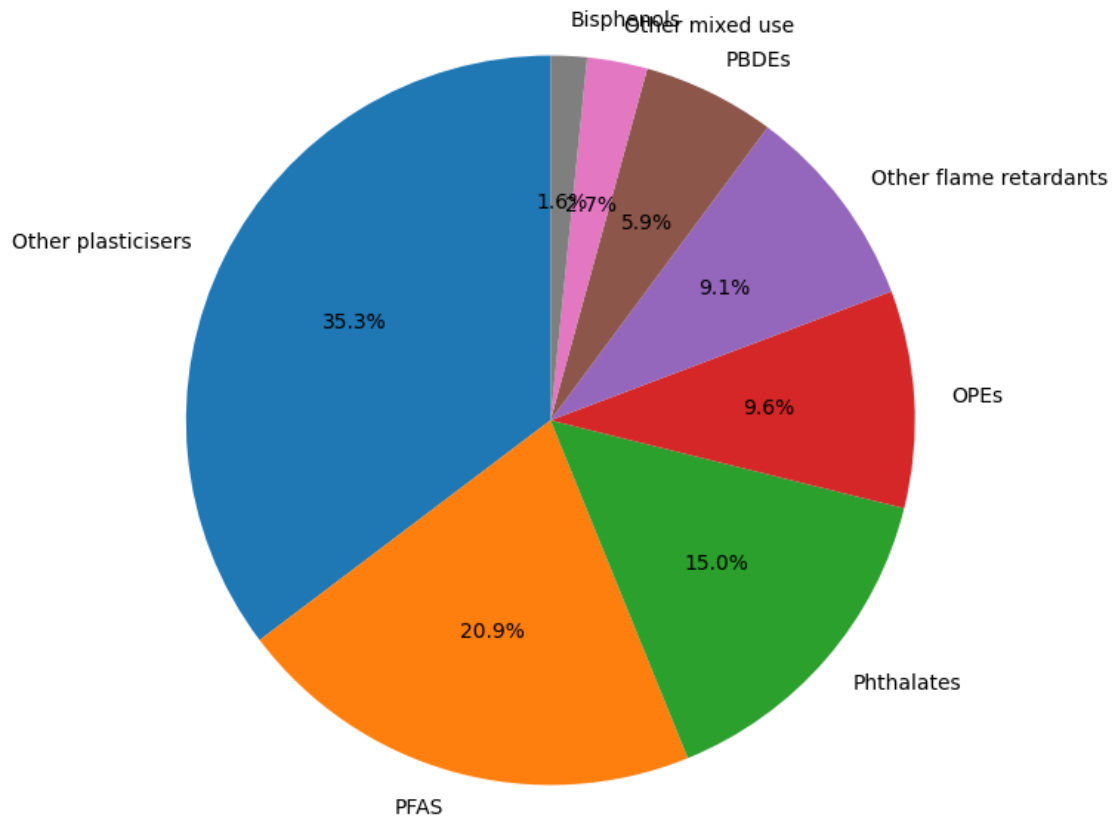
# Create a count plot (bar chart) to visualize the number of occurrences of
↳ each chemical class by ICD code
plt.figure(figsize=(12, 8))
sns.countplot(data=merged_data, x='Health outcome', hue='Chemical class')
plt.title('Chemical Classes by Health Conditions (ICD Codes)')
plt.xticks(rotation=45, ha='right')
plt.xlabel('ICD Health Conditions')
plt.ylabel('Count')
plt.tight_layout()
plt.show()
```



```
[7]: # In[20]:

# Plot pie chart for chemical class distribution
plt.figure(figsize=(8, 8))
merged_data['Chemical class'].value_counts().plot(kind='pie', autopct='%1.
    ↪1f%%', startangle=90)
plt.title("Distribution of Chemical Classes Related to Health Conditions")
plt.ylabel('')
plt.show()
```

Distribution of Chemical Classes Related to Health Conditions



```
[8]: # In[21]:

##
def categorize_risk(hazard_rating):
    if hazard_rating == 'high':
        return 'high risk'
    elif hazard_rating == 'low':
        return 'low risk'
    elif hazard_rating == 'medium':
        return 'medium risk'
    elif hazard_rating == 'No data available':
        return 'unknown risk'
    else:
        return 'unknown risk' # Optional: Handle unexpected cases

# Apply the function and create a new column
```

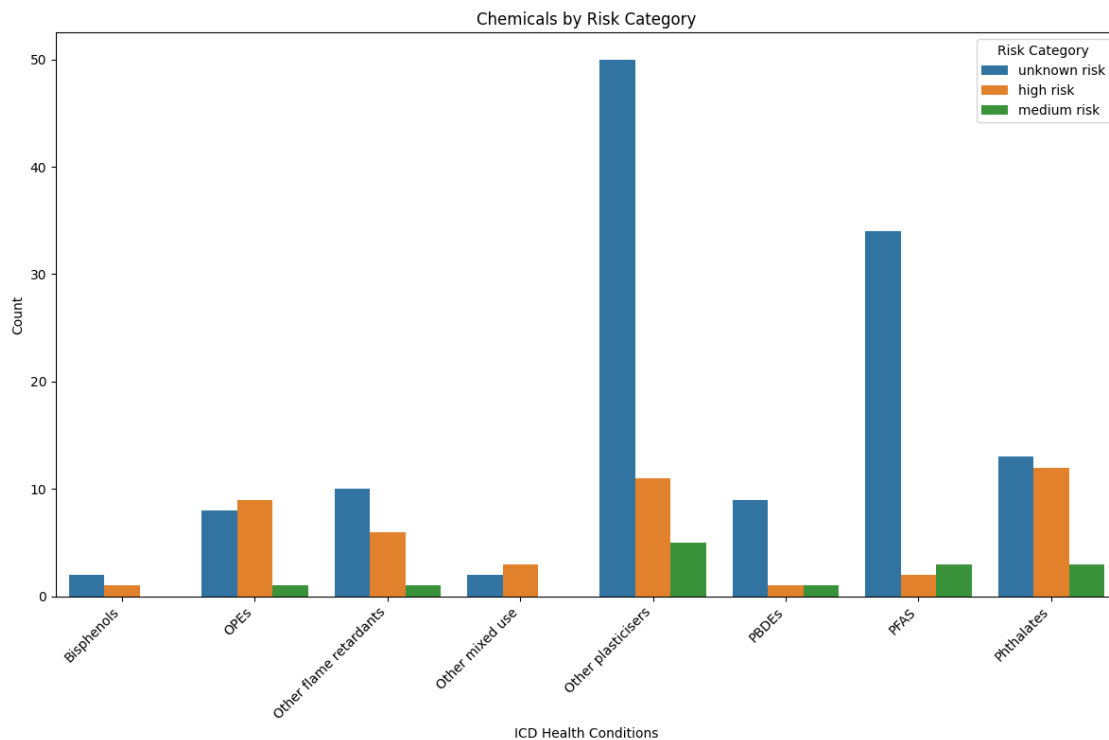
```
merged_data['Risk Category'] = merged_data['Hazard rating*'].
    ↪ apply(categorize_risk)
##

# Display the first few rows to verify
print(merged_data[['Chemical name', 'Risk Category']].head())
```

	Chemical name	Risk Category
0	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromo-2...	unknown risk
1	1,3-dibromo-5-[2-[3,5-dibromo-4-(2,3-dibromopr...	unknown risk
2	2,6-dibromo-4-[2-(3,5-dibromo-4-hydroxyphenyl)...	high risk
26	(3-diphenoxyphosphoryloxyphenyl) diphenyl phos...	unknown risk
30	2-[diethoxyphosphorylmethyl(2-hydroxyethyl)ami...	unknown risk

[9]: # In[22]:

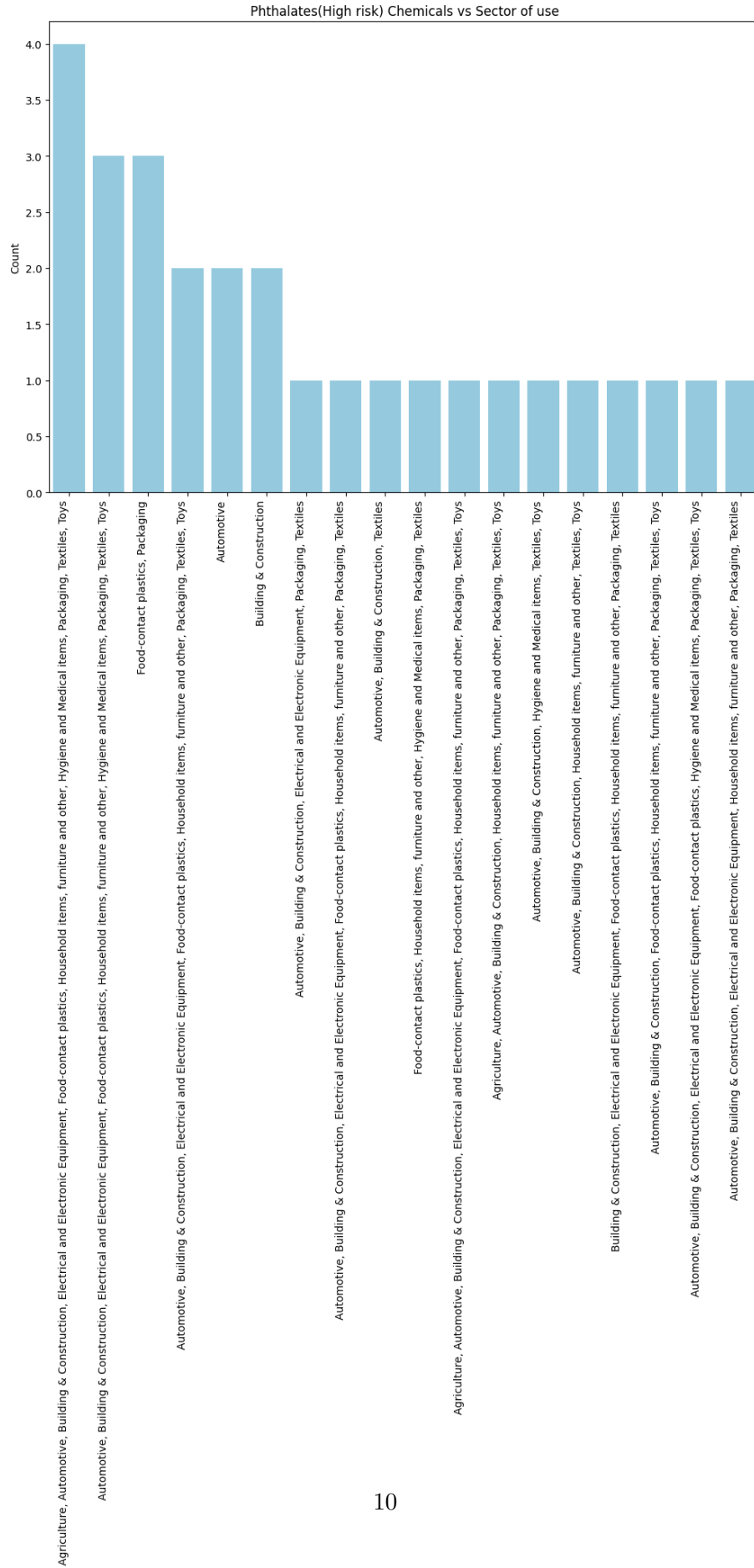
```
# Create a bar chart of chemicals by risk category and health outcomes
plt.figure(figsize=(12, 8))
sns.countplot(data=merged_data, x='Chemical class', hue='Risk Category')
plt.title("Chemicals by Risk Category")
plt.xticks(rotation=45, ha='right')
plt.xlabel('ICD Health Conditions')
plt.ylabel('Count')
plt.tight_layout()
plt.show()
```




```
[18]: # chemical class with the highest risk effect

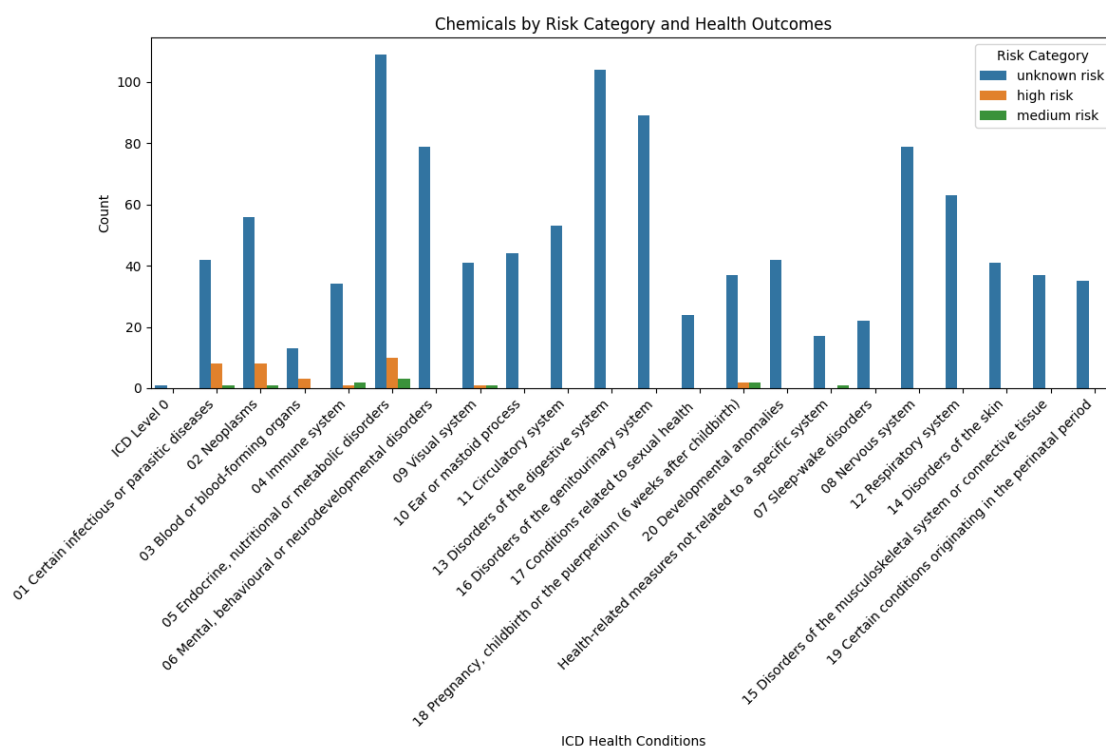
Phthalates = merged_data[merged_data['Chemical class'] == 'Phthalates']

# Create a bar chart of phthalates chemicals by sector of use
# Calculate the counts of each sector
sector_counts = Phthalates['Sector of use*'].value_counts().index.tolist()
plt.figure(figsize=(12, 8))
sns.countplot(data=Phthalates, x='Sector of use*', color='skyblue',
              order=sector_counts)
plt.title("Phthalates(High risk) Chemicals vs Sector of use")
plt.xticks(rotation=90, ha='right')
plt.xlabel('Sector of Use')
plt.ylabel('Count')
plt.show()
```



```
[ ]: # In[22]:
```

```
# Create a bar chart of chemicals by risk category and health outcomes
plt.figure(figsize=(12, 8))
sns.countplot(data=merged_data, x='Health outcome', hue='Risk Category')
plt.title("Chemicals by Risk Category and Health Outcomes")
plt.xticks(rotation=45, ha='right')
plt.xlabel('ICD Health Conditions')
plt.ylabel('Count')
plt.tight_layout()
plt.show()
```

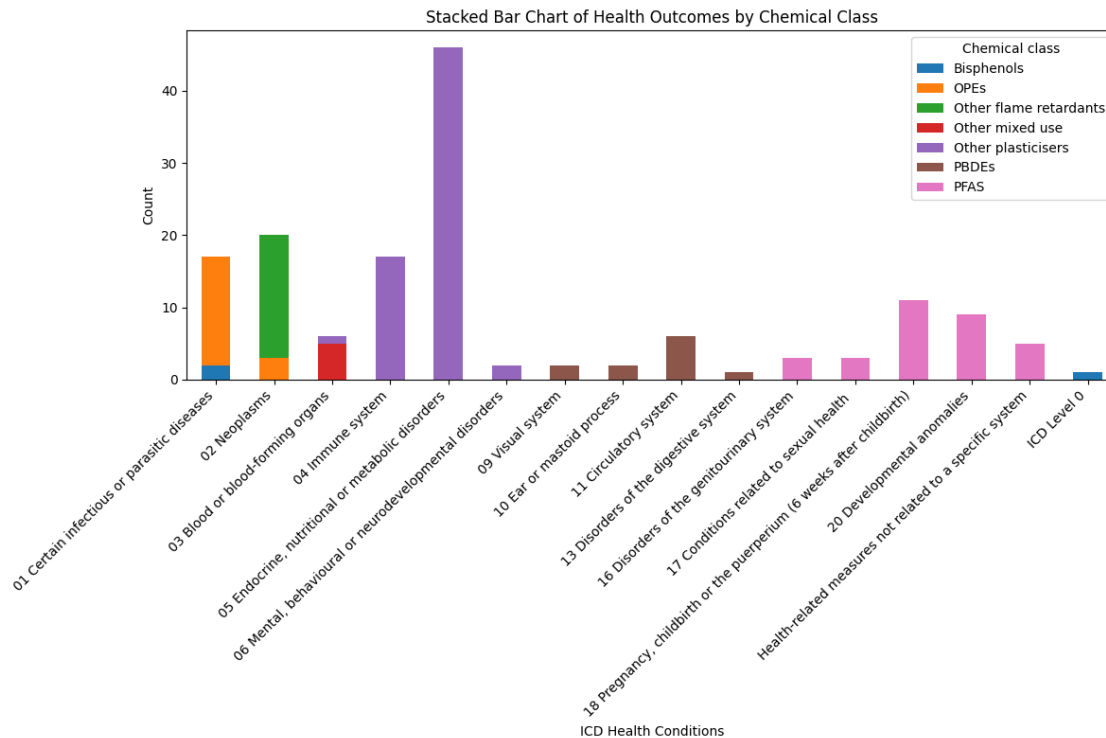


```
[ ]: # In[23]:
```

```
# Group the data by 'Health outcome' and 'Chemical class'
grouped_data = merged_data.groupby(['Health outcome', 'Chemical class']).size().
    unstack(fill_value=0)

# Plot a stacked bar chart
grouped_data.plot(kind='bar', stacked=True, figsize=(12, 8))
```

```
plt.title("Stacked Bar Chart of Health Outcomes by Chemical Class")
plt.xlabel('ICD Health Conditions')
plt.ylabel('Count')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```



```
[ ]: # In[45]:
```

```
print(merged_data.columns)
```

```
Index(['Chemical class', 'Chemical name', 'CAS number',
      'Function for SEM inclusion', 'Source', 'Found/Not found',
      'General function*', 'Sector of use*', 'Hazard rating*',
      'Health outcome', 'Group', 'Risk Category'],
      dtype='object')
```

```
[ ]: # In[46]:
```

```
# Create a binary target variable based on 'Risk Category'
# Assuming 'High Risk' indicates responsibility for health issues
merged_data['Responsible'] = merged_data['Risk Category'].apply(lambda x: 1 if
↳ x == 'high risk' else 0)
```

```

# Encode the features

from sklearn.preprocessing import LabelEncoder

# Instances for LabelEncoder
le_chemical = LabelEncoder()
le_health = LabelEncoder()

# Fit and transform the features
merged_data['Chemical class_encoded'] = le_chemical.
    ↳fit_transform(merged_data['Chemical class'])
merged_data['Health outcome_encoded'] = le_health.
    ↳fit_transform(merged_data['Health outcome'])

# Prepare features (X) and target (y)
X = merged_data[['Chemical class_encoded', 'Health outcome_encoded']] #
    ↳Features
y = merged_data['Responsible'] # Target variable: whether the chemical is
    ↳responsible

# Split the data into training and test sets
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↳random_state=42)

# Train a Random Forest model (or any other model you prefer)
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report

rf_classifier = RandomForestClassifier(random_state=42)
rf_classifier.fit(X_train, y_train)

# Make predictions
y_pred = rf_classifier.predict(X_test)

# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)
print(f"Random Forest Accuracy: {accuracy:.2f}")

# Print detailed classification report
print(classification_report(y_test, y_pred))

```

Random Forest Accuracy: 0.97

	precision	recall	f1-score	support
0	0.98	1.00	0.99	223

	1	0.50	0.17	0.25	6
accuracy				0.97	229
macro avg		0.74	0.58	0.62	229
weighted avg		0.97	0.97	0.97	229

```
[ ]: # In[51]:

print(merged_data['Risk Category'].value_counts())
# or check other columns that could serve as a target
```

```
Risk Category
unknown risk    1083
high risk        45
medium risk      14
Name: count, dtype: int64
```

```
[ ]: # In[53]:

from sklearn.neighbors import KNeighborsClassifier

# Initialize KNN model
knn_model = KNeighborsClassifier(n_neighbors=5)

# Train model
knn_model.fit(X_train, y_train)

# Predict
y_pred = knn_model.predict(X_test)

# Accuracy and classification report
accuracy = accuracy_score(y_test, y_pred)
print(f"KNN Accuracy: {accuracy:.2f}")
print(classification_report(y_test, y_pred))
```

```
KNN Accuracy: 0.97
```

		precision	recall	f1-score	support
	0	0.98	0.98	0.98	223
	1	0.33	0.33	0.33	6
accuracy				0.97	229
macro avg		0.66	0.66	0.66	229
weighted avg		0.97	0.97	0.97	229

```
[ ]: # In[54]:

from sklearn.tree import DecisionTreeClassifier

# Initialize Decision Tree model
dt_model = DecisionTreeClassifier(random_state=42)

# Train model
dt_model.fit(X_train, y_train)

# Predict
y_pred = dt_model.predict(X_test)

# Accuracy and classification report
accuracy = accuracy_score(y_test, y_pred)
print(f"Decision Tree Accuracy: {accuracy:.2f}")
print(classification_report(y_test, y_pred))
```

Decision Tree Accuracy: 0.97

	precision	recall	f1-score	support
0	0.97	1.00	0.98	223
1	0.00	0.00	0.00	6
accuracy			0.97	229
macro avg	0.49	0.50	0.49	229
weighted avg	0.95	0.97	0.96	229

	precision	recall	f1-score	support
0	0.97	1.00	0.98	223
1	0.00	0.00	0.00	6
accuracy			0.97	229
macro avg	0.49	0.50	0.49	229
weighted avg	0.95	0.97	0.96	229