# STATISTICAL CONSULTING HW2

## Data Analysis of mushroom

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## Easy look

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from scipy.stats import chi2_contingency

# File path
file_path = '/home/lee/workspace/STATISTICAL-CONSULTING_113-2/HW2/mushroom/primary_data.csv'

# Read the CSV file with semicolon as delimiter
df = pd.read_csv(file_path, delimiter=';')

# Display basic information about the dataset
print("Dataset Shape:", df.shape)
```

#### Dataset Shape: (173, 23)

```
print("Column Names:")
for col in df.columns:
    print(f"- {col}")
```

#### Column Names:

- family
- name
- class
- cap-diameter
- cap-shape
- Cap-surface
- cap-color
- does-bruise-or-bleed
- gill-attachment
- gill-spacing
- gill-color

```
- stem-height
```

- stem-width
- stem-root
- stem-surface
- stem-color
- veil-type
- veil-color
- has-ring
- ring-type
- Spore-print-color
- habitat
- season

```
print("Data Types:")
print(df.dtypes)
```

#### Data Types:

family object name object class object cap-diameter object cap-shape object Cap-surface object cap-color object does-bruise-or-bleed object gill-attachment object gill-spacing object gill-color object stem-height object stem-width object stem-root object stem-surface object stem-color object veil-type object veil-color object has-ring object ring-type object Spore-print-color object habitat object season object

dtype: object

print("Missing Values Count:") print(df.isna().sum())

### Missing Values Count:

family 0 0 name0 class 0 cap-diameter 0 cap-shape 40 Cap-surface cap-color 0 does-bruise-or-bleed 0 gill-attachment 28 71 gill-spacing

```
0
gill-color
stem-height
                          0
stem-width
                          0
stem-root
                        146
stem-surface
                        108
stem-color
                          0
veil-type
                        164
                        152
veil-color
                          0
has-ring
                          7
ring-type
Spore-print-color
                        155
habitat
                          0
                          0
season
dtype: int64
print("\nCount of Mushrooms by Family:")
family_counts = df['family'].value_counts()
print(family_counts)
Count of Mushrooms by Family:
family
Tricholoma Family
Russula Family
                          27
Bolete Family
                          14
Ink Cap Family
                          13
Cortinarius Family
                          11
Stropharia Family
                           8
                           8
Amanita Family
Wax Gill Family
                           8
Entoloma Family
                           7
                           7
Bracket Fungi
Mushroom Family
                           5
Bolbitius Family
                           3
Lepiota Family
                           3
Paxillus Family
                           3
Chanterelle Family
Pluteus Family
                           2
Oyster Mushroom Family
Crepidotus Family
Hydnum Family
Ear-Pick Family
                           1
Saddle-Cup Family
                           1
Morel Family
Jelly Discs Family
                           1
Name: count, dtype: int64
# Count of edible vs poisonous mushrooms
print("\nCount of Edible vs Poisonous Mushrooms:")
class_counts = df['class'].value_counts()
print(class_counts)
print("\nPercentage of Edible vs Poisonous Mushrooms:")
```

Count of Edible vs Poisonous Mushrooms:

print(class\_counts / len(df) \* 100)

```
class
     96
     77
Name: count, dtype: int64
Percentage of Edible vs Poisonous Mushrooms:
class
     55.491329
     44.508671
Name: count, dtype: float64
# Function to extract average from range values
def extract_average(range_str):
    if pd.isna(range_str) or range_str == '':
        return np.nan
    try:
        # Remove brackets and split by comma
        range_str = str(range_str).strip('[]')
        values = [float(x.strip()) for x in range_str.split(',')]
        return sum(values) / len(values)
    except:
        return np.nan
# Process range columns
range_columns = ['cap-diameter', 'stem-height', 'stem-width']
for col in range_columns:
    if col in df.columns:
        df[f'{col}_avg'] = df[col].apply(extract_average)
# Basic statistics for processed numerical columns
print("\nBasic Statistics for Numerical Features:")
numerical_cols = [col for col in df.columns if col.endswith('_avg')]
if numerical cols:
    print(df[numerical_cols].describe())
```

#### Basic Statistics for Numerical Features:

	cap-diameter_avg	stem-height_avg	stem-width_avg
count	173.000000	173.000000	173.000000
mean	6.739306	6.589595	12.157514
std	5.140185	3.264824	9.858009
min	0.700000	0.000000	0.000000
25%	3.500000	4.500000	5.500000
50%	6.000000	6.000000	10.000000
75%	8.500000	7.500000	15.000000
max	50.000000	25.000000	70.000000

#### Table 1

```
# Fixed function to analyze feature association with edibility
def analyze_feature_edibility(df, feature):
    """
    Analyze the association between a feature and mushroom edibility
    without using crosstab which has issues with duplicate labels.
    """
```

```
if feature not in df.columns:
        return None
    # Create a copy to avoid modifying the original data
    result = {"Feature values": [], "Edible count": [], "Poisonous count": []}
    # Process the feature values
    for value in df[feature].dropna().unique():
        # For features with multiple values like "[e, o]"
        value_str = str(value).strip('[]')
        values = [v.strip() for v in value_str.split(',')]
        for v in values:
            if v not in result["Feature values"]:
                # Filter rows with this feature value
                matches = df[df[feature].str.contains(v, na=False)]
                # Count edible and poisonous mushrooms
                edible_count = len(matches[matches['class'] == 'e'])
                poisonous_count = len(matches[matches['class'] == 'p'])
                # Add to results
                result["Feature values"].append(v)
                result["Edible count"].append(edible_count)
                result["Poisonous count"].append(poisonous_count)
    # Convert to DataFrame for better display
    result_df = pd.DataFrame(result)
    # Calculate total and percentages
   result_df["Total"] = result_df["Edible count"] + result_df["Poisonous count"]
    result_df["% Edible"] = (result_df["Edible count"] / result_df["Total"] * 100).round(1)
    result_df["% Poisonous"] = (result_df["Poisonous count"] / result_df["Total"] * 100).round(1)
    return result_df
# Analyze important features
print("Table1:")
important_features = ['cap-color', 'does-bruise-or-bleed', 'gill-attachment', 'has-ring', 'cap-shape',
for feature in important_features:
    if feature in df.columns:
        print(f"\nAssociation between {feature} and edibility:")
        result = analyze_feature_edibility(df, feature)
        if result is not None:
            print(result)
print("\nAnalysis completed!")
```

#### Table1:

```
Association between cap-color and edibility:
Feature values Edible count Poisonous count Total % Edible % Poisonous
```

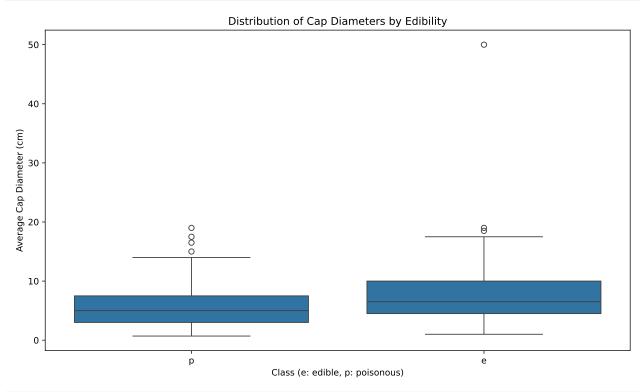
0	е	7		18	25		28.0		72.0
1		_		15	22		31.8		68.2
	0								
2	n			57	110		48.2		51.8
3	g	14		14	28		50.0		50.0
4	r	2		11	13		15.4		84.6
5	W	17		18	35		48.6		51.4
6				28			36.4		63.6
	У								
7	p	4		7	11		36.4		63.6
8	b	5		2	7		71.4		28.6
9	u	. 5		5	10		50.0		50.0
10	1	4		2			66.7		33.3
11	k	. 3		6	9		33.3		66.7
Asso	ociation betwe	en does-bruise			-				
F€	eature values	Edible count	Poisonous	count	Total	% :	Edible	%	Poisonous
0	f	63		80	143		44.1		55.9
1	t	14		16	30		46.7		53.3
1	C	14		10	30		40.7		55.5
		en gill-attach							
F€	eature values	Edible count	Poisonous	count	Total	% :	Edible	%	Poisonous
0	е	10		6	16		62.5		37.5
1	a	16		24	40		40.0		60.0
2	d	14		19	33		42.4		57.6
3	S	7		9	16		43.8		56.2
4	X	9		12	21		42.9		57.1
5	р	12		5	17		70.6		29.4
6	f	4		6	10		40.0		60.0
O	_	-		U	10		40.0		00.0
		en has-ring and							
	ociation betwe eature values				Total	% :	Edible	%	Poisonous
					Total 43	% :	Edible 39.5	%	Poisonous 60.5
F 6	eature values t	Edible count		count 26	43	% 1	39.5	%	60.5
Fe	eature values	Edible count		count		<b>%</b> 1		%	
F 6 0 1	eature values t f	Edible count 17 60	Poisonous	26 70	43	<b>%</b> 1	39.5	%	60.5
Fe 0 1	eature values t f ociation betwe	Edible count 17 60 en cap-shape a	Poisonous nd edibilit	26 70	43 130		39.5 46.2		60.5 53.8
Fe 0 1	eature values t f	Edible count 17 60 en cap-shape a	Poisonous nd edibilit	26 70	43		39.5		60.5
Fe 0 1	eature values t f ociation betwe	Edible count 17 60 en cap-shape a	Poisonous nd edibilit	26 70	43 130		39.5 46.2		60.5 53.8
Fe O 1 Asso	eature values t f ociation betwe eature values x	Edible count 17 60 en cap-shape a Edible count 54	Poisonous nd edibilit	count 26 70  Ey: count 56	43 130 Total 110		39.5 46.2 Edible 49.1		60.5 53.8 Poisonous 50.9
Fe O 1 Asso Fe O 1	eature values  t f  ociation betweeature values  x f	Edible count 17 60 en cap-shape a Edible count 54 36	Poisonous nd edibilit	26 70 2y: count 56 38	43 130 Total 110 74		39.5 46.2 Edible 49.1 48.6		60.5 53.8 Poisonous 50.9 51.4
Fe 0 1 Asso Fe 0 1 2	eature values  t f  ociation betweeature values  x f p	Edible count 17 60 en cap-shape at Edible count 54 36 10	Poisonous nd edibilit	26 70 27: count 56 38 5	43 130 Total 110 74 15		39.5 46.2 Edible 49.1 48.6 66.7		60.5 53.8 Poisonous 50.9 51.4 33.3
Fe 0 1 Asso Fe 0 1 2 3	eature values  t f  ociation betweeature values  x f	Edible count 17 60 en cap-shape at Edible count 54 36 10 5	Poisonous nd edibilit	count 26 70 cy: count 56 38 5 18	43 130 Total 110 74 15 23		39.5 46.2 Edible 49.1 48.6 66.7 21.7		60.5 53.8 Poisonous 50.9 51.4 33.3 78.3
Fe 0 1 Asso Fe 0 1 2	eature values  t f  ociation betweeature values  x f p	Edible count 17 60 en cap-shape at Edible count 54 36 10	Poisonous nd edibilit	26 70 27: count 56 38 5	43 130 Total 110 74 15		39.5 46.2 Edible 49.1 48.6 66.7		60.5 53.8 Poisonous 50.9 51.4 33.3
Fe 0 1 Asso 0 1 2 3 4	eature values t f  occiation betwe eature values x f p b	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4	Poisonous nd edibilit	count 26 70 cy: count 56 38 5 18 4	43 130 Total 110 74 15 23 8		39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0		60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0
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Fe 0 1 Asso 0 1 2 3 4	eature values t f  occiation betwee eature values x f p b c	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4	Poisonous nd edibilit	count 26 70 cy: count 56 38 5 18 4	43 130 Total 110 74 15 23 8		39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0		60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0
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Fe 0 1 Asso Fe 0 1 2 3 4 5 6 Asso	eature values  t f  ociation betwee eature values  x f p b c s o	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4 17 4 en gill-color	Poisonous  nd edibilit Poisonous  and edibili	count 26 70 cy: count 56 38 5 18 4 19 8	43 130 Total 110 74 15 23 8 36 12	% 1	39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0 47.2 33.3	%	60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0 52.8 66.7
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Fe 0 1 Asso Fe 0 1 2 3 4 5 6 Asso	eature values  t f  ociation betwee eature values  x f p b c s o	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4 17 4 en gill-color Edible count	Poisonous  nd edibilit Poisonous  and edibili Poisonous	count 26 70 cy: count 56 38 5 18 4 19 8	43 130 Total 110 74 15 23 8 36 12	% 1	39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0 47.2 33.3	%	60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0 52.8 66.7
F6 0 1 Asso F6 0 1 2 3 4 5 6	eature values  t f  occiation betwee eature values  x f p b c s o occiation betwee	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4 17 4 en gill-color Edible count	Poisonous  nd edibilit Poisonous  and edibili Poisonous	count 26 70 cy: count 56 38 5 18 4 19 8 cty: s count	43 130 Total 110 74 15 23 8 36 12	% 1	39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0 47.2 33.3	%	60.5 53.8  Poisonous 50.9 51.4 33.3 78.3 50.0 52.8 66.7
Fe 0 1 Asso Fe 0 1 2 3 4 4 5 6 Asso H 0 1	eature values  t f  occiation betwee eature values  x f p b c s o occiation betwee	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4 17 4 en gill-color Edible count 38	Poisonous  nd edibilit Poisonous  and edibili Poisonous	count 26 70 cy: count 56 38 5 18 4 19 8 count 35 32	43 130 Total 110 74 15 23 8 36 12 Total 73 47	% 1	39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0 47.2 33.3 Edible 52.1 31.9	%	60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0 52.8 66.7
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Fe 0 1 Asso Fe 0 1 2 3 4 5 6 Asso I 0 1 2 3 4 5 5	eature values  t f  occiation betwee eature values  x f p b c s o occiation betwee Feature values  w n p u b g	Edible count 17 60 en cap-shape at Edible count 54 36 10 5 4 17 4 en gill-color Edible count 38 15 12 3 10 17 2	Poisonous  nd edibilit Poisonous  and edibili Poisonous	count 26 70 29: count 56 38 5 18 4 19 8 25: count 35 32 16 4 2 13	43 130 Total 110 74 15 23 8 36 12 Total 73 47 28 7 5 23	% 1	39.5 46.2 Edible 49.1 48.6 66.7 21.7 50.0 47.2 33.3 Edible 52.1 31.9 42.9 60.0 43.5	%	60.5 53.8 Poisonous 50.9 51.4 33.3 78.3 50.0 52.8 66.7 Poisonous 47.9 68.1 57.1 40.0 56.5

9	0	5	8	13	38.5	61.5
10	k	6	9	15	40.0	60.0
11	f	4	6	10	40.0	60.0

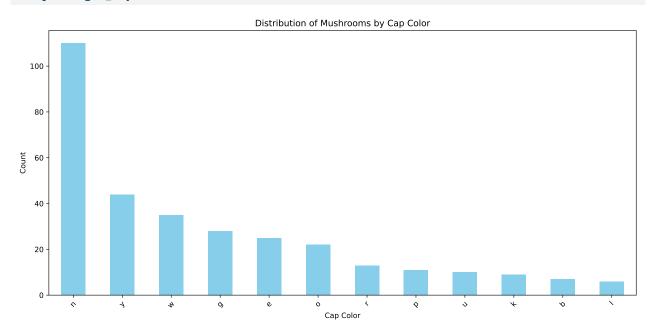
Analysis completed!

### plot

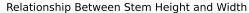
```
plt.figure(figsize=(10, 6))
if 'cap-diameter_avg' in df.columns:
    sns.boxplot(x='class', y='cap-diameter_avg', data=df)
    plt.title('Distribution of Cap Diameters by Edibility')
    plt.xlabel('Class (e: edible, p: poisonous)')
    plt.ylabel('Average Cap Diameter (cm)')
    plt.tight_layout()
```

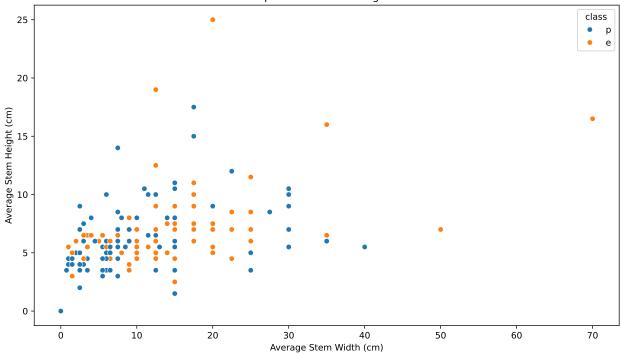


```
plt.xticks(rotation=45)
plt.tight_layout()
```



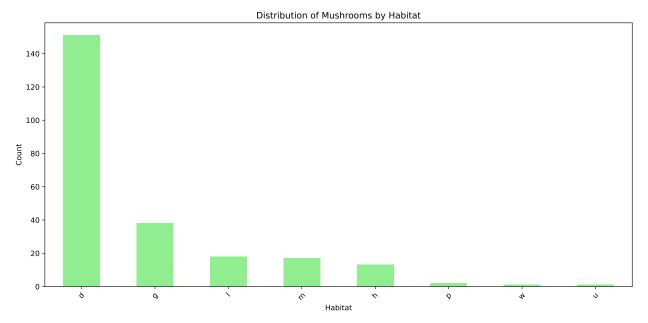
```
plt.figure(figsize=(10, 6))
if 'stem-height_avg' in df.columns and 'stem-width_avg' in df.columns:
    sns.scatterplot(x='stem-width_avg', y='stem-height_avg', hue='class', data=df)
    plt.title('Relationship Between Stem Height and Width')
    plt.xlabel('Average Stem Width (cm)')
    plt.ylabel('Average Stem Height (cm)')
    plt.tight_layout()
```





```
plt.figure(figsize=(12, 6))
if 'habitat' in df.columns:
    # Process multi-value fields properly
habitats = []
for habitat_str in df['habitat'].dropna():
    # Remove brackets and split by comma
    habitat_list = str(habitat_str).strip('[]').split(',')
    habitats.extend([h.strip() for h in habitat_list])

habitat_counts = pd.Series(habitats).value_counts()
habitat_counts.plot(kind='bar', color='lightgreen')
plt.title('Distribution of Mushrooms by Habitat')
plt.xlabel('Habitat')
plt.ylabel('Count')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.tight_layout()
```



```
plt.figure(figsize=(12, 6))
edibility_by_family = pd.crosstab(df['family'], df['class'])
edibility_by_family_pct = edibility_by_family.div(edibility_by_family.sum(axis=1), axis=0) * 100
edibility_by_family_pct.plot(kind='bar', stacked=True)
plt.title('Percentage of Edible vs Poisonous Mushrooms by Family')
plt.xlabel('Mushroom Family')
plt.ylabel('Percentage')
plt.legend(title='Class', labels=['Edible', 'Poisonous'])
plt.xticks(rotation=45)
plt.tight_layout()
```

<Figure size 3600x1800 with 0 Axes>

