```
import gdown
file_id = '1MCmntA3BOquC7BWxLcYDItohDgTPRKx8'
url = f'https://drive.google.com/uc?id={file_id}'
output = 'Crop Data Final.csv'
gdown.download(url, output, quiet=False)
→ Downloading...
   From: https://drive.google.com/uc?id=1MCmntA3BOquC7BWxLcYDItohDgTPRKx8
   To: /content/Crop_Data_Final.csv
   100% | 214k/214k [00:00<00:00, 48.6MB/s]
num_itemsets = frequent_itemsets['itemsets'].apply(len).value_counts().to_dict()
rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.5, num_itemsets
   NameError
                                   Traceback (most recent call last)
   <ipython-input-2-3cf694829650> in <cell line: 0>()
   ---> 1 num_itemsets = frequent_itemsets['itemsets'].apply(len).value_counts().to_dict()
        2 rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.5, num_itemsets=num_itemsets)
   NameError: name 'frequent_itemsets' is not defined
import pandas as pd
from mlxtend.frequent patterns import apriori, association rules
from sklearn.preprocessing import KBinsDiscretizer
# Load dataset
file_path = '/content/Crop_Data_Final.csv'
data = pd.read_csv(file_path)
# Districts to process
districts = data['Dist Name'].unique()
# Binarization thresholds
binarize columns = [
    'RICE YIELD (Kg per ha)', 'WHEAT YIELD (Kg per ha)', 'SORGHUM YIELD (Kg per ha)',
    'PEARL MILLET YIELD (Kg per ha)', 'MAIZE YIELD (Kg per ha)', 'CHICKPEA YIELD (Kg per ha)',
    'PIGEONPEA YIELD (Kg per ha)', 'GROUNDNUT YIELD (Kg per ha)', 'SESAMUM YIELD (Kg per ha)',
    'OILSEEDS YIELD (Kg per ha)', 'SUGARCANE YIELD (Kg per ha)', 'COTTON YIELD (Kg per ha)',
    'Precipitation (mm)', 'Annual Rainfall'
]
# Function to discretize continuous variables
def discretize data(df, columns):
    discretizer = KBinsDiscretizer(n_bins=2, encode='ordinal', strategy='uniform')
    df = df.copy() # Prevent SettingWithCopyWarning
    for col in columns:
        if col in df.columns:
            df[col + '_Binned'] = discretizer.fit_transform(df[[col]]).astype(int)
    return df
# Function to generate transactions for Apriori
def generate_transactions(df):
    transactions = pd.DataFrame()
    for col in df.columns:
        if col.endswith('_Binned'):
             transactions[col + '_High'] = (df[col] == 1).astype(int)
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transactions[col + '_Low'] = (df[col] == 0).astype(int)
   return transactions
# Function to process each district
def process_district(district):
   # Filter district-specific data and discretize
   district_data = data[data['Dist Name'] == district].copy()
   district data = discretize data(district data, binarize columns)
   # Generate transactions
   transactions = generate_transactions(district_data)
   # Run Apriori algorithm
   frequent_itemsets = apriori(transactions, min_support=0.2, use_colnames=True)
   num_itemsets = frequent_itemsets['itemsets'].apply(len).value_counts().to_dict()
   rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.5, num_item
   # rules = association rules(frequent itemsets, metric="lift", min threshold=1.0)
   # Add district name to results
   rules['District'] = district
   return rules
# Process all districts and collect results
all_rules = []
for district in districts:
   try:
        district_rules = process_district(district)
       all_rules.append(district_rules)
   except Exception as e:
       print(f"Error processing district {district}: {e}")
# Combine results for all districts
all_rules_df = pd.concat(all_rules, ignore_index=True)
# Save results to a CSV file
output_path = '/content/district_apriori_rules.csv'
all_rules_df.to_csv(output_path, index=False)
print(f"Apriori results saved to {output_path}")
```

```
import pandas as pd
from mlxtend.frequent_patterns import apriori, association_rules
from sklearn.preprocessing import KBinsDiscretizer
# Load dataset
file path = '/content/Crop Data Final.csv'
data = pd.read_csv(file_path)
# Districts to process
districts = data['Dist Name'].unique()
# Binarization thresholds
binarize columns = [
    'RICE YIELD (Kg per ha)', 'WHEAT YIELD (Kg per ha)', 'SORGHUM YIELD (Kg per ha)',
    'PEARL MILLET YIELD (Kg per ha)', 'MAIZE YIELD (Kg per ha)', 'CHICKPEA YIELD (Kg per ha)',
    'PIGEONPEA YIELD (Kg per ha)', 'GROUNDNUT YIELD (Kg per ha)', 'SESAMUM YIELD (Kg per ha)',
    'OILSEEDS YIELD (Kg per ha)', 'SUGARCANE YIELD (Kg per ha)', 'COTTON YIELD (Kg per ha)',
    'Precipitation (mm)', 'Annual Rainfall'
1
# Function to discretize continuous variables
def discretize_data(df, columns):
    discretizer = KBinsDiscretizer(n bins=2, encode='ordinal', strategy='uniform')
    df = df.copy() # Prevent SettingWithCopyWarning
    for col in columns:
        if col in df.columns:
            df[col + '_Binned'] = discretizer.fit_transform(df[[col]]).astype(int)
    return df
# Function to generate transactions for Apriori
def generate transactions(df):
    transactions = pd.DataFrame()
    for col in df.columns:
        if col.endswith('_Binned'):
            transactions[col + '_High'] = (df[col] == 1).astype(int)
            transactions[col + ' Low'] = (df[col] == 0).astype(int)
    return transactions
# Function to process each district
def process_district(district):
    # Filter district-specific data and discretize
    district_data = data[data['Dist Name'] == district].copy()
    district data = discretize data(district data, binarize columns)
    # Generate transactions
    transactions = generate_transactions(district_data)
    # Run Apriori algorithm
    frequent_itemsets = apriori(transactions, min_support=0.2, use_colnames=True)
    num_itemsets = frequent_itemsets['itemsets'].apply(len).value_counts().to_dict()
    rules = association rules(frequent itemsets, metric="confidence", min threshold=0.5)
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# Add district name to results
       rules['District'] = district
       # Save rules for this district to its own CSV
       safe_district_name = district.replace(' ', '_').replace('/', '_')
       output path = f'/content/{safe_district_name}_apriori_rules.csv'
       rules.to csv(output path, index=False)
       print(f"Saved rules for district: {district} to {output path}")
       return rules
# Process all districts and collect results
all rules = []
for district in districts:
       try:
               district rules = process district(district)
               all rules.append(district rules)
       except Exception as e:
               print(f"Error processing district {district}: {e}")
# Combine results for all districts (optional)
all_rules_df = pd.concat(all_rules, ignore_index=True)
# Save combined results to a CSV file
output path = '/content/district apriori rules combined.csv'
all_rules_df.to_csv(output_path, index=False)
print(f"Combined Apriori results saved to {output path}")
🚁 /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ 🐣
         warnings.warn(
      Saved rules for district: Ahmednagar to /content/Ahmednagar_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Akola to /content/Akola_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Amarawati to /content/Amarawati apriori rules.csv
      /usr/local/lib/python 3.11/dist-packages/sklearn/preprocessing/\_discretization.py: 262: \ UserWarning: \ Feature \ 0 \ is \ constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ and \ will \ b \ UserWarning: \ Feature \ 0 \ is \ Constant \ And \
         warnings.warn(
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/association_rules.py:186: RuntimeWarning: invalid value encounter
        cert metric = np.where(certainty denom == 0, 0, certainty num / certainty denom)
      Saved rules for district: Aurangabad to /content/Aurangabad_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Beed to /content/Beed apriori rules.csv
      /usr/local/lib/python3.11/dist-packages/sklearn/preprocessing/_discretization.py:262: UserWarning: Feature 0 is constant and will b
         warnings.warn(
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/association_rules.py:186: RuntimeWarning: invalid value encounter
         cert_metric = np.where(certainty_denom == 0, 0, certainty_num / certainty_denom)
      Saved rules for district: Bhandara to /content/Bhandara_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
        warnings.warn(
      Saved rules for district: Buldhana to /content/Buldhana_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Chandrapur to /content/Chandrapur_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Dhule to /content/Dhule apriori rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Jalgaon to /content/Jalgaon_apriori_rules.csv
      /usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ
         warnings.warn(
      Saved rules for district: Kolhapur to /content/Kolhapur apriori rules.csv
```

/usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ warnings.warn(

Saved rules for district: Nagpur to /content/Nagpur_apriori_rules.csv

/usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ warnings.warn(

Saved rules for district: Nanded to /content/Nanded_apriori_rules.csv

/usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ warnings.warn(

Saved rules for district: Nasik to /content/Nasik_apriori_rules.csv

/usr/local/lib/python3.11/dist-packages/mlxtend/frequent_patterns/fpcommon.py:161: DeprecationWarning: DataFrames with non-bool typ warnings.warn(

Saved rules for district: Osmanabad to /content/Osmanabad_apriori_rules.csv