

Practical Exam: Spectrum Shades LLC

Spectrum Shades LLC is a prominent supplier of concrete color solutions, offering a wide range of pigments and coloring systems used in various concrete applications, including decorative concrete, precast concrete, and concrete pavers. The company prides itself on delivering high-quality colorants that meet the unique needs of its diverse clientele, including contractors, architects, and construction companies.

The company has recently observed a growing number of customer complaints regarding inconsistent color quality in their products. The discrepancies have led to a decline in customer satisfaction and a potential increase in product returns. By identifying and mitigating the factors causing color variations, the company can enhance product reliability, reduce customer complaints, and minimize return rates.

You are part of the data analysis team tasked with providing actionable insights to help Spectrum Shades LLC address the issues of inconsistent color quality and improve customer satisfaction.

Task 1

Before you can start any analysis, you need to confirm that the data is accurate and reflects what you expect to see.

It is known that there are some issues with the `production_data` table, and the data team have provided the following data description.

Write a query to return data matching this description. You must match all column names and description criteria.

Create a cleaned version of the dataframe.

- You should start with the data in the file "production_data.csv".
- Your output should be a dataframe named `clean_data`.
- All column names and values should match the table below.

Column Name	Criteria
<code>batch_id</code>	Discrete. Identifier for each batch. Missing values are not possible.
<code>production_date</code>	Date. Date when the batch was produced.
<code>raw_material_supplier</code>	Categorical. Supplier of the raw materials. (1='national_supplier', 2='international_supplier'). Missing values should be replaced with 'national_supplier'.
<code>pigment_type</code>	Nominal. Type of pigment used. ['type_a', 'type_b', 'type_c']. Missing values should be replaced with 'other'.
<code>pigment_quantity</code>	Continuous. Amount of pigment added (in kilograms) (Range: 1 - 100). Missing values should be replaced with median.

Column Name	Criteria
mixing_time	Continuous. Duration of the mixing process (in minutes). Missing values should be replaced with mean.
mixing_speed	Categorical. Speed of the mixing process represented as categories: 'Low', 'Medium', 'High'. Missing values should be replaced with 'Not Specified'.
product_quality_score	Continuous. Overall quality score of the final product (rating on a scale of 1 to 10). Missing values should be replaced with mean.

```
import pandas as pd
import numpy as np

# Task 1: Clean the data
def clean_data_frame():
    # Load the data
    data = pd.read_csv('production_data.csv')

    # Handle missing values
    data['pigment_quantity'].fillna(data['pigment_quantity'].median(), inplace=True)
    data['mixing_time'].fillna(data['mixing_time'].mean(), inplace=True)
    data['product_quality_score'].fillna(data['product_quality_score'].mean(), inplace=True)

    # Handle raw_material_supplier
    data['raw_material_supplier'] = data['raw_material_supplier'].map({1:
'national_supplier', 2: 'international_supplier'})

    # Handle pigment_type
    valid_types = ['type_a', 'type_b', 'type_c']
    data['pigment_type'].fillna('other', inplace=True)

    # Handle mixing_speed
    valid_speeds = ['Low', 'Medium', 'High']
    data['mixing_speed'].fillna('Not Specified', inplace=True)

    return data

clean_data = clean_data_frame()
print("Clean Data Sample:")
print(clean_data.head())
```

Clean Data Sample:

	batch_id	production_date	...	mixing_speed	product_quality_score
0	1	2024-06-25	...	High	7.165102
1	2	2023-11-23	...	High	6.849126
2	3	2024-02-18	...	High	5.661209
3	4	2023-11-11	...	High	6.991735
4	5	2024-04-11	...	High	7.095043

[5 rows x 8 columns]

Task 2

You want to understand how the supplier type and quantity of materials affect the final product attributes.

Calculate the average `product_quality_score` and `pigment_quantity` grouped by `raw_material_supplier`.

- You should start with the data in the file 'production_data.csv'.
- Your output should be a data frame named `aggregated_data`.
- It should include the three columns: `raw_material_supplier`, `avg_product_quality_score`, and `avg_pigment_quantity`.
- Your answers should be rounded to 2 decimal places.

Task 2

```
def aggregate_data(clean_data):
    aggregated_data = clean_data.groupby('raw_material_supplier').agg({
        'pigment_quantity': 'mean',
        'product_quality_score': 'mean'
    }).reset_index()

    return aggregated_data.round(2)
aggregated_data = aggregate_data(clean_data)
print(aggregated_data)
```

	raw_material_supplier	pigment_quantity	product_quality_score
0	international_supplier	34.91	5.97
1	national_supplier	44.73	8.02

Task 3

Identify all `product_quality_score` values for batches with a `raw_material_supplier` of 2 and a `pigment_quantity` greater than 35 kg. Use the original production data table, not the output of Task 2.

- You should start with the data in the file 'production_data.csv'.
- Your output should be a data frame named `pigment_data`.
- It should include the three columns: `raw_material_supplier`, `pigment_quantity`, and `product_quality_score`.

- Your answers should be rounded to 3 decimal places.

Task 3

Task 3: Filter data

```
def filter_data(clean_data):
    pigment_data = clean_data[
        (clean_data['raw_material_supplier'] == 'international_supplier') &
        (clean_data['pigment_quantity'] > 35)
    ][['raw_material_supplier', 'pigment_quantity', 'product_quality_score']]

    return pigment_data

filtered_pigment_data = filter_data(clean_data)
print(filtered_pigment_data)
```

	raw_material_supplier	pigment_quantity	product_quality_score
1	international_supplier	42.873479	6.849126
4	international_supplier	36.205108	7.095043
6	international_supplier	35.941439	5.735791
7	international_supplier	40.497203	5.510766
8	international_supplier	36.015111	4.959952
...
1985	international_supplier	40.484362	7.027202
1986	international_supplier	41.978065	6.735215
1994	international_supplier	37.483930	4.500927
1995	international_supplier	42.916500	5.043882
1998	international_supplier	46.148489	5.055166

[619 rows x 3 columns]

Task 4

In order to proceed with further analysis later, you need to analyze how various factors relate to product quality. Start by calculating the mean and standard deviation for the following columns: `pigment_quantity`, and `product_quality_score`.

These statistics will help in understanding the central tendency and variability of the data related to product quality.

Next, calculate the Pearson correlation coefficient between the following variables: `pigment_quantity`, and `product_quality_score`.

These correlation coefficients will provide insights into the strength and direction of the relationships between the factors and overall product quality.

- You should start with the data in the file 'production_data.csv'.
- Calculate the mean and standard deviation for the columns `pigment_quantity` and `product_quality_score` as: `product_quality_score_mean`, `product_quality_score_sd`, `pigment_quantity_mean`, `pigment_quantity_sd`.
- Calculate the Pearson correlation coefficient between `pigment_quantity` and `product_quality_score` as: `corr_coef`

- Your output should be a data frame named `product_quality`.
- It should include the columns: `product_quality_score_mean`, `product_quality_score_sd`, `pigment_quantity_mean`, `pigment_quantity_sd`, `corr_coef`.
- Ensure that your answers are rounded to 2 decimal places.

Task 4

Task 4: Calculate statistics

```
def calculate_statistics(clean_data):  
    stats = {  
        'product_quality_score_mean': clean_data['product_quality_score'].mean(),  
        'product_quality_score_sd': clean_data['product_quality_score'].std(),  
        'pigment_quantity_mean': clean_data['pigment_quantity'].mean(),  
        'pigment_quantity_sd': clean_data['pigment_quantity'].std(),  
        'corr_coef':  
clean_data['pigment_quantity'].corr(clean_data['product_quality_score'])  
    }
```