```
In [1]: ##Step1
        import pandas as pd
        #Load data
        df = pd.read csv("C:/Users/sridh/Desktop/Personal/Projects/Growth Analyst II Kerala Ayurveda Assignment/Amazon
        #Rename columns for consistency
        df.columns = df.columns.str.strip().str.replace(' ', '_').str.lower()
        #Convert 'date' to datetime
        df['date'] = pd.to datetime(df['date'])
        # Step 4: Clean columns with comma strings
        cols to clean = [
            'total_sales', 'organic_sales', 'ad sales',
            'total_traffic', 'organic_traffic', 'ad_traffic',
            'ad impressions', 'ad spends'
        1
        for col in cols_to_clean:
            df[col] = df[col].astype(str).str.replace(',', '').astype(float)
        #Check the result
        print(df.dtypes)
        print(df.head())
       product code
                                object
       date
                         datetime64[ns]
       total sales
                               float64
       organic sales
                                float64
       ad_sales
                               float64
       total traffic
                               float64
                              float64
       organic_traffic
                              float64
       ad traffic
       overall_units
                                 int64
      organic units
                                float64
       ad units
                               float64
       ad impressions
                               float64
       ad spends
                               float64
       dtype: object
        product_code
                          date total_sales organic_sales ad_sales \
       0 B07F5NCTN28 2024-04-01
                                    31500.0 13848.0 11262.0
       1 B07R3ZKB7D8 2024-04-01
                                     26680.0
                                                   16399.0
                                                              6436.0
                                                   9158.0
6919.0
         B07F5M62172 2024-04-01
                                     26600.0
                                                             13452.0
      3 B07P8FP14D3 2024-04-01
                                     12780.0
                                                              5330.0
       4 B07F5LZHVL1 2024-04-01
                                  11960.0
                                                  10764.0
                                                                 0.0
         total traffic organic traffic ad traffic overall units organic units \
       0
                1548.0
                                1279.55
                                            268.45
                                                              90
                                                                            52.0
                 774.0
                                677.58
                                             96.42
                                                                            35.0
                                                             95
                1086.0
                                822.23
       2
                                            263.77
                                                                            39.0
       3
                 626.0
                                 492.41
                                            133.59
                                                               71
                                                                            38.0
                                                              52
                 154.0
                                153.18
                                              0.82
                                                                            52.0
       4
         ad_units ad_impressions ad_spends
                   72433.0
24171.0
       0
             38.0
                                      5561.0
             23.0
                                     2186.0
       1
       2
             56.0
                         41932.0
                                     4935.0
       3
             33.0
                         26649.0
                                     1089.0
       4
              0.0
                            123.0
                                        5.0
In [4]: ##step2
        #Filter April and May data
        april = df[df['date'].dt.month == 4]
        may = df[df['date'].dt.month == 5]
        #Aggregate by month
        def compute_metrics(data):
            total sales = data['total sales'].sum()
            overall_units = data['overall_units'].sum()
            total traffic = data['total traffic'].sum()
            ad_spends = data['ad_spends'].sum()
           ad sales = data['ad sales'].sum()
            conversion rate = overall units / total traffic if total traffic else 0
            asp = total_sales / overall_units if overall_units else \theta
            roas = ad sales / ad spends if ad spends else 0
           acos = ad_spends / ad_sales if ad_sales else 0
            return {
                'Revenue (₹)': total_sales,
```

'Units Sold': overall_units,

```
'Conversion Rate': conversion_rate,
                 'ASP (₹)': asp,
                 'Ad Spend (₹)': ad_spends,
                'ROAS': roas,
                'ACOS': acos
            }
        april metrics = compute metrics(april)
        may_metrics = compute_metrics(may)
        # alculate deltas and % change
        mom table = pd.DataFrame([april metrics, may metrics], index=['April', 'May'])
        mom_table.loc['^ (Change)'] = mom_table.loc['May'] - mom_table.loc['April']
        mom table.loc['% Change'] = ((mom table.loc['May'] - mom table.loc['April']) / mom table.loc['April']) * 100
        # Show result
        print(mom_table)
                                                     Traffic Conversion Rate \
                    Revenue (₹)
                                   Units Sold
                   9.271914e+06 30755.000000 262250.00000
       April
                                                                     0.117274
                   9.354981e+06 30108.000000 222669.00000
                                                                     0.135214
       Mav
       ^ (Change) 8.306708e+04 -647.000000 -39581.00000
                                                                     0.017941
       % Change
                   8.959001e-01
                                    -2.103723
                                                  -15.09285
                                                                    15.298037
                      ASP (₹) Ad Spend (₹)
                                                    R0AS
                                                               AC0S
                   301.476623 832984.000000 2.742891 0.364579
       April
                   310.714117 723522.000000 3.287223 0.304208
9.237494 -109462.000000 0.544332 -0.060371
       Mav
       ^ (Change)
                                 -13.140949 19.845201 -16.559029
       % Change
                     3.064083
In [5]: ##step3
        import numpy as np
        # Add helper function to calculate decomposition for any stream
        def decompose_sales_change(traffic, units, sales):
            CR_apr = april[units].sum() / april[traffic].sum()
            CR_may = may[units].sum() / may[traffic].sum()
            ASP_apr = april[sales].sum() / april[units].sum()
            ASP_may = may[sales].sum() / may[units].sum()
            T_apr = april[traffic].sum()
            T_may = may[traffic].sum()
            # Chain rule log-decomposition
            \Delta \log sales = np.log(T_may * CR_may * ASP_may) - np.log(T_apr * CR_apr * ASP_apr)
            \Delta \log_{Taffic} = np.\log(T_may) - np.\log(T_apr)
            \Delta \log_{cr} = np.\log(CR_{may}) - np.\log(CR_{apr})
            \Delta_log_asp = np.log(ASP_may) - np.log(ASP_apr)
            return {
                 'Traffic %': (Δ log traffic / Δ log sales) * 100,
                 'Conversion Rate %': (Δ_log_cr / Δ_log_sales) * 100,
                 'ASP %': (Δ log asp / Δ log sales) * 100
            }
        #Run for Organic Sales
        organic_decomp = decompose_sales_change('organic_traffic', 'organic_units', 'organic_sales')
        #Run for Ad Sales
        ad decomp = decompose sales change('ad traffic', 'ad units', 'ad sales')
        #Display both
        print(" Organic Sales Decomposition:")
        print(pd.DataFrame([organic_decomp]))
        print("\n Ad Sales Decomposition:")
        print(pd.DataFrame([ad_decomp]))
        Organic Sales Decomposition:
            Traffic % Conversion Rate %
                             -921.735531 -171.690337
         1193.425867
        Ad Sales Decomposition:
           Traffic % Conversion Rate %
                                              ASP %
       0 -458.283742
                            490.427938 67.855804
In [7]: ##step4
        #Add ASP and Conversion columns
        df['conversion rate'] = df['overall units'] / df['total traffic']
        df['asp'] = df['total_sales'] / df['overall_units']
        #Separate April & May again
        april df = df[df['date'].dt.month == 4].copy()
        may_df = df[df['date'].dt.month == 5].copy()
```

'Traffic': total traffic,

```
#Group by ASIN (Product Code) and calculate aggregates
 def asin_monthly_summary(data):
     return data.groupby('product_code').agg({
         'total_sales': 'sum',
         'total_traffic': 'sum',
         'overall units': 'sum',
         'conversion rate': 'mean',
         'asp': 'mean'
     }).rename(columns={
         'total_sales': 'sales',
         'total traffic': 'traffic',
         'overall_units': 'units'
     })
 april summary = asin monthly summary(april df)
 may summary = asin monthly summary(may df)
 #Calculate deltas
 asin change = may summary - april summary
 asin change['conversion_rate_change'] = may_summary['conversion_rate'] - april_summary['conversion_rate']
 asin_change['asp_change'] = may_summary['asp'] - april_summary['asp']
 #Top ASINs by drop in Conversion Rate
 top_cr_drop = asin_change.sort_values('conversion_rate_change').head(10)
 #Top ASINs by drop in Traffic
 top traffic drop = asin change.sort values('traffic').head(10)
 #Outliers: ASINs with traffic > 1000 and units == 0
 outliers = may df.groupby('product code').agg({
     'total_traffic': 'sum',
'overall_units': 'sum'
 }).query('total traffic > 1000 and overall units == 0')
 print(" Top 10 ASINs with Conversion Rate Drop:\n", top_cr_drop[['conversion_rate_change']])
 print("\n Top 10 ASINs with Traffic Drop:\n", top traffic drop[['traffic']])
 print("\n Outlier ASINs (High Traffic, Zero Units):\n", outliers)
Top 10 ASINs with Conversion Rate Drop:
               conversion rate change
product code
B07R3.1.1GYW2
                                 -inf
B009YDXV904
                                 -inf
B07MSPKHCC7
                                 -inf
B07R4TV4SF5
                                 -inf
B07P6CCH3R5
                                -inf
B07R4TK5RB7
                                 -inf
B07P9K24MW1
                                -inf
B07P8FP3MN7
                                 -inf
B07H70N4SF7
                                -inf
B0823RGLLS2
                                 -inf
Top 10 ASINs with Traffic Drop:
               traffic
product_code
B07F5NCTN28 -12169.0
B07R3ZKB7D8
             -4050.0
B0DNM0VKFL5 -3677.0
B07P8FP14D3 -2647.0
B07F5M62172
             -2117.0
B07R4WTRN21
              -1890.0
B07CGNCPKT6
             -1780.0
B07PCPQ57N6 -1691.0
B07PBJNS5X9
             -1523.0
B07MQ63G912
             -1512.0
Outlier ASINs (High Traffic, Zero Units):
Empty DataFrame
Columns: [total_traffic, overall_units]
Index: []
```