```
import os
import pandas as pd
import numpy as np
from scipy.optimize import curve_fit, minimize, least_squares
import matplotlib.pyplot as plt
def profit_function(price, unit_cost, initial_demand, initial_price, elasticity):
    price = np.atleast_1d(price)
    price = np.clip(price, 0.5 * unit_cost, None) # Assuming unit cost is a
reasonable minimum
    demand = initial_demand * ((price / initial_price) ** elasticity)
    profit = (price - unit_cost) * demand
    return profit.ravel() # Ensure the return is 1-D array
def calculate_elasticity_coefficient(product_data):
    product_data.loc[:, 'Unit Cost'] = product_data['Unit Cost'].replace('[\$,]',
'', regex=True).astype(float)
    product_data.loc[:, 'Unit Price'] = product_data['Unit
Price'].replace('[\$,]', '', regex=True).astype(float)
    price_demand_data = product_data.groupby('Unit Price')['Order
Quantity'].sum().reset_index()
    def demand_model(x, a, b):
        return a * x**b
    popt, _ = curve_fit(demand_model, price_demand_data['Unit Price'],
price_demand_data['Order Quantity'])
    elasticity coefficient = popt[1] * 25
    return elasticity_coefficient
def optimize_price_demand(product_data, elasticity_coefficient):
    initial_price = product_data['Unit Price'].iloc[0]
    initial_demand = product_data['Order Quantity'].sum()
    unit_cost = product_data['Unit Cost'].iloc[0]
    learning_rate = 0.01
    num_iterations = 1000
    prices, profits = [initial price], []
    best_price, best_profit = initial_price, 0
    for _ in range(num_iterations):
        current_price = prices[-1]
        demand = initial_demand * ((current_price / initial_price) **
elasticity_coefficient)
        profit_gradient = demand + elasticity_coefficient * (current_price -
unit_cost) * demand / current_price
        updated_price = current_price + learning_rate * profit_gradient
        updated_profit = profit_function(updated_price, unit_cost, initial_demand,
initial_price, elasticity_coefficient)
        prices.append(updated price)
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profits.append(updated_profit)
        if updated_profit > best_profit:
            best_profit = updated_profit
            best_price = updated_price
   return initial_price, best_price, best_profit, prices, profits
# Using price-based gradient descent optimization
def optimize_price_only(product_data, elasticity_coefficient):
   initial_price = product_data['Unit Price'].iloc[0]
   initial_demand = product_data['Order Quantity'].sum()
   unit_cost = product_data['Unit Cost'].iloc[0]
   learning rate = 0.01
   num_iterations = 1000
   prices, profits = [initial_price], []
   best_price, best_profit = initial_price, 0
   for _ in range(num_iterations):
        current_price = prices[-1]
        profit_gradient = np.sum(-elasticity_coefficient * initial_demand *
((current_price / initial_price) ** (elasticity_coefficient - 1)))
        updated_price = current_price + learning_rate * profit_gradient
        updated_profit = profit_function(updated_price, unit_cost, initial_demand,
initial_price, elasticity_coefficient)
        prices.append(updated_price)
        profits.append(updated_profit)
        if updated profit > best profit:
            best profit = updated profit
            best_price = updated_price
   return initial_price, best_price, best_profit, prices, profits
def optimize_with_bfgs(product_data, elasticity_coefficient):
    initial_price = product_data['Unit Price'].iloc[0]
   prices, profits = [], []
   def callback(price):
        profit = profit_function(price, product_data['Unit Cost'].iloc[0],
product_data['Order Quantity'].sum(),
                                 product data['Unit Price'].iloc[0],
elasticity coefficient)
        prices.append(price[0])
        profits.append(profit[0])
   # Define the objective function to minimize
   def objective(price):
        return -profit_function(price, product_data['Unit Cost'].iloc[0],
product_data['Order Quantity'].sum(),
                                product_data['Unit Price'].iloc[0],
elasticity coefficient)
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# Bounds ensure price does not go below half of unit cost or unreasonably high
    bounds = [(0.5 * product_data['Unit Cost'].iloc[0], None)]
    # Minimize using BFGS with bounds
    result = minimize(objective, [initial_price], method='L-BFGS-B',
bounds=bounds, callback=callback)
    best price = result.x[0]
    best_profit = -result.fun
    return initial_price, best_price, best_profit, prices, profits
def optimize_with_kkt(product_data, elasticity_coefficient):
    initial_price = product_data['Unit Price'].iloc[0]
    prices, profits = [], []
    def callback(price):
        profit = profit_function(price, product_data['Unit Cost'].iloc[0],
product_data['Order Quantity'].sum(),
                                product_data['Unit Price'].iloc[0],
elasticity_coefficient)
        prices.append(price[0])
        profits.append(profit[0])
   # Define the objective function
    def objective(price):
        return -profit_function(price, product_data['Unit Cost'].iloc[0],
product_data['Order Quantity'].sum(),
                                product_data['Unit Price'].iloc[0],
elasticity_coefficient)
    # Define constraints
    cons = ({'type': 'ineq', 'fun': lambda price: price - 0.5 * product_data['Unit
Cost'].iloc[0]})
    # Minimize with constraints (KKT)
    result = minimize(objective, [initial_price], method='SLSQP',
constraints=cons, callback=callback)
    best_price = result.x[0]
    best_profit = -result.fun
    return initial_price, best_price, best_profit, prices, profits
# Update the analyze product and analyze all products methods to include these
optimizations
def analyze_product(product_id, data_path, output_folder, plot=False,
display_plot=False, method_type='GM0'):
    sales_data = pd.read_csv(data_path)
    product_data = sales_data[sales_data['_ProductID'] == product_id]
    elasticity_coefficient = calculate_elasticity_coefficient(product_data)
    if method type == 'BFGS':
        initial_price, best_price, best_profit, prices, profits =
optimize with bfgs(product data, elasticity coefficient)
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method_name = "BFGS"
    elif method_type == 'KKT':
        initial_price, best_price, best_profit, prices, profits =
optimize_with_kkt(product_data, elasticity_coefficient)
        method name = "KKT"
    elif method_type == "GM0":
        initial_price, best_price, best_profit, prices, profits =
optimize_price_only(product_data, elasticity_coefficient)
        method_name = "GradiantMethods_PriceOnly"
    elif method_type == "GM1":
        initial_price, best_price, best_profit, prices, profits =
optimize_price_demand(product_data, elasticity_coefficient)
        method_name = "GradiantMethods_PriceAndDemand"
    else:
        return "Wrong typed: method type"
    # Plotting logic if needed
    if plot:
        plt.figure(figsize=(12, 6))
        plt.subplot(1, 2, 1)
        plt.plot(prices, marker='o', linestyle='-')
        plt.title(f'Price Optimization Over Iterations ({method_name})')
        plt.xlabel('Iteration')
        plt.ylabel('Price ($)')
        plt.grid(True)
        plt.subplot(1, 2, 2)
        plt.plot(profits, marker='o', linestyle='-', color='red')
        plt.title(f'Profit Optimization Over Iterations ({method_name})')
        plt.xlabel('Iteration')
        plt.ylabel('Profit ($)')
        plt.grid(True)
        plt.tight_layout()
        plt.draw()
plt.savefig(f'{output_folder}/Product_{product_id}_{method_name}_Iteration.png')
        if display_plot:
            plt.show()
        plt.close()
    return product_data['Unit Price'].iloc[0], best_price, best_profit,
method name
def analyze_all_products(data_path, output_folder, method_type="GM0",
display_plots=False):
    # Create method-specific output folder
    method_output_folder = os.path.join(output_folder, method_type)
    if not os.path.exists(method_output_folder):
        os.makedirs(method_output_folder)
    sales_data = pd.read_csv(data_path)
    product ids = sales data[' ProductID'].unique()
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results = []
   for product_id in product_ids:
        initial_price, best_price, best_profit, method_name = analyze_product(
           product id, data path, method output folder, plot=True,
display_plot=display_plots, method_type=method_type)
        results.append([product_id, initial_price, best_price, best_profit,
method name])
   results_df = pd.DataFrame(results, columns=['ProductID', 'Initial Price',
'Best Price', 'Best Profit', 'Method'])
results_df.to_csv(f'{method_output_folder}/All_Products_Optimization_{method_name}
.csv')
def describe methods():
   print("Available Optimization Methods and their Usage:")
   print("----")
   print("1. Gradient Method (GM):")
   print(" - 'GM0' for PriceOnly Gradient Descent Optimization")
             - 'GM1' for Demand and Price Gradient Descent Optimization")
   print(" Use these methods to optimize price based on gradient descent
technique.")
   print("\n2. BFGS:")
   print(" - 'BFGS' for Broyden-Fletcher-Goldfarb-Shanno optimization
algorithm")
   print(" Use BFGS to find the local maximum of profit function using a quasi-
Newton method.")
   print("\n3. Karush-Kuhn-Tucker (KKT):")
   print(" - 'KKT' for optimization using Karush-Kuhn-Tucker conditions")
   print(" Use KKT when there are constraints that the solution needs to
satisfy.")
   print("\nFunction Usage:")
   print("----")
   print("analyze product(product id, data path, output folder, plot=False,
display_plot=False, method_type='GM0')")
   print(" - product_id: ID of the product to analyze.")
   print(" - data path: Path to the CSV file containing product data.")
   print(" - output folder: Directory to save the output plots and data files.")
   print(" - plot: Set to True to generate plots.")
   print(" - display plot: Set to True to display plots during execution.")
   print(" - method type: Specifies the optimization method to use.")
   print(" Options are 'GM0', 'GM1', 'BFGS', 'KKT'.")
   print("\nanalyze_all_products(data_path, output_folder, method_type='GMO',
display plots=False)")
   print(" - data_path: Path to the CSV file containing all products data.")
   print(" - output_folder: Directory to save the output plots and data files for
all products.")
    print(" - method_type: Specifies the optimization method to use for all
products.")
```

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print(" - display_plots: Set to True to display plots during execution.")
  print(" Options are 'GMO', 'GM1', 'BFGS', 'KKT'.")
  print("\nUse these functions to analyze product profitability under various
pricing scenarios using different optimization methods.")

# usage introduction:
describe_methods()
```

Available Optimization Methods and their Usage:

- 1. Gradient Method (GM):
 - 'GM0' for PriceOnly Gradient Descent Optimization
 - 'GM1' for Demand and Price Gradient Descent Optimization
 Use these methods to optimize price based on gradient descent technique.

2. BFGS:

- 'BFGS' for Broyden-Fletcher-Goldfarb-Shanno optimization algorithm Use BFGS to find the local maximum of profit function using a quasi-Newton method.
- Karush-Kuhn-Tucker (KKT):
 - 'KKT' for optimization using Karush-Kuhn-Tucker conditions
 Use KKT when there are constraints that the solution needs to satisfy.

Function Usage:

analyze_product(product_id, data_path, output_folder, plot=False,
display_plot=False, method_type='GM0')

- product_id: ID of the product to analyze.
- data path: Path to the CSV file containing product data.
- output_folder: Directory to save the output plots and data files.
- plot: Set to True to generate plots.
- display_plot: Set to True to display plots during execution.
- method_type: Specifies the optimization method to use.
 Options are 'GM0', 'GM1', 'BFGS', 'KKT'.

analyze_all_products(data_path, output_folder, method_type='GM0',
display plots=False)

- data_path: Path to the CSV file containing all products data.
- output_folder: Directory to save the output plots and data files for all products.
 - method_type: Specifies the optimization method to use for all products.
- display_plots: Set to True to display plots during execution.
 Options are 'GMO', 'GM1', 'BFGS', 'KKT'.

Use these functions to analyze product profitability under various pricing scenarios using different optimization methods.

```
analyze_all_products('US_Regional_Sales_Data.csv', 'result_images', "GM0")

analyze_all_products('US_Regional_Sales_Data.csv', 'result_images', "GM1")

analyze_all_products('US_Regional_Sales_Data.csv', 'result_images', "BFGS")

analyze_all_products('US_Regional_Sales_Data.csv', 'result_images', "KKT")
```