1.1_main_script_gas

March 31, 2025

[]: import numpy as np

```
import os
     import pymrio
     import pandas as pd
     from IPython.display import display
     import country converter as coco
     from scipy.sparse.linalg import inv
     import scipy.sparse as sp
     import matplotlib.pyplot as plt
[]: pd.set_option('display.max_rows', 200)
     pd.set_option('display.max_columns', 100)
     pd.set_option('display.width', 1000)
     pd.set_option('display.max_colwidth', 100)
[]: # Define EU28 country codes
     eu28_countries = [
         "AT", "BE", "BG", "CY", "CZ", "DE", "DK", "EE", "ES", "FI", "FR", "GB", [
      ⇔"GR", "HR", "HU",
         "IE", "IT", "LT", "LU", "LV", "MT", "NL", "PL", "PT", "RO", "SE", "SI". "SK"
     ]
[]: | # Define the storing folder for Exiobase3 data
     exio3_folder = 'C:/Users/danie/Nextcloud/Coding/Masterthesis/exiobase'
     download_folder = os.path.join(exio3_folder, 'exio_download')
     # Check if the exio_download folder exists, create if not
     if not os.path.exists(download folder):
         os.makedirs(download folder)
         print(f"Created directory: {download folder}")
     else:
         print(f"Directory already exists: {download_folder}")
     # Download Exiobase3 data to the specified folder
     exio_downloadlog = pymrio.download_exiobase3(storage_folder=download_folder,_u
      →system="ixi", years=[2018, 2019, 2020, 2021, 2022])
     print(exio_downloadlog)
```

```
[]: # Parse Exiobase3 (2021) data
    exio3 = pymrio.parse_exiobase3(path='C:/Users/danie/Nextcloud/Coding/
      →Masterthesis/exiobase/exio_download/IOT_2021_ixi.zip')
[]: # Assess meta data
    print(exio3.meta)
[]: ### Check for geographical sampling differences between FIGARO and EXIOBASE 31
    # FIGARO countries
    figaro_countries = [
        'AR', 'AT', 'AU', 'BE', 'BG', 'BR', 'CA', 'CH', 'CN', 'CY', 'CZ', 'DE',
     'GR', 'HR', 'HU', 'ID', 'IE', 'IN', 'IT', 'JP', 'KR', 'LT', 'LU', 'LV', L
     ↔ 'MT', 'MX', 'NL', 'NO', 'PL', 'PT', 'RO', 'RU',
        'SA', 'SE', 'SI', 'SK', 'TR', 'US', 'ZA'
    # Extract country codes from EXIOBASE 3 dataset
    exio countries = exio3.get regions()
    # Compare country codes
    common_countries = sorted(set(figaro_countries).intersection(exio_countries))
    figaro_only_countries = sorted(set(figaro_countries) - set(exio_countries))
    exio_only_countries = sorted(set(exio_countries) - set(figaro_countries))
    print("Common countries:", common_countries)
    print("Countries only in FIGARO:", figaro_only_countries)
    print("Countries only in EXIOBASE 3:", exio_only_countries)
[]: # Argentina and Saudi Arabia are not in EXIOBASE 3, but in FIGARO
    # Taiwan is in EXIOBASE 3, but not in FIGARO
    # FIGW1 is ROW in Figaro
    # WA (Asia), WE (Europe), WF (Africa), WL (Latin America), WM (Middle East), WPu
     → (Pacific) are ROW regions in Exiobase3
    # Collect all RoW regions and Taiwan in one FIGARO region
    exio3.rename_regions({'WA': 'FIGW1', 'WE': 'FIGW1', 'WF': 'FIGW1', 'WL':
     # Aggregate EXIOBASE 3 data to FIGARO regions
    exio3.aggregate_duplicates(inplace=True)
    exio3.Z.to_csv('C:/Users/danie/Nextcloud/Coding/Masterthesis/exiobase/
     ⇔country_agg_exio3_figaro.csv')
```

```
⇔names and aggregation levels
    mrio_class = pymrio.get_classification(mrio_name='exio3_ixi')
[]: # Display the full mrio_class
    display(mrio_class)
[]: # Create a conversion dictionnary from ExioName to ExioLabel and check for
     ⇔correctness by displaying it
    conv_dict = mrio_class.get_sector_dict(mrio_class.sectors.ExioName, mrio_class.
     ⇒sectors.ExioLabel)
    display(conv_dict)
    # Rename sectors in the pymrio object
    exio3.rename_sectors(conv_dict)
    # Check if the renaming was successful
    print(exio3.Z.index)
[]: ### Aggregate Exiobase3 data ###
    # Done through renaming, which also helps to adapt it to eurostat data
    # Renaming of sectors requires mappping of ExioLabel to NACE classification
    rename_dict_exio3_NACE = {
        "A PARI": "A01",
        "A_WHEA": "A01",
        "A_OCER": "A01",
        "A_FVEG": "A01",
        "A_OILS": "A01",
        "A_SUGB": "A01",
        "A_FIBR": "A01",
        "A_OTCR": "A01",
        "A_CATL": "A01",
        "A_PIGS": "A01",
        "A_PLTR": "A01",
        "A_OMEA": "A01",
        "A_OANP": "AO1",
        "A MILK": "A01",
        "A_WOOL": "A01",
        "A_MANC": "A01",
        "A_MANB": "AO1",
        "A_FORE": "A02",
        "A_FISH": "A03",
        "A_GASE": "B_gas",
        "A_OGPL": "B_gas",
        "A_COAL": "B_nongas",
```

```
"A_COIL": "B_nongas",
"A_ORAN": "B_nongas",
"A_IRON": "B_nongas",
"A_COPO": "B_nongas",
"A_NIKO": "B_nongas",
"A_ALUO": "B_nongas",
"A_PREO": "B_nongas",
"A_LZTO": "B_nongas",
"A_ONFO": "B_nongas",
"A_STON": "B_nongas",
"A_SDCL": "B_nongas",
"A_CHMF": "B_nongas",
"A_PCAT": "C10-12",
"A_PPIG": "C10-12",
"A_PPLT": "C10-12",
"A_POME": "C10-12",
"A_VOIL": "C10-12",
"A_DAIR": "C10-12",
"A_RICE": "C10-12",
"A_SUGR": "C10-12",
"A_OFOD": "C10-12",
"A BEVR": "C10-12",
"A_FSHP": "C10-12",
"A TOBC": "C10-12",
"A TEXT": "C13-15",
"A GARM": "C13-15",
"A_LETH": "C13-15",
"A WOOD": "C16",
"A_WOOW": "C16",
"A_PULP": "C17",
"A_PAPR": "C17",
"A_PAPE": "C17",
"A_MDIA": "C18",
"A_COKE": "C19",
"A_REFN": "C19",
"A_PLAS": "C20_21",
"A PLAW": "C20 21",
"A_NFER": "C20_21",
"A PFER": "C20 21",
"A_CHEM": "C20_21",
"A_RUBP": "C22",
"A GLAS": "C23",
"A_GLAW": "C23",
"A_CRMC": "C23",
"A_BRIK": "C23",
"A_CMNT": "C23",
"A_ASHW": "C23",
```

```
"A_ONMM": "C23",
"A_NUCF": "C24",
"A STEL": "C24".
"A_STEW": "C24",
"A_PREM": "C24",
"A_PREW": "C24",
"A ALUM": "C24",
"A_ALUW": "C24",
"A_LZTP": "C24",
"A LZTW": "C24",
"A COPP": "C24",
"A_COPW": "C24",
"A_ONFM": "C24",
"A_ONFW": "C24",
"A_METC": "C24",
"A_FABM": "C25-33",
"A_MACH": "C25-33",
"A_OFMA": "C25-33",
"A_ELMA": "C25-33",
"A_RATV": "C25-33",
"A_MEIN": "C25-33",
"A MOTO": "C25-33",
"A_OTRE": "C25-33",
"A FURN": "C25-33",
"A_POWC": "D35",
"A POWG": "D35",
"A_POWN": "D35",
"A POWH": "D35",
"A_POWW": "D35",
"A_POWP": "D35",
"A_POWB": "D35",
"A_POWS": "D35",
"A_POWE": "D35",
"A_POWO": "D35",
"A_POWM": "D35",
"A_POWZ": "D35",
"A POWT": "D35",
"A_POWD": "D35",
"A GASD": "D35",
"A_HWAT": "D35",
"A_WATR": "E36",
"A RYMS": "E37-39",
"A_BOTW": "E37-39",
"A_INCF": "E37-39",
"A_INCP": "E37-39",
"A_INCL": "E37-39",
"A_INCM": "E37-39",
```

```
"A_INCT": "E37-39",
"A_INCW": "E37-39",
"A INCO": "E37-39".
"A_BIOF": "E37-39",
"A_BIOP": "E37-39",
"A_BIOS": "E37-39",
"A_COMF": "E37-39",
"A_COMW": "E37-39",
"A_WASF": "E37-39",
"A WASO": "E37-39",
"A LANF": "E37-39",
"A_LANP": "E37-39",
"A_LANL": "E37-39",
"A_LANI": "E37-39",
"A_LANT": "E37-39",
"A_LANW": "E37-39",
"A_CONS": "F",
"A_CONW": "F",
"A_TDMO": "G45",
"A_TDWH": "G46",
"A_TDFU": "G47",
"A TDRT": "G47",
"A_TRAI": "H49",
"A TLND": "H49",
"A TPIP": "H49",
"A TWAS": "H50",
"A_TWAI": "H50",
"A TAIR": "H51",
"A_TAUX": "H52",
"A_PTEL": "H53",
"A_HORE": "I",
"A_COMP": "J",
"A_FINT": "K64",
"A_FINS": "K65",
"A_FAUX": "K66",
"A_REAL": "L68",
"A_RESD": "M_N",
"A_OBUS": "M_N",
"A MARE": "M N",
"A_PADF": "084",
"A_EDUC": "P85",
"A HEAL": "Q",
"A_RECR": "R_S",
"A_ORGA": "R_S",
"A_OSER": "R_S",
"A_PRHH": "T",
"A_EXTO": "U"
```

```
}
[]: # Apply mapping with the rename_sectors tool of pymrio
     exio3.rename_sectors(rename_dict_exio3_NACE)
     print(exio3.Z.index)
     # Aggregate duplicates
     exio3.aggregate_duplicates()
     print(exio3.Z)
[]: exio_Z_df = exio3.Z
     display(exio_Z_df)
[]: | # Extract energy inputs for B_gas and B_nongas (rows)
     energy_inputs = exio_Z_df.loc[(slice(None), ['B gas', 'B nongas']), :]
     # Compute total energy input per row (sum of B_gas and B_nongas)
     total_energy_inputs_row = energy_inputs.groupby(level=0).sum()
     total_energy_inputs_row.replace(0, np.nan, inplace=True) # Avoid division by_
      ⇒zero
     # Compute row-wise shares (only for B_gas and B_nongas rows)
     energy_shares_row = energy_inputs.div(total_energy_inputs_row, level=0).
      ofillna(0)
     # Create a copy of the original DataFrame to preserve structure
     exio_Z_with_row_shares = exio_Z_df.copy()
     # Add energy share values to the original dataframe (only updating B gas and )
      \rightarrow B_nongas rows)
     exio_Z_with_row_shares.loc[energy_shares_row.index] = energy_shares_row
     # Display the full table with updated B_gas and B_nongas rows
     print(exio_Z_with_row_shares)
[]: # Create a copy of the DataFrame to store both row-wise and column-wise shares
     exio_Z_with_column_shares = exio_Z_df.copy()
     # Identify\ energy-related\ columns\ (MultiIndex:\ first\ level=country,\ second_{\sqcup}
      \rightarrow level = B_{gas}/B_{nongas})
     energy_columns = ['B_gas', 'B_nongas']
     # Iterate over each country in the column MultiIndex
     for country in exio_Z_df.columns.levels[0]: # First level = country names
```

```
# Define MultiIndex column tuples for the current country's B gas and
      \hookrightarrow B_nongas
         country_energy_cols = [(country, col) for col in energy_columns]
         # Ensure the country has both B_gas and B_nongas columns (avoid KeyErrors)
         existing_energy_cols = [col for col in country_energy_cols if col in_
      \rightarrowexio_Z_df.columns]
         if len(existing_energy_cols) < 2: # Skip countries missing B_gas or_
      \hookrightarrow B_nongas
             continue
         # Compute total energy input per row (sum of only that country's B_gas and_
      \hookrightarrow B nongas values)
         total_energy_inputs_row = exio_Z_df.loc[:, existing_energy_cols].sum(axis=1)
         # Avoid division by zero
         total_energy_inputs_row.replace(0, np.nan, inplace=True)
         # Compute column-wise shares for that country's B gas and B nongas
         energy_shares_col = exio_Z_df.loc[:, existing_energy_cols].
      →div(total_energy_inputs_row, axis=0).fillna(0)
         # Store the computed shares into the existing table
         exio Z with column shares loc[:, existing energy cols] = energy shares col
     # Display the final DataFrame
     print(exio_Z_with_column_shares)
[]: \# A condition has to be added, that if both B_gas and B_nongas are zero,
      \rightarrow B_nongas should be 1 and B_gas should be 0
[]: # Step 1: Create a new DataFrame with the same structure, filled with 1s
     exio_Z_final = pd.DataFrame(1, index=exio_Z_df.index, columns=exio_Z_df.columns)
     # Step 2: Identify B_qas and B_nongas rows and columns
     row_sectors = ["B_gas", "B_nongas"]
     column_sectors = ["B_gas", "B_nongas"]
     # Step 3: Apply row-wise shares **excluding B quadrants**
     exio_Z_final.loc[exio_Z_df.index.get_level_values(1).isin(row_sectors),
                       ~exio_Z_df.columns.get_level_values(1).isin(column_sectors)] =_
      →(
         exio_Z_with_row_shares.loc[exio_Z_df.index.get_level_values(1).
      →isin(row_sectors),
```

```
~exio_Z_df.columns.get_level_values(1).
 ⇒isin(column_sectors)].values
# Step 4: Apply column-wise shares **excluding B quadrants**
exio Z final.loc[~exio Z df.index.get level values(1).isin(row sectors),
                 exio_Z_df.columns.get_level_values(1).isin(column_sectors)] = (
    exio_Z_with_column_shares.loc[~exio_Z_df.index.get_level_values(1).
 ⇔isin(row_sectors),
                                  exio_Z_df.columns.get_level_values(1).
 →isin(column_sectors)].values
# Step 5: Assign B quadrant shares for each (supplier country, consuming
 ⇔country) pair
for supplier_country in exio_Z_df.index.get_level_values(0).unique():
    for consuming_country in exio_Z_df.columns.get_level_values(0).unique():
        # Identify the specific 2x2 quadrant for this country pair
        supplier rows = (exio Z df.index.get level values(0) ==___
 supplier_country) & (exio_Z_df.index.get_level_values(1).isin(row_sectors))
        consumer_cols = (exio_Z_df.columns.get_level_values(0) ==__
 →consuming_country) & (exio_Z_df.columns.get_level_values(1).
 ⇔isin(column_sectors))
        # Extract the total energy input for this quadrant from original exio3.Z
        b_quadrant_values = exio3.Z.loc[supplier_rows, consumer_cols]
        # Sum total energy input for this quadrant
        total_energy_quadrant = b_quadrant_values.sum().sum()
        # Normalize: Compute each cell's share of the total energy input
        if total_energy_quadrant > 0: # Avoid division by zero
            b_quadrant_shares = b_quadrant_values / total_energy_quadrant
        else:
            b_quadrant_shares = b_quadrant_values * 0 # If no energy input,_
 ⇔set to zero
        # Apply the modification: If both B_{gas} and B_{nongas} are 0, set
 \hookrightarrow B_nongas = 1, B_gas = 0
        if (b_quadrant_shares.loc[(supplier_country, "B_gas"),__
 ⇔(consuming_country, "B_gas")] == 0 and
            b_quadrant_shares.loc[(supplier_country, "B_nongas"),__
 ⇔(consuming country, "B nongas")] == 0):
```

```
b_quadrant_shares.loc[(supplier_country, "B_nongas"),__
      ⇔(consuming_country, "B_nongas")] = 1
                 b_quadrant_shares.loc[(supplier_country, "B_gas"),__
      ⇔(consuming country, "B gas")] = 0
             # Insert the calculated shares back into the final DataFrame
             exio_Z_final.loc[supplier_rows, consumer_cols] = b_quadrant_shares.
      ⇔values
     # Step 6: Done! The final DataFrame now correctly integrates all three parts
     print(exio_Z_final)
[]: # Save the final DataFrame
     exio_Z_final.to_csv('C:/Users/danie/Nextcloud/Coding/Masterthesis/exiobase/
      →exio3_ixi_2021_energy_shares.csv')
[]: ### Repeat similar proedure for the Y matrix (demand) as a prerequesite for cpiu
      ⇔weighing ###
[]: # Insepct the Y matrix
     display(exio3.Y)
     exio_Y_df = exio3.Y
[]: # Extract private household demand for B_gas and B_nongas (rows)
     energy_inputs = exio_Y_df.loc[(slice(None), ['B_gas', 'B_nongas']), :]
     # Compute demand per row (sum of B_qas and B_nonqas)
     total_energy_inputs_row = energy_inputs.groupby(level=0).sum()
     total_energy_inputs_row.replace(0, np.nan, inplace=True) # Avoid division by_
      \rightarrow zero
     # Compute row-wise shares (only for B_gas and B_nongas rows)
     energy_shares_row = energy_inputs.div(total_energy_inputs_row, level=0).
      ofillna(0)
     # Create a copy of the original DataFrame to preserve structure
     exio_Y_row_shares = exio_Y_df.copy()
     # Add energy share values to the original dataframe (only updating B gas and I
      \hookrightarrow B nongas rows)
     exio_Y_row shares.loc[energy shares_row.index] = energy shares_row
     # Filter columns to only include 'Final consumption expenditure by households'
     exio_Y_row_shares = exio_Y_row_shares.xs('Final consumption expenditure by_
      ⇔households', axis=1, level=1, drop_level=False)
```

```
# Display the filtered table with updated B_gas and B_nongas rows
display(exio_Y_row_shares)
```

```
[]: ### Introducing the Figaro data ###
     # Load the Figaro data and prepare for further usage
     # Split the index and columns of the Figaro into a MultiIndex
     def split_index_to_multiindex(df):
         11 11 11
         Split the index and columns of the DataFrame into a MultiIndex.
         The index and columns are expected to have a structure like,
      ⇔'XX_sector_code'.
         11 11 11
         def split_index(index):
             return pd.MultiIndex.from_tuples([tuple(i.split('_', 1)) for i in_
      →index], names=['Country', 'Sector'])
         df.index = split_index(df.index)
         df.columns = split_index(df.columns)
         return df
     def process_files_and_split_index(input_dir, output_dir):
         Process all CSV files in the input folder, split the index and columns into
      \hookrightarrow a MultiIndex,
         and save the new files in the specified output directory with a_{\sqcup}
      → 'multiindex ' prefix.
         # Ensure the output directory exists
         if not os.path.exists(output_dir):
             os.makedirs(output_dir)
         for filename in os.listdir(input_dir):
             if filename.endswith(".csv"):
                 file_path = os.path.join(input_dir, filename)
                 print(f"Processing {filename}...")
```

```
df = pd.read_csv(file_path, index_col=0)
                df = split_index_to_multiindex(df)
                 # Save the modified DataFrame to the output directory with
      → 'multiindex_ ' prefix
                output_file_path = os.path.join(output_dir,_
      df.to_csv(output_file_path)
                print(f"Processed and saved {filename} to {output_file_path}")
     # Example usage
    input_dir = 'C:/Users/danie/Nextcloud/Coding/Masterthesis/data/raw/
     ⇔figaro_tables'
    output_dir = 'C:/Users/danie/Nextcloud/Coding/Masterthesis/notebooks/
     ⇔NB_exio3_figaro_gas/figaro_multi_index'
    process_files_and_split_index(input_dir, output_dir)
[]: # Define the file path for the 2021 multiindex figure table
    file_path_2021 = os.path.join(output_dir, 'multiindex_2021_figaro_64.csv')
     # Load the 2021 multiindex figaro table
    df_2021 = pd.read_csv(file_path_2021, index_col=[0, 1], header=[0, 1])
     # Display the first few rows of the dataframe to verify
    print(df_2021.head())
[]: | ### Apply sector mapping to the Figaro data ###
     # 1. Sector Mapping (Renaming + Aggregation)
    sector_mapping = {
        "C10T12": "C10-12",
        "C13T15": "C13-15",
        "C20": "C20_21",
        "C21": "C20_21",
        "E37T39": "E37-39",
        "J58": "J", "J59_60": "J", "J61": "J", "J62_63": "J",
        "M69_70": "M_N", "M71": "M_N", "M72": "M_N", "M73": "M_N", "M74_75": "M_N",
        "N77": "M_N", "N78": "M_N", "N79": "M_N", "N80T82": "M_N",
        "Q86": "Q", "Q87_88": "Q",
        "R90T92": "R_S", "R93": "R_S", "S94": "R_S", "S95": "R_S", "S96": "R S",
        "1.": "1.68"
    }
```

```
# 3. Function to Apply Sector Mapping and Aggregate
     def apply_sector_mapping(df, sector_mapping):
         Rename and aggregate sectors in both rows and columns using the provided \sqcup
      \hookrightarrow mapping.
         HHHH
         # Step 1: Rename Row Index (Industries)
         new_row_index = [(country, sector_mapping.get(sector, sector)) for country,__
      ⇒sector in df.index]
         df.index = pd.MultiIndex.from_tuples(new_row_index, names=['Country',_
      # Step 2: Rename Column Index (Industries)
         new_col_index = [(country, sector_mapping.get(sector, sector)) for country,_
      ⇒sector in df.columns]
         df.columns = pd.MultiIndex.from_tuples(new_col_index, names=['Country',_

¬'Sector'])
         # Step 3: Aggregate Mapped Sectors
         df = df.groupby(level=['Country', 'Sector']).sum() # Aggregate rows
         df = df.groupby(level=['Country', 'Sector'], axis=1).sum() # Aggregate_\( \)
      \hookrightarrow columns
         return df
[]: # -----
     # 4. Apply Sector Mapping to FIGARO Data
     df_2021_mapped = apply_sector_mapping(df_2021, sector_mapping)
     # Display the first few rows to verify the aggregation worked
     display(df_2021_mapped)
     df_2021_mapped.to_csv('C:/Users/danie/Nextcloud/Coding/Masterthesis/notebooks/
      →NB_exio3_figaro_gas/figaro_mapped.csv')
[]: ### Merge countries in the Figaro data to match the Exiobase 3 data ###
     def merge_countries(df, countries_to_merge, target='FIGW1'):
         Relabels rows and columns so that any country in countries to merge is_{\sqcup}
      \negreplaced by target.
         Then groups by the MultiIndex to sum the duplicated entries.
```

```
Assumes both rows and columns are MultiIndex with levels ['Country', ]

    'Sector'].
         Parameters:
           df: pd.DataFrame with MultiIndex for both rows and columns
           countries_to_merge: list of country codes to merge (e.g., ['AR', 'SA'])
           target: the target country code to absorb the values (default 'FIGW1')
         Returns:
           A DataFrame with the specified countries merged into the target.
         # Save the original index names (should be ['Country', 'Sector'])
         row_index_names = df.index.names
         col_index_names = df.columns.names
         # --- Relabel row index: Replace countries in countries to merge with target
         new_row_index = [
             (target if country in countries_to_merge else country, sector)
             for country, sector in df.index
         df.index = pd.MultiIndex.from_tuples(new_row_index, names=row_index_names)
         # --- Relabel column index: Replace countries in countries to merge with
      \hookrightarrow target
         new_col_index = [
             (target if country in countries_to_merge else country, sector)
             for country, sector in df.columns
         df.columns = pd.MultiIndex.from_tuples(new_col_index, names=col_index_names)
         \# --- Group by the MultiIndex levels to aggregate duplicate entries \sqcup
      ⇔(summing over duplicates)
         df = df.groupby(level=row_index_names).sum()
         df = df.groupby(axis=1, level=col_index_names).sum()
         return df
     merged_df = merge_countries(df_2021_mapped, ['AR', 'SA'])
     display(merged_df)
[]: # Add gross ouput row
     def add_gross_output_row(df):
         gross_output = df.sum(axis=0)
```

gross_output.name = ('GO', 'GO')

```
df = pd.concat([df, pd.DataFrame(gross_output).T])
         return df
     df_2021_rdy = add_gross_output_row(merged_df)
     display(df_2021_rdy)
[]: df 2021 rdy.to csv('C:/Users/danie/Nextcloud/Coding/Masterthesis/notebooks/
     →NB_exio3_figaro_gas/figaro_mapped_2021_merged.csv')
[]: ### Calculation of CPI weights ###
     import pandas as pd
     def prepare_cpi_weight_data(df):
         Prepares the FIGARO final consumption data for CPI weight calculation.
         This involves:
         - Extracting only `P3_S14` columns
         - Removing the `GO` row
         - Duplicating `B` sector rows into `B_gas` and `B_nongas`
         Parameters:
           df : pd.DataFrame
               The FIGARO final demand table with MultiIndex (Country, Sector)
         Returns:
           pd.DataFrame
               Modified DataFrame with `B` sector split into `B_gas` and `B_nongas`,\Box
      ⇔without `GO`
         11 11 11
         final sector = "P3 S14"
         # **Step 1: Extract only the relevant final consumption column**
         df_fc = df.xs(final_sector, axis=1, level=1, drop_level=False)
         print(f" Extracted only `{final_sector}` from FIGARO.")
         # **Step 2: Exclude the `GO` row**
         df_fc = df_fc[df_fc.index.get_level_values(1) != "GO"]
         print(" `GO` row successfully removed.")
         # **Step 3: Identify `B` sector rows**
         b_rows = df_fc[df_fc.index.get_level_values(1) == "B"].copy()
         # **Step 4: Duplicate and rename `B` sector rows**
         b_rows_gas = b_rows.copy()
```

```
b_rows_nongas = b_rows.copy()
         b_rows_gas.index = [(c, "B_gas") for c, _ in b_rows.index]
         b_rows_nongas.index = [(c, "B_nongas") for c, _ in b_rows.index]
         # **Step 5: Remove original `B` row and add new ones**
         df_fc = df_fc[df_fc.index.get_level_values(1) != "B"]
         df_fc = pd.concat([df_fc, b_rows_gas, b_rows_nongas]).sort_index()
         print(" `B` sector successfully split into `B gas` and `B nongas`.")
         return df_fc
     # **Run the preparation function**
     df_cpi_ready = prepare_cpi_weight_data(df_2021_rdy)
[]: display(df_cpi_ready)
[]: print(exio_Y_row_shares.index)
     print(exio_Y_row_shares.columns)
     print(df_cpi_ready.index)
     print(df_cpi_ready.columns)
[]: import pandas as pd
     def apply_energy_shares(df_cpi_ready, energy_shares_df):
         Apply energy shares from `exio_Y_row_shares` to `B_gas` and `B_nongas` in ∪
      \hookrightarrow `df_cpi_ready`.
         Fixes:
            Ensures correct index naming for `df_cpi_ready`.
            Extracts the energy shares from the correct **column**, not index.
         Parameters:
           df_cpi_ready : pd.DataFrame
               FIGARO final consumption table with 'B gas' and 'B nongas' rows.
           energy_shares_df : pd.DataFrame
               EXIOBASE Y matrix shares for energy sectors.
         Returns:
           pd.DataFrame
               Updated `df_cpi_ready` with adjusted `B_gas` and `B_nongas` values.
         11 11 11
         # Step 1: Ensure `df_cpi_ready` has correct index names
         df_cpi_ready.index.set_names(["Country", "Sector"], inplace=True)
         # Step 2: Ensure energy shares dataframe has correct index names
```

```
¬"sector": "Sector", "category": "Sector"})
         # Step 3: Extract the column `"Final consumption expenditure by "
      →households"` from the correct **country-wise** level
        final_sector_exio = "Final consumption expenditure by households"
        if final_sector_exio not in energy_shares_df.columns.get_level_values(1):
            raise KeyError(f" `{final_sector_exio}` not found in⊔
      → `exio_Y_row_shares` columns! Available: {energy_shares_df.columns.

¬get_level_values(1).unique()}")
        # **Extract Energy Shares Per Country**
        energy_shares = energy_shares_df.xs(final_sector_exio, axis=1, level=1)
        energy_shares.columns = pd.MultiIndex.from_product([energy_shares.columns,_
      print(energy_shares.index)
        print(energy_shares.columns)
        print(f" Extracted energy shares for '{final_sector_exio}' from_
      ⇔`exio Y row shares`.")
        # Step 4: Identify affected rows (B_gas and B_nongas)
        energy_rows = df_cpi_ready.index.get_level_values(1).isin(['B_gas',_

¬'B_nongas'])
        # Step 5: Apply energy shares **only to B_gas and B_nongas**, matching by
      \hookrightarrow country
        df_cpi_ready.loc[energy_rows] *= energy_shares
        print(" Energy shares successfully applied to `B_gas` and `B_nongas`.")
        return df_cpi_ready
     # Apply the energy shares BEFORE CPI calculation
    df_cpi_ready = apply_energy_shares(df_cpi_ready, exio_Y_row_shares)
[]: ### Calculate CPI weights ###
    def calculate_cpi_weights(df_cpi_ready):
        Compute CPI weights by normalizing each column.
```

energy_shares_df = energy_shares_df.rename_axis(index={"region": "Country",__

```
Steps:
       Divide each cell by the total column sum.
       Preserve MultiIndex structure.
       Handle cases where column sums are zero (avoid division errors).
   Parameters:
     df_cpi_ready : pd.DataFrame
         The adjusted FIGARO final consumption table with energy shares 
 \hookrightarrow applied.
   Returns:
     pd.DataFrame
         CPI weights as a normalized version of `df_cpi_ready`.
    11 11 11
   # Step 1: Compute column sums (total final consumption per country)
   column_sums = df_cpi_ready.sum(axis=0)
   # Step 2: Avoid division by zero (replace zero sums with NaN to preventu
 ⇔errors)
   column_sums.replace(0, pd.NA, inplace=True)
   # Step 3: Normalize each column (divide each cell by its column sum)
   cpi_weights = df_cpi_ready.div(column_sums, axis=1)
   print(" CPI weights successfully calculated.")
   return cpi_weights
# Apply CPI weight calculation
df_cpi_weights = calculate_cpi_weights(df_cpi_ready)
# Define the output directory and ensure it exists
base_output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
 ⇔cpi_weights"
final_output_dir = os.path.join(base_output_dir, "exio3_figaro_cpi_weights")
→doesn't exist
# Define file path and save
output_file = os.path.join(final_output_dir, "cpi_weights_figaro_2021.csv")
df_cpi_weights.to_csv(output_file)
print(f" CPI weights saved to {output_file}")
```

```
[]: | ### Create EU28 household demand column based only on EU28 countries ###
     # Step 1: Filter for EU28 countries and 'P3_S14' only
     eu28_p3_s14_columns = [(country, 'P3_S14') for country in eu28_countries]
     df_p3_s14_eu28 = df_cpi_ready.loc[:, eu28_p3_s14_columns]
     # Step 2: Sum across these EU28 household demand columns
     df_cpi_ready_eu28 = df_p3_s14_eu28.sum(axis=1).to_frame(name='EU28')
     # Step 3: Calculate CPI weights using the existing function
     df_cpi_weights_eu28 = calculate_cpi_weights(df_cpi_ready_eu28)
     # Save result
     output_file_eu28 = os.path.join(final_output_dir, "cpi_weights_figaro_2021_EU28.
     df_cpi_weights_eu28.to_csv(output_file_eu28)
     print(f" CPI weights for EU28 saved to {output_file_eu28}")
[]: ### Technical Coefficient Matrix Calculation ###
     # ---- STEP 0: Extract the gross output row (since it's unique, with index
     → ("GO", "GO"))
     go_row = df_2021_rdy.loc[("GO", "GO")]
     # ---- STEP 1: Build a dictionary for gross output for each supplier column
     gross_output = {}
     for col in merged df.columns:
         country, supplier_sector = col
         if supplier sector != "GO": # Only for supplier columns
            try:
                 go_value = go_row[col]
                gross_output[col] = go_value
             except KeyError:
                 print(f"Warning: Gross output for supplier column {col} not found; ⊔

defaulting to 1.")
                 gross_output[col] = 1
     # ---- STEP 2: Create the interindustry block by excluding rows and columns
      ⇔where Sector == "GO"
     rows_mask = merged_df.index.get_level_values("Sector") != "GO"
     cols_mask = merged_df.columns.get_level_values("Sector") != "GO"
     interindustry_block = merged_df.loc[rows_mask, cols_mask]
     # ---- STEP 3: Compute the technical coefficients
     def normalize_column(col):
        supplier_key = col.name # a tuple (country, supplier_sector)
```

```
denominator = gross_output.get(supplier_key, 1)
   return col / denominator
tech_coeff = interindustry_block.apply(normalize_column, axis=0)
# ---- STEP 4: Duplicate B rows with new sector names "B_gas" and "B_nongas"
b_rows = tech_coeff.loc[tech_coeff.index.get_level_values("Sector") == "B"].
 ⇔copy()
# Function to update MultiIndex for row transformations
def update_sector(index, new_label):
   new_tuples = [(country, new_label if sector == "B" else sector) for⊔
 ⇔country, sector in index]
   return pd.MultiIndex.from_tuples(new_tuples, names=index.names)
# Duplicate rows
b_rows_gas = b_rows.copy()
b_rows_nongas = b_rows.copy()
b_rows_gas.index = update_sector(b_rows.index, "B_gas")
b_rows_nongas.index = update_sector(b_rows.index, "B_nongas")
# Remove original B rows and add new ones
non_b_rows = tech_coeff.loc[tech_coeff.index.get_level_values("Sector") != "B"]
tech_coeff_rows_modified = pd.concat([non_b_rows, b_rows_gas, b_rows_nongas]).
⇔sort_index()
# ---- STEP 4a (New): Duplicate B columns with new sector names "B gas" and
→"B nongas"
b_columns = tech_coeff_rows_modified.loc[:, tech_coeff_rows_modified.columns.
 # Function to update MultiIndex for column transformations
def update column(index, new label):
   new_tuples = [(country, new_label if sector == "B" else sector) for_
 ⇔country, sector in index]
   return pd.MultiIndex.from_tuples(new_tuples, names=index.names)
# Duplicate columns
b columns gas = b columns.copy()
b_columns_nongas = b_columns.copy()
b_columns_gas.columns = update_column(b_columns.columns, "B_gas")
b_columns_nongas.columns = update_column(b_columns.columns, "B_nongas")
# Remove original B columns and add new ones
non_b_columns = tech_coeff_rows_modified.loc[:, tech_coeff_rows_modified.
 ⇔columns.get_level_values("Sector") != "B"]
```

```
tech_coeff_final = pd.concat([non_b_columns, b_columns_gas, b_columns_nongas],_
      ⇒axis=1).sort_index(axis=1)
     # ---- STEP 6: Save the final modified table to CSV
    tech_coeff_final.to_csv("C:/Users/danie/Nextcloud/Coding/Masterthesis/notebooks/
      →NB exio3 figaro gas/technical coefficients modified.csv")
    print("Modified technical coefficients table saved to__
      display(tech coeff final)
[]: # Assume figaro B is the Figaro technical coefficients table (B matrix) with a
     \hookrightarrow MultiIndex
    figaro_B = tech_coeff_final
    weights_B = exio_Z_final
[]: # Define aggregated sector mapping
    aggregated_sector = "C25-33"
    disaggregated_sectors = ['C25', 'C26', 'C27', 'C28', 'C29', 'C30', 'C31_32', __
     # Ensure MultiIndexes are correctly assigned
    figaro_B.index.names = ['Country', 'Sector']
    figaro_B.columns.names = ['Country', 'Sector']
    weights_B.index.names = ['Country', 'Sector']
    weights_B.columns.names = ['Country', 'Sector']
    display(figaro B.index)
    display(weights_B.index)
[]: | ### Step 1: Apply row-wise weightings (B_gas and B_nongas rows)
    for country in figaro_B.index.get_level_values(0).unique():
        for row in ['B_gas', 'B_nongas']:
            if (country, row) in figaro_B.index and (country, row) in weights_B.
      ⇒index:
                for target_country, sector in figaro_B.columns:
                    # Normal case: Use direct weight if sector exists in weights B
                    if (target_country, sector) in weights_B.columns:
                        weight = weights_B.loc[(country, row), (target_country,__
      ⇔sector)]
                        figaro_B.loc[(country, row), (target_country, sector)] *=__
      ⇔weight
                    # Special handling for disaggregated C25-33 sectors
```

```
⇒aggregated_sector) in weights_B.columns:
                       weight = weights_B.loc[(country, row), (target_country,__
     →aggregated sector)]
                       figaro_B.loc[(country, row), (target_country, sector)] *=__
     ⇔weight
                       print(f"Applied ROW weight to disaggregated sector:
     →row), (target_country, sector)]}") # Debugging
    ### Step 2: Apply column-wise weightings (only to B_gas and B_nongas columns)
    for country in figaro_B.columns.get_level_values(0).unique():
        for col in ['B_gas', 'B_nongas']:
            if (country, col) in figaro_B.columns and (country, col) in weights_B.
     ⇔columns:
               for target_country, row in figaro_B.index:
                   if row not in ['B_gas', 'B_nongas']: # Skip these rows
                       # Normal case: Use direct weight if sector exists in_
     \hookrightarrow weights_B
                       if (target_country, row) in weights_B.index:
                          weight = weights_B.loc[(target_country, row), (country,__
     ⇔col)]
                          figaro B.loc[(target country, row), (country, col)] *=___
     ⇔weight
                       # Special handling for disaggregated C25-33 sectors
                       elif row in disaggregated_sectors and (target_country, __
     →aggregated_sector) in weights_B.index:
                          weight = weights_B.loc[(target_country,__
     →aggregated_sector), (country, col)]
                          figaro_B.loc[(target_country, row), (country, col)] *=__
     ⇔weight
                          print(f"Applied COLUMN weight to disaggregated sector:
     →row), (country, col)]}") # Debugging
    # Display results
    display(figaro_B)
[]: # Define the final demand sectors to remove
    final_demand_sectors = ['P3_S13', 'P3_S14', 'P3_S15', 'P51G', 'P5M']
    # Step 1: Drop all columns where the sector is in final demand sectors
    figaro_df = figaro_B.drop(columns=[col for col in figaro_B.columns if col[1] in_
     →final_demand_sectors])
```

elif sector in disaggregated_sectors and (target_country, __

```
# Step 2: Drop all rows where 'Country' is 'W2'
figaro_df = figaro_df.drop(index='W2', level=0) # Drops rows where first index______
level (Country) is 'W2'

# Step 3: Replace all NaN values with 0
figaro_df = figaro_df.fillna(0)

# Display the cleaned DataFrame
display(figaro_df)
```

```
[]: ### 500% shock to imported gas into EU28 from outside EU28 ###
     # Define output directory
     output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
     ⇔results"
     os.makedirs(output_dir, exist_ok=True) # Ensure directory exists
     # Define the shock factor (500% increase)
     shock_factor = 6.0 # 500% increase
     no_shock_factor = 1.0 # No change for EU countries
     # Define energy and non-energy sectors
     energy_sectors = ['B_gas']
     non_energy_sectors = [s for s in figaro_df.index.get_level_values(1).unique()_u
     →if s not in energy_sectors]
     # Ensure MultiIndex is correctly set
     figaro_df.index = figaro_df.index.set_names(['Country', 'Sector'])
     figaro_df.columns = figaro_df.columns.set_names(['Country', 'Sector'])
     # Placeholder for results
     shock_impacts = []
     # **Apply shock for EU28 collectively (imported gas from non-EU28)**
     print("Processing imported gas shock into EU28")
     # Define block partitioning for energy vs. non-energy (within EU28)
     N_indices = [(c, sec) for c, sec in figaro_df.index if c in eu28_countries and_
      →sec in non_energy_sectors]
```

```
E indices = [(c, sec) for c, sec in figaro_df.index if sec in energy_sectors] ___
 →# Include ALL countries
# Partition the IO coefficients (EU28)
A_EE = figaro_df.loc[N_indices, N_indices].values # endogenous (non-energy) to_{\square}
⇔endogenous (non-energy)
A_XE = figaro_df.loc[E_indices, N_indices].values # endogenous (non-energy)_
⇔consuming exogenous (energy)
# Ensure A_XE is correctly shaped: (N \times E)
A_XE_T = A_XE.T # Now (E \times N), ensuring proper multiplication
# Identity matrix for energy block
I_EE = np.eye(A_EE.shape[0])
# Compute the Leontief inverse for EU28's energy sectors
try:
    L_{EE} = np.linalg.inv(I_{EE} - A_{EE}.T) # (I - A'_{EE})^-1
except np.linalg.LinAlgError:
    print("Singular matrix encountered for EU28. Skipping...")
    exit()
# Start with a default price vector (all set to 1)
P_X = np.ones((A_XE.shape[0], 1)) # Each row corresponds to an energy sector_
 ⇒in each country
# Iterate over all energy sectors (suppliers)
for i, (supplier_country, sec) in enumerate(E_indices):
    if sec == 'B_gas': # Only shock B_gas
        for j, (buyer_country, buyer_sec) in enumerate(N_indices): # Iterate_
 ⇔over non-energy buyers
            if buyer_country in eu28_countries and supplier_country not in_
 ⇔eu28 countries:
                # If gas is being delivered from a non-EU country to an EU28_
 →country, apply the shock
                P X[i, 0] = shock factor # Set price to 5 for affected gas,
 \hookrightarrow imports
# Compute price changes for EU28's energy sectors
delta_P_E = L_EE @ (A_XE_T @ P_X)
# Store results in a properly structured DataFrame
```

```
[]: ### 500% shock to ALL imported gas ###
     # Define output directory
     output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      ⇔results"
     os.makedirs(output_dir, exist_ok=True) # Ensure directory exists
     # Define the shock factor (500% increase)
     shock_factor = 6.0 # 500% increase
     no_shock_factor = 1.0 # No change for EU countries
     # Define energy and non-energy sectors
     energy_sectors = ['B_gas']
     non_energy_sectors = [s for s in figaro_df.index.get_level_values(1).unique()__
      →if s not in energy_sectors]
     # Ensure MultiIndex is correctly set
     figaro_df.index = figaro_df.index.set_names(['Country', 'Sector'])
     figaro_df.columns = figaro_df.columns.set_names(['Country', 'Sector'])
     # Placeholder for results
     shock_impacts = []
     # **Apply shock for EU28 collectively (imported gas from non-EU28)**
     print("Processing imported gas shock into EU28")
     # Define block partitioning for energy vs. non-energy (within EU28)
     N_indices = [(c, sec) for c, sec in figaro_df.index if c in eu28_countries and_
      ⇒sec in non_energy_sectors]
     E_indices = [(c, sec) for c, sec in figaro_df.index if sec in energy_sectors] __
     →# Include ALL countries
     # Partition the IO coefficients (EU28)
     A_EE = figaro_df.loc[N_indices, N_indices].values # endogenous (non-energy) to_{\square}
     ⇔endogenous (non-energy)
```

```
A XE = figaro_df.loc[E_indices, N_indices].values # endogenous (non-energy)_
⇔consuming exogenous (energy)
# Ensure A XE is correctly shaped: (N \times E)
A_XE_T = A_XE.T # Now (E \times N), ensuring proper multiplication
# Identity matrix for energy block
I EE = np.eye(A EE.shape[0])
# Compute the Leontief inverse for EU28's energy sectors
try:
    L_{EE} = np.linalg.inv(I_{EE} - A_{EE}.T) # (I - A'_{EE})^{-1}
except np.linalg.LinAlgError:
    print("Singular matrix encountered for EU28. Skipping...")
    exit()
# Start with 1s (no shock)
P_X = np.ones((A_XE.shape[0], 1))
# Apply shocks to extra-EU imports (like in Scenario 1)
for i, (supplier country, sec) in enumerate(E indices):
    if sec == 'B_gas' and supplier_country not in eu28_countries:
        P_X[i, 0] = shock_factor # Shock extra-EU gas imports (5.0)
# Apply shocks to intra-EU imports (Scenario 2 addition)
for i, (buyer_country, sec) in enumerate(E_indices):
    if sec == 'B_gas' and buyer_country in eu28_countries:
        for j, (supplier_country, sec2) in enumerate(E_indices):
            if sec2 == 'B_gas' and supplier_country in eu28_countries and_
 ⇒supplier_country != buyer_country:
                P_X[j, 0] = shock_factor # Shock intra-EU gas imports (5.0)
# Compute price changes for EU28's energy sectors
delta P E = L EE @ (A XE T @ P X)
# Store results in a properly structured DataFrame
shock_results_intra_extra_df = pd.DataFrame(delta P E.flatten(), index=pd.
MultiIndex.from_tuples(N_indices, names=["Country", "Sector"]),
⇔columns=["Price Change"])
# Save results in three-column format
output_file = os.path.join(output_dir, 'imported_gas_shock_intra_extra_EU28.
shock_results_intra_extra_df.to_csv(output_file, index=True) # Keeps11
 \hookrightarrow MultiIndex
```

```
print(f"Shock impacts saved to {output_file}")
[]: print(shock_results_extra_df.index)
     print(shock_results_extra_df.columns)
     print(shock_results_intra_extra_df.index)
     print(shock_results_intra_extra_df.columns)
     print(df_cpi_weights.index)
     print(df_cpi_weights.columns)
[]: display(df_cpi_weights)
     display(shock_results_extra_df)
[]: # Define the output directory
     output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
     →results/cpi_weighted_shocks"
     os.makedirs(output_dir, exist_ok=True) # Ensure the directory exists
     # Define file paths for the CPI-weighted results
     file_cpi_scenario_1 = os.path.join(output_dir,__

¬"cpi_weighted_imported_gas_shock_extra_EU28.csv")
     file_cpi_scenario_2 = os.path.join(output_dir,__

¬"cpi_weighted_imported_gas_shock_intra_extra_EU28.csv")

     # Define weights and both shock scenarios
     weights = df_cpi_weights
     shock_scenarios = {
         file_cpi_scenario_1: shock_results_extra_df,
         file_cpi_scenario_2: shock_results_intra_extra_df
     }
     # Drop the unnecessary second level in the columns of weights (P3 S14)
     if isinstance(weights.columns, pd.MultiIndex) and len(weights.columns.levels) > __
         weights = weights.droplevel(level=1, axis=1) # Ensure we only have
     → 'Country' as columns
     # Get the list of consuming countries
     consuming countries = weights.columns # Now only country names remain as_{\sqcup}
     ⇔column labels
     # Loop over both scenarios
     for file_path, shock_results in shock_scenarios.items():
         # Initialize an empty DataFrame with the MultiIndex from shock results
         cpi_weighted_results = pd.DataFrame(index=shock_results.index)
         # Loop through each consuming country
```

```
for consuming_country in consuming_countries:
             # Select the CPI weights for this consuming country
             country weights = weights[consuming country] # This should now be a_
      \hookrightarrowSeries
             # Broadcast these weights to the full (Country, Sector) index of []
      ⇔shock results
             aligned_weights = country_weights.reindex(shock_results.index,_

fill_value=0)
             # Multiply price changes by weights
             cpi_weighted_values = shock_results['Price Change'] * aligned_weights
             # Ensure we store a **flat** 1D Series (not a DataFrame)
             cpi_weighted_results[consuming_country] = cpi_weighted_values.squeeze()_
      → # Convert to 1D Series
         # Save results
         cpi weighted results.to csv(file path)
         print(f"CPI-weighted shocks saved to {file_path}")
[]: # Load EU28 CPI weights
     cpi_weights_eu28_path = os.path.join(final_output_dir,_

¬"cpi_weights_figaro_2021_EU28.csv")
     weights_eu28 = pd.read_csv(cpi_weights_eu28_path, index_col=[0, 1])['EU28']
     # Define output file paths
     file_cpi_scenario_1_eu28 = os.path.join(output_dir,_
     →"cpi_weighted_imported_gas_shock_extra_EU28_eu28weighted.csv")
     file_cpi_scenario_2_eu28 = os.path.join(output_dir,_

¬"cpi_weighted_imported_gas_shock_intra_extra_EU28_eu28weighted.csv")

     # Define mapping of files and shocks
     shock_scenarios_eu28 = {
         file_cpi_scenario_1_eu28: shock_results_extra_df,
         file_cpi_scenario_2_eu28: shock_results_intra_extra_df
     }
     # Loop through each scenario and apply EU28 weights
     for file_path, shock_results in shock_scenarios_eu28.items():
         # Align the EU28 weights to the index of the shock result
         aligned_weights = weights_eu28.reindex(shock_results.index, fill_value=0)
```

Apply the EU28 weights to the price change column

```
[]: | ### Visualization of CPI-weighted Results in a stacked bar chart ###
     # Define file paths for the CPI-weighted results
     file_cpi_scenario_1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

¬processed/results/cpi_weighted_shocks/

¬cpi_weighted_imported_gas_shock_extra_EU28.csv"

     file_cpi_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

¬processed/results/cpi_weighted_shocks/
      ⇒cpi_weighted_imported_gas_shock_intra_extra_EU28.csv"
     # Define base output directory
     base_output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
     figures_dir_cpi = os.path.join(base_output_dir,__

¬"figures_cpi_weighted_gas_shocks")
     os.makedirs(figures_dir_cpi, exist_ok=True) # Ensure folder exists
     # Read the CSV files into DataFrames (CPI-weighted results)
     df_cpi_scenario_1 = pd.read_csv(file_cpi_scenario_1, index_col=[0, 1])
     df_cpi_scenario_2 = pd.read_csv(file_cpi_scenario_2, index_col=[0, 1])
     # Multiply by 100 to convert to percentage values
     df_cpi_scenario_1 *= 100
     df_cpi_scenario_2 *= 100
     # Define EU28 country codes
     eu28 countries = [
         "AT", "BE", "BG", "CY", "CZ", "DE", "DK", "EE", "ES", "FI", "FR", "GB", |

¬"GR", "HR", "HU",

         "IE", "IT", "LT", "LU", "LV", "MT", "NL", "PL", "PT", "RO", "SE", "SI", "SK"
     ]
     # Function to determine the y-axis limit based on max impact
     def get_dynamic_y_limit(max_shock):
         step_size = 0.05 # Define step size for y-axis scaling
         return np.ceil(max_shock / step_size) * step_size # Round up to the next_
      \hookrightarrowstep
     # Iterate over each EU28 country to generate CPI-weighted plots
     for country in eu28_countries:
         try:
```

```
# Extract the CPI-weighted impact for this country
      country_cpi_scenario_1 = df_cpi_scenario_1[country].
⇔sort_values(ascending=False)
      country_cpi_scenario_2 = df_cpi_scenario_2[country].
⇒sort_values(ascending=False)
      # Select the top 20 affected sectors
      top_20_country_cpi_2 = country_cpi_scenario_2.head(20) # Scenario_2_
⇔(full intra + extra shock)
      # Convert MultiIndex (Country, Sector) into readable strings
      top_20_labels = [f''[idx[0]] - [idx[1]]'' for idx in top_20_country_cpi_2.
⊶indexl
      # Calculate `max_shock_2` inside the loop, after extracting top sectors
      max_shock_2 = top_20_country_cpi_2.max()
      y_axis_limit = get_dynamic_y_limit(max_shock_2)
      ### **Bar Chart for Scenario 2 (Intra- & Extra-EU Imports Shock)**
      plt.figure(figsize=(12, 6))
      plt.bar(top_20_labels, top_20_country_cpi_2, color='darkorange')
      plt.title(f"Top 20 CPI-Weighted Sectors in {country} (Intra- & Extra-EU⊔
Gas Shock)")
      plt.ylabel("CPI-Weighted Price Impact (%)")
      plt.xlabel("Sectors")
      plt.xticks(rotation=45, ha="right")
      plt.ylim(0, y_axis_limit)
      plt.grid(axis="y", linestyle="--", alpha=0.7)
      plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:
plt.savefig(os.path.join(figures_dir_cpi,__
→f"Top_20_CPI_Sectors_{country}.png"), bbox_inches="tight")
      plt.close()
       ### **Stacked Bar Chart for Intra vs. Extra EU Impact**
      diff cpi scenario = country cpi scenario 2.loc[top 20 country cpi 2.
-index] - country_cpi_scenario_1.loc[top_20_country_cpi_2.index]
      plt.figure(figsize=(12, 6))
      plt.bar(
          top_20_labels,
          country_cpi_scenario_1.loc[top_20_country_cpi_2.index],
          label="Extra-EU Imports Shock",
          color="royalblue"
      plt.bar(
```

```
top_20_labels,
                 diff_cpi_scenario,
                 bottom=country_cpi_scenario_1.loc[top_20_country_cpi_2.index],
                 label="Intra-EU Impact",
                 color="darkorange"
             )
             # Format plot
             plt.title(f"CPI-Weighted Impact in {country}: Gas Shock (Stacked)")
             plt.ylabel("CPI-Weighted Price Impact (%)")
             plt.xlabel("Sectors")
             plt.xticks(rotation=45, ha="right")
             plt.ylim(0, y_axis_limit) # Set scaling
             plt.grid(axis="y", linestyle="--", alpha=0.7)
             plt.legend()
             plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:
      # Save figure
             plt.savefig(os.path.join(figures_dir_cpi,__
      of "Stacked_CPI_Sectors_{country}.png"), bbox_inches="tight")
             plt.close()
             print(f" Saved CPI-weighted visualizations for {country} in _{\!\sqcup}

√{figures_dir_cpi}")

         except KeyError:
             print(f" Skipping {country}: No CPI-weighted data available.")
[]: # Choose scenario (e.g., intra + extra EU shock)
     df_cpi_weighted = df_cpi_scenario_2 *100 # Or df_cpi_scenario_1 for only_
      \hookrightarrow Extra-EU
     # Step 1: Filter to EU28 countries only (columns)
     df_cpi_weighted_eu28 = df_cpi_weighted[eu28_countries]
     # Step 2: Compute average CPI-weighted impact for each sector across all EU28_
     # Group by the 'Sector' level of the index (level=1)
     sector_avg impact = df_cpi_weighted_eu28.groupby(level="Sector").mean().
      →mean(axis=1)
     # Step 3: Sort and select top sectors
     top_20_sectors = sector_avg_impact.sort_values(ascending=False).head(20)
     # Step 4: Plot
     plt.figure(figsize=(12, 6))
```

```
[]: # === Add EU28 CPI-weighted stacked bar chart ===
     # File paths for EU28-weighted results
     file_cpi_eu28_extra = os.path.join(output_dir,__

¬"cpi_weighted_imported_gas_shock_extra_EU28_eu28weighted.csv")

     file_cpi_eu28_intra_extra = os.path.join(output_dir,__
      -"cpi_weighted_imported_gas_shock_intra_extra_EU28_eu28weighted.csv")
     # Load and convert to percentage
     df_eu28_extra = pd.read_csv(file_cpi_eu28_extra, index_col=[0, 1]) * 100
     df_eu28_intra_extra = pd.read_csv(file_cpi_eu28_intra_extra, index_col=[0, 1])_u
      →* 100
     # Sort sectors by descending impact (Scenario 2 = intra + extra)
     top_20_eu28 = df_eu28_intra_extra['EU28'].sort_values(ascending=False).head(40)
     # Generate readable labels
     top_20\_labels = [f''\{idx[0]\} - \{idx[1]\}'' for idx in top_20\_eu28.index]
     # Determine y-axis limit
     max_shock_eu28 = top_20_eu28.max()
     y_axis_limit = get_dynamic_y_limit(max_shock_eu28)
     # Compute difference = intra-EU contribution
     diff_eu28 = df_eu28_intra_extra.loc[top_20_eu28.index, 'EU28'] - df_eu28_extra.
      ⇒loc[top_20_eu28.index, 'EU28']
     # === Stacked Bar Chart for EU28 ===
     plt.figure(figsize=(12, 6))
     plt.bar(
         top 20 labels,
         df_eu28_extra.loc[top_20_eu28.index, 'EU28'],
         label="Extra-EU Imports Shock",
         color="royalblue"
     plt.bar(
```

```
top_20_labels,
         diff eu28,
         bottom=df_eu28_extra.loc[top_20_eu28.index, 'EU28'],
         label="Intra-EU Impact",
         color="darkorange"
     )
     # Format plot
     plt.title("EU28 CPI-Weighted Impact: Gas Shock (Stacked)")
     plt.ylabel("CPI-Weighted Price Impact (%)")
     plt.xlabel("Sectors")
     plt.xticks(rotation=45, ha="right")
     plt.ylim(0, y_axis_limit)
    plt.grid(axis="y", linestyle="--", alpha=0.7)
     plt.legend()
    plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:.2f}%"))
     # Save plot
     plt.savefig(os.path.join(figures_dir_cpi, "Stacked_CPI_Sectors_EU28_t40.png"), u
      ⇔bbox_inches="tight")
     plt.close()
     print(" Saved CPI-weighted stacked bar chart for EU28 in", figures_dir_cpi)
[]: | # Load EU28-weighted CPI shock results
     file_cpi_eu28_extra = os.path.join(output_dir,__

¬"cpi_weighted_imported_gas_shock_extra_EU28_eu28weighted.csv")

     file cpi eu28 intra extra = os.path.join(output dir,
      -"cpi_weighted_imported_gas_shock_intra_extra_EU28_eu28weighted.csv")
     df_eu28_extra = pd.read_csv(file_cpi_eu28_extra, index_col=[0, 1]) * 100
     df_eu28_intra_extra = pd.read_csv(file_cpi_eu28_intra_extra, index_col=[0, 1])__
      →* 100
     # Step 1: Accumulate sectoral impacts by **producing country** (first level of
     impact_extra_by_origin = df_eu28_extra.groupby(level=0)['EU28'].sum()
     impact_intra_extra_by_origin = df_eu28_intra_extra.groupby(level=0)['EU28'].
      ⇒sum()
```

Step 2: Compute the **intra-EU contribution** per origin country

impact_intra_only_by_origin = impact_intra_extra_by_origin -__

→impact_extra_by_origin

Step 3: Sort by total impact (descending)

```
sorted_countries = impact_intra_extra_by_origin.sort_values(ascending=False).
      ⊶index
     # Step 4: Plot stacked bar chart
     plt.figure(figsize=(14, 7))
     plt.bar(
         sorted countries,
         impact extra by origin.loc[sorted countries],
         label="Extra-EU Shock Impact",
         color="royalblue"
     plt.bar(
         sorted_countries,
         impact_intra_only_by_origin.loc[sorted_countries],
         bottom=impact_extra_by_origin.loc[sorted_countries],
         label="Intra-EU Impact",
         color="darkorange"
     )
     # Format plot
     plt.title("Origin Countries' Contribution to EU28 CPI-Weighted Inflation,
      plt.ylabel("Total CPI-Weighted Inflation Impact (%)")
     plt.xlabel("Origin Country")
     plt.xticks(rotation=45, ha="right")
     plt.grid(axis="y", linestyle="--", alpha=0.7)
     plt.legend()
     plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:.2f}%"))
     # Save figure
     output_path = os.path.join(figures_dir_cpi,__

¬"Country_Origin_Impact_on_EU28_Inflation_Stacked.png")

     plt.savefig(output_path, bbox_inches="tight")
     plt.close()
     print(" Saved origin country stacked CPI-weighted impact chart to:", ___
      →output_path)
[]: | ### Cumulative Inflation Impact by EU28-Country (Stacked) ###
     # Define file paths for the CPI-weighted results
     file_cpi_scenario_1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

→processed/results/cpi_weighted_shocks/

¬cpi_weighted_imported_gas_shock_extra_EU28.csv"
```

```
file_cpi_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/
 ⇒processed/results/cpi_weighted_shocks/

¬cpi_weighted_imported_gas_shock_intra_extra_EU28.csv"

# Define output directory for charts
base output dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
figures_dir_cpi = os.path.join(base_output_dir,_

¬"figures_cpi_weighted_gas_shocks")
os.makedirs(figures_dir_cpi, exist_ok=True) # Ensure folder exists
# Read CPI-weighted results into DataFrames
df cpi scenario 1 = pd.read csv(file cpi scenario 1, index col=[0, 1])
df_cpi_scenario_2 = pd.read_csv(file_cpi_scenario_2, index_col=[0, 1])
# Convert values to percentage
df_cpi_scenario_1 *= 100
df_cpi_scenario_2 *= 100
# Filter columns to only include EU28 countries
df_cpi_scenario_1 = df_cpi_scenario_1[eu28_countries]
df_cpi_scenario_2 = df_cpi_scenario_2[eu28_countries]
# Sum total inflation impact across all sectors for each EU28 country
cumulative_impact_scenario_1 = df_cpi_scenario_1.sum(axis=0) # Extra-EU Gas_u
 \hookrightarrow Shock
cumulative_impact_scenario_2 = df_cpi_scenario_2.sum(axis=0) # Intra+Extra-EU_
 →Gas Shock
# Calculate difference (intra-EU contribution)
cumulative_intra_impact = cumulative_impact_scenario_2 -_
 →cumulative_impact_scenario_1
# Sort countries by total impact (Scenario 2)
sorted_countries = cumulative_impact_scenario_2.sort_values(ascending=False).
 ⊶index
# Generate stacked bar chart
plt.figure(figsize=(14, 7))
plt.bar(sorted_countries, cumulative_impact_scenario_1.loc[sorted_countries],_
 ⇔label="Extra-EU Gas Shock", color="royalblue")
plt.bar(sorted_countries, cumulative_intra_impact.loc[sorted_countries],
        bottom=cumulative_impact_scenario_1.loc[sorted_countries],
        label="Intra-EU Impact", color="darkorange")
# Format plot
plt.title("Cumulative CPI-Weighted Inflation Impact by Country (Stacked)")
```

```
plt.ylabel("Total CPI-Weighted Inflation Impact (%)")
plt.xlabel("Country")
plt.xticks(rotation=45, ha="right")
plt.grid(axis="y", linestyle="--", alpha=0.7)
plt.legend()
plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:.2f}%"))

# Save figure
output_path = os.path.join(figures_dir_cpi, "Cumulative_CPI_Impact_Stacked.png")
plt.savefig(output_path, bbox_inches="tight")
plt.close()

print(f" Saved cumulative CPI-weighted stacked bar chart: {output_path}")
```

```
[]: ### Cumulative Inflation Impact by Country (Stacked) also outside EUL
     ⇔countries###
     # Define file paths for the CPI-weighted results
     file_cpi_scenario_1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

¬processed/results/cpi_weighted_shocks/
      ⇔cpi_weighted_imported_gas_shock_extra_EU28.csv"
     file_cpi_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

¬processed/results/cpi_weighted_shocks/

¬cpi_weighted_imported_gas_shock_intra_extra_EU28.csv"

     # Define output directory for charts
     base_output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
     figures_dir_cpi = os.path.join(base_output_dir,__

¬"figures_cpi_weighted_gas_shocks")
     os.makedirs(figures_dir_cpi, exist_ok=True) # Ensure folder exists
     # Read CPI-weighted results into DataFrames
     df_cpi_scenario_1 = pd.read_csv(file_cpi_scenario_1, index_col=[0, 1])
     df_cpi_scenario_2 = pd.read_csv(file_cpi_scenario_2, index_col=[0, 1])
     # Convert values to percentage
     df_cpi_scenario_1 *= 100
     df_cpi_scenario_2 *= 100
     # Sum total inflation impact across all sectors for each country
     cumulative_impact_scenario_1 = df_cpi_scenario_1.sum(axis=0) # Extra-EU Gas_
      \hookrightarrow Shock
     cumulative_impact_scenario_2 = df_cpi_scenario_2.sum(axis=0) # Intra+Extra-EU_
      →Gas Shock
```

```
cumulative_intra_impact = cumulative_impact_scenario_2 -__
      ⇒cumulative_impact_scenario_1
     # Sort countries by total impact
     sorted_countries = cumulative_impact_scenario_2.sort_values(ascending=False).
      ⇔index
     # Generate stacked bar chart
     plt.figure(figsize=(14, 7))
     plt.bar(sorted_countries, cumulative_impact_scenario_1.loc[sorted_countries],_
      ⇔label="Extra-EU Gas Shock", color="royalblue")
     plt.bar(sorted_countries, cumulative_intra_impact.loc[sorted_countries],
             bottom=cumulative_impact_scenario_1.loc[sorted_countries],
             label="Intra-EU Impact", color="darkorange")
     # Format plot
     plt.title("Cumulative CPI-Weighted Inflation Impact by Country (Stacked)")
     plt.ylabel("Total CPI-Weighted Inflation Impact (%)")
     plt.xlabel("Country")
     plt.xticks(rotation=45, ha="right")
     plt.grid(axis="y", linestyle="--", alpha=0.7)
     plt.legend()
     plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:.2f}%"))
     # Save figure
     output_path = os.path.join(figures_dir_cpi, "Cumulative_CPI_Impact_Stacked.png")
     plt.savefig(output_path, bbox_inches="tight")
     plt.close()
    print(f" Saved cumulative CPI-weighted stacked bar chart: {output_path}")
[]: ### Overall Impact Visualization ###
     # Define file path for the CPI-weighted results
     file_cpi_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/

¬processed/results/cpi_weighted_shocks/
     ⇔cpi_weighted_imported_gas_shock_intra_extra_EU28.csv"
     # Define output directory for visualization
     base_output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
     figures_dir_cpi = os.path.join(base_output_dir,__

¬"figures_cpi_weighted_gas_shocks")
     os.makedirs(figures_dir_cpi, exist_ok=True) # Ensure folder exists
     # Read the CPI-weighted results into a DataFrame
```

Calculate difference (intra-EU contribution)

```
df_cpi_scenario_2 = pd.read_csv(file_cpi_scenario_2, index_col=[0, 1])
     # Multiply by 10,000 to get percentage values
     df_cpi_scenario_2 *= 100
     # Compute the total CPI-weighted impact per country (sum of all sectoral
      ⇔effects)
     total_cpi_impact = df_cpi_scenario_2.sum()
     # Sort countries by total impact
     total_cpi_impact_sorted = total_cpi_impact.sort_values(ascending=False)
     # Determine dynamic y-axis limit starting from 0.5, increasing in 0.5 steps
     y_axis_limit = np.ceil(total_cpi_impact_sorted.max() / 0.5) * 0.5
     # Create bar chart
     plt.figure(figsize=(12, 6))
     total_cpi_impact_sorted.plot(kind="bar", color="darkblue")
     plt.title("Total CPI-Weighted Impact per Country (Intra- & Extra-EU Gas Shock)")
     plt.ylabel("Summed CPI-Weighted Price Impact (%)")
     plt.xlabel("Country")
     plt.xticks(rotation=45, ha="right")
     plt.ylim(0, y_axis_limit)
     plt.grid(axis="y", linestyle="--", alpha=0.7)
     plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:.1f}%"))
     # Save figure
     plot_path = os.path.join(figures_dir_cpi, "Total_CPI_Impact_Per_Country.png")
     plt.savefig(plot_path, bbox_inches="tight")
     plt.close()
     print(f" Saved CPI-weighted impact plot to: {plot_path}")
[]: ### Visualization ###
     ### Unweighted shock impacts ###
     # Load the computed shock results
     output_file = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
     ⇔results/imported_gas_shock_intra_extra_EU28.csv"
```

shock_results_df = pd.read_csv(output_file, index_col=["Country", "Sector"])

Read the CSV file into a DataFrame with MultiIndex

Sort sectors by the absolute price change (largest impacts)

```
shock_results_sorted = shock_results_df.abs().sort_values(by="Price Change",_
 ⇔ascending=False)
# Select the top 20 sectors with the highest price impact
top_20_sectors = shock_results_sorted.head(20)
# Plot the results
plt.figure(figsize=(12, 6))
top_20_sectors.plot(kind="bar", legend=False, title="Top 20 Sectors by Price_∪
 →Impact (500% Gas Shock to imports into EU28)")
plt.ylabel("Relative Price Change")
plt.xlabel("Sectors")
plt.xticks(rotation=45, ha="right")
plt.grid(axis="y", linestyle="--", alpha=0.7)
# Show the plot
plt.show()
import pandas as pd
import matplotlib.pyplot as plt
\# Load the computed shock results while preserving MultiIndex
output_file = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
 →results/imported_gas_shock_into_EU28_fixed.csv"
# Read CSV and restore MultiIndex
shock_results_df = pd.read_csv(output_file, index_col=["Country", "Sector"])
# Convert to percentage changes (already stored in % but ensuring consistency)
shock_results_df["Price Change"] *= 100
# Normalize by taking the mean absolute price impact per sector across all \sqcup
 \hookrightarrow countries
sector_impact_normalized = shock_results_df.groupby(level="Sector")["Price__
# Sort sectors by the most affected
top_20_sectors = sector_impact_normalized.sort_values(ascending=False).head(20)
# Plot the results
plt.figure(figsize=(12, 6))
top_20_sectors.plot(kind="bar", legend=False, title="Top 20 Most Affected_
 ⇔Sectors (Normalized)")
plt.ylabel("Average Price Impact (%)") # Normalized
plt.xlabel("Sectors")
plt.xticks(rotation=45, ha="right")
```

```
plt.grid(axis="y", linestyle="--", alpha=0.7)
# Show the plot
plt.show()
```

```
[]: ### Visualization loop over all countries ###
     # Define file paths for the first two scenarios
     file scenario 1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      →results/imported_gas_shock_extra_EU28.csv"
     file scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      →results/imported_gas_shock_intra_extra_EU28.csv"
     # Read the CSV files into DataFrames with MultiIndex
     df_scenario_1 = pd.read_csv(file_scenario_1, index_col=[0, 1])
     df_scenario_2 = pd.read_csv(file_scenario_2, index_col=[0, 1])
     # Define EU28 country codes
     eu28_countries = [
         "AT", "BE", "BG", "CY", "CZ", "DE", "DK", "EE", "ES", "FI", "FR", "GB", "

→ "GR", "HR", "HU",

         "IE", "IT", "LT", "LU", "LV", "MT", "NL", "PL", "PT", "RO", "SE", "SI", "SK"
     ]
     # Iterate over each EU28 country to generate plots
     for country in eu28_countries:
         try:
             # Filter for the current country
            df_country_scenario_1 = df_scenario_1.loc[country].
      ⇔sort_values(by="Price Change", ascending=False)
             df_country_scenario_2 = df_scenario_2.loc[country].
      ⇔sort_values(by="Price Change", ascending=False)
             # Select the top 20 affected sectors
            top_20_country_scenario_1 = df_country_scenario_1.head(20)
            top_20_country_scenario_2 = df_country_scenario_2.head(20)
             # Plot Scenario 1 for the country
            plt.figure(figsize=(12, 6))
            top_20_country_scenario_1.plot(kind="bar", legend=False, title=f"Top_20_
      →Sectors in {country} (Scenario 1: Extra-EU Gas Shock)")
            plt.ylabel("Relative Price Change (%)")
            plt.xlabel("Sectors")
            plt.xticks(rotation=45, ha="right")
            plt.grid(axis="y", linestyle="--", alpha=0.7)
            plt.show()
```

```
# Plot Scenario 2 for the country
plt.figure(figsize=(12, 6))
top_20_country_scenario_2.plot(kind="bar", legend=False, title=f"Top 20_
Sectors in {country} (Scenario 2: Intra- & Extra-EU Gas Shock)")
plt.ylabel("Relative Price Change (%)")
plt.xlabel("Sectors")
plt.xlabel("Sectors")
plt.xticks(rotation=45, ha="right")
plt.grid(axis="y", linestyle="--", alpha=0.7)
plt.show()

except KeyError:
print(f"Skipping {country}: No data available.")
```

```
[]: import os
     import pandas as pd
     import matplotlib.pyplot as plt
     import numpy as np
     # Define file paths for the first two scenarios
     file_scenario_1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      ⇔results/imported_gas_shock_extra_EU28.csv"
     file_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      →results/imported_gas_shock_intra_extra_EU28.csv"
     # Define base output directory
     base_output_dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
     figures_dir = os.path.join(base_output_dir, "figures_gas_shocks")
     # Ensure the output directory exists
     os.makedirs(figures_dir, exist_ok=True)
     # Read the CSV files into DataFrames with MultiIndex
     df_scenario_1 = pd.read_csv(file_scenario_1, index_col=[0, 1])
     df_scenario_2 = pd.read_csv(file_scenario_2, index_col=[0, 1])
     # Multiply price changes by 100 to get percentages
     df_scenario_1["Price Change"] *= 100
     df_scenario_2["Price Change"] *= 100
     # Define EU28 country codes
     eu28 countries = [
         "AT", "BE", "BG", "CY", "CZ", "DE", "DK", "EE", "ES", "FI", "FR", "GB", [
     →"GR", "HR", "HU",
         "IE", "IT", "LT", "LU", "LV", "MT", "NL", "PL", "PT", "RO", "SE", "SI", "SK"
     ]
     # Function to determine the y-axis limit based on max shock
```

```
def get_y_axis_limit(max_shock):
    if max_shock <= 20:</pre>
        return 20
    elif max_shock <= 30:</pre>
        return 30
    elif max_shock <= 40:</pre>
        return 40
    elif max_shock <= 50:</pre>
        return 50
    else:
        return np.ceil(max_shock / 10) * 10 # Round up to the next multiple of __
 →10
# Iterate over each EU28 country to generate plots
for country in eu28_countries:
    try:
        # Filter for the current country
        df_country_scenario_1 = df_scenario_1.loc[country].
 ⇔sort_values(by="Price Change", ascending=False)
        df_country_scenario_2 = df_scenario_2.loc[country].
 ⇒sort_values(by="Price Change", ascending=False)
        # Select the top 20 affected sectors
        top_20_country_scenario_1 = df_country_scenario_1.head(20)
        top_20_country_scenario_2 = df_country_scenario_2.head(20)
        # Determine the y-axis limit based on max price change in each country
        max_shock_1 = top_20_country_scenario_1["Price Change"].max()
        max_shock_2 = top_20_country_scenario_2["Price Change"].max()
        y_axis_limit = get_y_axis_limit(max(max_shock_1, max_shock_2))
        # Plot and save Scenario 1 for the country
        plt.figure(figsize=(12, 6))
        top_20_country_scenario_1["Price Change"].plot(kind="bar", __
 →legend=False, color='royalblue')
        plt.title(f"Top 20 Sectors in {country} (Scenario 1: Extra-EU Gas⊔

¬Shock)")
        plt.ylabel("Relative Price Change (%)")
        plt.xlabel("Sectors")
        plt.xticks(rotation=45, ha="right")
        plt.ylim(0, y_axis_limit) # Set country-specific scaling
        plt.grid(axis="y", linestyle="--", alpha=0.7)
        plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:
 \hookrightarrow .Of}%")) # Format y-axis as percentage
        plt.savefig(os.path.join(figures_dir,__

¬f"Top_20_Sectors_{country}_Scenario1.png"), bbox_inches="tight")
```

```
plt.close()
        # Plot and save Scenario 2 for the country
        plt.figure(figsize=(12, 6))
        top_20_country_scenario_2["Price Change"].plot(kind="bar", __
 →legend=False, color='darkorange')
        plt.title(f"Top 20 Sectors in {country} (Scenario 2: Intra- & Extra-EU, |
 Gas Shock)")
        plt.ylabel("Relative Price Change (%)")
        plt.xlabel("Sectors")
        plt.xticks(rotation=45, ha="right")
        plt.ylim(0, y axis limit) # Set country-specific scaling
        plt.grid(axis="y", linestyle="--", alpha=0.7)
        plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:
 \hookrightarrow .Of}%")) # Format y-axis as percentage
        plt.savefig(os.path.join(figures_dir,__
 of "Top_20_Sectors_{country}_Scenario2.png"), bbox_inches="tight")
        plt.close()
        print(f"Saved figures for {country} in {figures dir}")
    except KeyError:
        print(f"Skipping {country}: No data available.")
import pandas as pd
```

```
[]: import os
     import matplotlib.pyplot as plt
     import numpy as np
     # Define file paths for the two scenarios
     file_scenario_1 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      →results/imported_gas_shock_extra_EU28.csv"
     file_scenario_2 = "C:/Users/danie/Nextcloud/Coding/Masterthesis/data/processed/
      ⇔results/imported_gas_shock_intra_extra_EU28.csv"
     # Define output directory
     base output dir = "C:/Users/danie/Nextcloud/Coding/Masterthesis/visualization"
     figures_dir = os.path.join(base_output_dir, "figures_gas_shocks_combined")
     os.makedirs(figures_dir, exist_ok=True) # Ensure the folder exists
     # Read the CSV files into DataFrames with MultiIndex
     df_scenario_1 = pd.read_csv(file_scenario_1, index_col=[0, 1])
     df_scenario_2 = pd.read_csv(file_scenario_2, index_col=[0, 1])
     # Convert price changes to percentages
     df_scenario_1["Price Change"] *= 100
     df_scenario_2["Price Change"] *= 100
```

```
# Define EU28 country codes
eu28 countries = [
    "AT", "BE", "BG", "CY", "CZ", "DE", "DK", "EE", "ES", "FI", "FR", "GB", "
 ⇔"GR", "HR", "HU",
    "IE", "IT", "LT", "LU", "LV", "MT", "NL", "PL", "PT", "RO", "SE", "SI". "SK"
]
# Function to determine the y-axis limit based on max shock
def get_y_axis_limit(max_shock):
    if max_shock <= 10:</pre>
        return 10
    elif max_shock <= 20:</pre>
        return 20
    elif max shock <= 30:
        return 30
    elif max shock <= 40:</pre>
        return 40
    elif max shock <= 50:</pre>
        return 50
    else:
        return np.ceil(max_shock / 10) * 10 # Round up to the next multiple of ___
 →10
# Iterate over each EU28 country to generate stacked bar plots
for country in eu28_countries:
    try:
        # Filter data for the current country
        df_country_scenario_1 = df_scenario_1.loc[country]
        df_country_scenario_2 = df_scenario_2.loc[country]
        # Compute total shock (Scenario 2)
        total_impact = df_country_scenario_2["Price Change"]
        # Sort sectors by total impact in descending order
        sorted_indices = total_impact.sort_values(ascending=False).index
        # Select the top 20 affected sectors
        top_20_country_scenario_1 = df_country_scenario_1.loc[sorted_indices].
 \rightarrowhead(20)
        top_20_country_scenario_2 = df_country_scenario_2.loc[sorted_indices].
 \rightarrowhead(20)
        # Compute the additional impact from Intra-EU imports (Scenario 2 -
 \hookrightarrowScenario 1)
        diff_scenario = top_20_country_scenario_2["Price Change"] -__

→top_20_country_scenario_1["Price Change"]
```

```
# Determine the y-axis limit
      max_shock = top_20_country_scenario_2["Price Change"].max()
      y_axis_limit = get_y_axis_limit(max_shock)
      # Create stacked bar chart
      plt.figure(figsize=(12, 6))
      plt.bar(
          top_20_country_scenario_1.index,
          top_20_country_scenario_1["Price Change"],
          label="Extra-EU Imports Shock",
          color="royalblue"
      )
      plt.bar(
          top_20_country_scenario_1.index,
          diff_scenario,
          bottom=top_20_country_scenario_1["Price Change"],
          label="Intra-EU Impact",
          color="darkorange"
      )
      # Format plot
      plt.title(f"Top 20 Sectors in {country}: Gas Shock (Stacked)")
      plt.ylabel("Relative Price Change (%)")
      plt.xlabel("Sectors")
      plt.xticks(rotation=45, ha="right")
      plt.ylim(0, y_axis_limit) # Set scaling
      plt.grid(axis="y", linestyle="--", alpha=0.7)
      plt.legend()
      plt.gca().yaxis.set_major_formatter(plt.FuncFormatter(lambda x, _: f"{x:
\hookrightarrow.0f}%")) # Format y-axis as percentage
      # Save figure
      plt.savefig(os.path.join(figures_dir, f"Stacked_Sectors_{country}).
→png"), bbox_inches="tight")
      plt.close()
      print(f"Saved stacked bar figure for {country}")
  except KeyError:
      print(f"Skipping {country}: No data available.")
```

```
# Read the CSV file into a DataFrame with MultiIndex
shock_results_df = pd.read_csv(output_file, index_col=["Country", "Sector"])
# Sort sectors by the absolute price change (largest impacts)
shock_results_sorted = shock_results_df.abs().sort_values(by="Price Change",__
 ⇔ascending=False)
# Select the top 20 sectors with the highest price impact
top_20_sectors = shock_results_sorted.head(20)
# Plot the results
plt.figure(figsize=(12, 6))
top_20_sectors.plot(kind="bar", legend=False, title="Top 20 Sectors by Price_
 →Impact (500% Gas Shock to imports into EU28)")
plt.ylabel("Relative Price Change")
plt.xlabel("Sectors")
plt.xticks(rotation=45, ha="right")
plt.grid(axis="y", linestyle="--", alpha=0.7)
# Show the plot
plt.show()
# Read CSV and restore MultiIndex
shock_results_df = pd.read_csv(output_file, index_col=["Country", "Sector"])
# Convert to percentage changes (already stored in % but ensuring consistency)
shock_results_df["Price Change"] *= 100
# Normalize by taking the mean absolute price impact per sector across all _{\sqcup}
 -countries
sector_impact_normalized = shock_results_df.groupby(level="Sector")["Price_
 →Change"].mean().abs()
# Sort sectors by the most affected
top_20_sectors = sector_impact_normalized.sort_values(ascending=False).head(20)
# Plot the results
plt.figure(figsize=(12, 6))
top_20_sectors.plot(kind="bar", legend=False, title="Top 20 Most Affected_u
 ⇔Sectors (Normalized)")
plt.ylabel("Average Price Impact (%)") # Normalized
plt.xlabel("Sectors")
plt.xticks(rotation=45, ha="right")
plt.grid(axis="y", linestyle="--", alpha=0.7)
# Show the plot
```

plt.show()