## harry-browne-permanent-portfolio

April 16, 2025

# 1 Does Harry Browne's permanent portfolio withstand the test of time?

## 1.1 Python Imports

```
[1]: # Standard Library
     import os
     import sys
     import datetime
     import random
     import warnings
     import random
     from pathlib import Path
     # Data Handling
     import pandas as pd
     import numpy as np
     # Data Visualization
     import matplotlib.pyplot as plt
     import matplotlib.ticker as mtick
     import matplotlib.dates as mdates
     import seaborn as sns
     # import dataframe_image as dfi
     from matplotlib.ticker import FuncFormatter, FormatStrFormatter, MultipleLocator
     # Data Sources
     import yfinance as yf
     # Statistical Analysis
     import statsmodels.api as sm
     from sklearn.decomposition import PCA
     from sklearn.preprocessing import StandardScaler
     # Suppress warnings if needed
     # warnings.filterwarnings("ignore")
```

## 1.2 Set Base Directory

```
[2]: # Add the base directory to the system path
base_directory = "/home/jared/Cloud_Storage/Dropbox/Quant_Finance_Research"
sys.path.append(base_directory)
```

## 1.3 Track Index Dependencies

```
[3]: # Define the dependency tracker
dep_file = Path("index_dep.txt")
Path("index_dep.txt").write_text("")

def export_and_track(md_filename: str, content: str):
    """Export markdown and append to index_dependencies.txt"""
    Path(md_filename).write_text(content)
    with dep_file.open("a") as f:
        f.write(md_filename + "\n")
    print(f" Exported and tracked: {md_filename}")
```

## 1.4 Import Functions

```
[4]: # TBD
```

## 2 Begin Index.md Here

#### 2.1 Python Functions

### 2.1.1 bb\_data\_updater

```
[5]: # This function takes an excel export from Bloomberg and
# removes all excess data leaving date and close columns

# Imports
import pandas as pd

# Function definition
def bb_data_updater(fund):

# File name variable
file = fund + ".xlsx"

# Import data from file as a pandas dataframe
df = pd.read_excel(file, sheet_name = 'Worksheet', engine='openpyxl')

# Set the column headings from row 5 (which is physically row 6)
df.columns = df.iloc[5]
```

```
# Set the column heading for the index to be "None"
df.rename_axis(None, axis=1, inplace = True)
# Drop the first 6 rows, 0 - 5
df.drop(df.index[0:6], inplace=True)
# Set the date column as the index
df.set_index('Date', inplace = True)
# Drop the volume column
try:
   df.drop(columns = {'PX_VOLUME'}, inplace = True)
except KeyError:
   pass
# Rename column
df.rename(columns = {'PX_LAST':'Close'}, inplace = True)
# Sort by date
df.sort_values(by=['Date'], inplace = True)
# Export data to excel
file = fund + "_Clean.xlsx"
df.to_excel(file, sheet_name='data')
# Output confirmation
print(f"The last date of data for {fund} is: ")
print(df[-1:])
print(f"Bloomberg data conversion complete for {fund} data")
return print(f"----")
```

#### 2.1.2 dp

```
[6]: # Need to update

# Set number of decimal places in pandas

def dp(decimal_places):
    pd.set_option('display.float_format', lambda x: f'%.{decimal_places}f' % x)

dp(3)
```

#### 2.1.3 df info

```
[7]: # Need to update

# The `df_info` function returns some useful information about
# a dataframe, such as the columns, data types, and size.

def df_info(df):
    print('The columns, shape, and data types are:')
    print(df.info())
    print('The first 5 rows are:')
    display(df.head())
    print('The last 5 rows are:')
    display(df.tail())
```

#### 2.1.4 load data

```
[8]: def load_data(file):
    # Import CSV
    try:
        df = pd.read_csv(file)
    except:
        pass

# Import excel
    try:
        df = pd.read_excel(file, sheet_name='data', engine='openpyxl')
    except:
        pass
    return df
```

## 2.1.5 strategy

```
[9]: def strategy(
    fund_list,
    starting_cash,
    cash_contrib,
    close_prices_df,
    rebal_month,
    rebal_day,
    rebal_per_high,
    rebal_per_low
):

####

Execute the rebalance strategy based on specified criteria.
```

```
Args:
      fund_list (str): List of funds for data to be combined from. Funds are
\hookrightarrowstrings in the form "BTC-USD".
      starting_cash (int): Starting investment balance.
      cash contrib (int): Cash contribution to be made daily.
      close\_prices\_df (pd.DataFrame): DataFrame containing date and close_{\sqcup}
⇔prices for all funds to be included.
      rebal_month (int): Month for annual rebalance.
      rebal_day (int): Day for annual rebalance.
      rebal_per_high (float): High percentage for rebalance.
      rebal_per_low (float): Low percentage for rebalance.
  Returns:
      pd.DataFrame: DataFrame containing strategy data for all funds to be_{\sqcup}
→included. Also dumps the df to excel for reference later.
  num_funds = len(fund_list)
  df = close_prices_df.copy()
  df.reset_index(inplace = True)
  # Date to be used for annual rebalance
  target month = rebal month
  target_day = rebal_day
  # Create a dataframe with dates from the specific month
  rebal_date = df[df['Date'].dt.month == target_month]
  # Specify the date or the next closest
  rebal_date = rebal_date[rebal_date['Date'].dt.day >= target_day]
  # Group by year and take the first entry for each year
  rebal_dates_by_year = rebal_date.groupby(rebal_date['Date'].dt.year).
→first().reset_index(drop=True)
  Column order for the dataframe:
  df[fund + "_BA_Shares"]
  df[fund + "_BA_$_Invested"]
  df[fund + "_BA_Port_%"]
  df['Total_BA_$_Invested']
  df['Contribution']
  df['Rebalance']
  df[fund + "_AA_Shares"]
  df[fund + "_AA_$_Invested"]
```

```
df[fund + "_AA_Port_%"]
  df['Total_AA_$_Invested']
  # Calculate the columns and initial values for before action (BA) shares, \$_{\sqcup}
⇒invested, and port %
  for fund in fund list:
       df[fund + "_BA_Shares"] = starting_cash / num_funds / df[fund +__

¬"_Close"]

      df[fund + "_BA_$_Invested"] = df[fund + "_BA_Shares"] * df[fund +__

¬"_Close"]

      df[fund + "_BA_Port_%"] = 0.25
  # Set column values initially
  df['Total_BA_$_Invested'] = starting_cash
  df['Contribution'] = 0
  # df['Contribution'] = cash_contrib
  df['Rebalance'] = "No"
  # Set columns and values initially for after action (AA) shares, \$_{\sqcup}
⇒invested, and port %
  for fund in fund list:
      df[fund + "_AA_Shares"] = starting_cash / num_funds / df[fund +__

¬"_Close"]

      df[fund + "_AA_$_Invested"] = df[fund + "_AA_Shares"] * df[fund +__

¬"_Close"]

      df[fund + "_AA_Port_%"] = 0.25
  # Set column value for after action (AA) total $ invested
  df['Total_AA_$_Invested'] = starting_cash
  # Iterate through the dataframe and execute the strategy
  for index, row in df.iterrows():
       # Ensure there's a previous row to reference by checking the index value
      if index > 0:
           # Initialize variable
           Total BA Invested = 0
           # Calculate before action (BA) shares and $ invested values
           for fund in fund_list:
               df.at[index, fund + "_BA_Shares"] = df.at[index - 1, fund +__
→" AA Shares"]
               df.at[index, fund + "_BA_$_Invested"] = df.at[index, fund +__

¬"_BA_Shares"] * row[fund + "_Close"]
```

```
# Sum the asset values to find the total
               Total_BA_Invested = Total_BA_Invested + df.at[index, fund +

¬"_BA_$_Invested"]

           # Calculate before action (BA) port % values
           for fund in fund list:
               df.at[index, fund + "_BA_Port_%"] = df.at[index, fund +__
→"_BA_$_Invested"] / Total_BA_Invested
           # Set column for before action (BA) total $ invested
           df.at[index, 'Total_BA_$_Invested'] = Total_BA_Invested
           # Initialize variables
           rebalance = "No"
           date = row['Date']
           # Check for a specific date annually
           # Simple if statement to check if date_to_check is in_
→ jan_28_or_after_each_year
           if date in rebal_dates_by_year['Date'].values:
               rebalance = "Yes"
           else:
               pass
           # Check to see if any asset has portfolio percentage of greater_
⇔than 35% or less than 15% and if so set variable
           for fund in fund_list:
               if df.at[index, fund + "_BA_Port_%"] > rebal_per_high or df.
→at[index, fund + "_BA_Port_%"] < rebal_per_low:</pre>
                   rebalance = "Yes"
               else:
                   pass
           # If rebalance is required, rebalance back to 25% for each asset,_{\sf L}
⇔else just divide contribution evenly across assets
           if rebalance == "Yes":
               df.at[index, 'Rebalance'] = rebalance
               for fund in fund list:
                       df.at[index, fund + "_AA_$_Invested"] =__
→(Total_BA_Invested + df.at[index, 'Contribution']) * 0.25
           else:
               df.at[index, 'Rebalance'] = rebalance
               for fund in fund_list:
                       df.at[index, fund + "_AA_$_Invested"] = df.at[index,__

¬fund + "_BA_$_Invested"] + df.at[index, 'Contribution'] * 0.25
```

```
# Initialize variable
          Total_AA_Invested = 0
           # Set column values for after action (AA) shares and port %
          for fund in fund_list:
               df.at[index, fund + "_AA_Shares"] = df.at[index, fund +__

¬"_AA_$_Invested"] / row[fund + "_Close"]

               # Sum the asset values to find the total
               Total_AA_Invested = Total_AA_Invested + df.at[index, fund +

¬"_AA_$_Invested"]

           # Calculate after action (AA) port % values
          for fund in fund_list:
               df.at[index, fund + "_AA_Port_%"] = df.at[index, fund +__
→"_AA_$_Invested"] / Total_AA_Invested
           # Set column for after action (AA) total $ invested
           df.at[index, 'Total_AA_$_Invested'] = Total_AA_Invested
       # If this is the first row
      else:
          pass
  df['Return'] = df['Total_AA_$_Invested'].pct_change()
  df['Cumulative_Return'] = (1 + df['Return']).cumprod()
  plan_name = '_'.join(fund_list)
  file = plan_name + "_Strategy.xlsx"
  location = file
  df.to_excel(location, sheet_name="data")
  print(f"Strategy complete for {plan_name}.")
  return df
```

#### 2.1.6 summary stats

```
[10]: # Stats for entire data set
def summary_stats(
    fund_list,
    df,
    period,
    excel_export
):
    """
    Calculate summary statistics for the given fund list and return data.
```

```
Args:
       fund_list (str): List of funds for data to be combined from. Funds are
\hookrightarrowstrings in the form "BTC-USD".
       df (df): Dataframe with return data.
       period (str): Period for which to calculate statistics. Options are
→ "Monthly", "Weekly", "Daily", "Hourly".
       excel_export (bool): If True, export to excel file.
  Returns:
      pd.DataFrame: DataFrame containing various portfolio statistics.
  if period == "Monthly":
      timeframe = 12 # months
  elif period == "Weekly":
      timeframe = 52 # weeks
  elif period == "Daily":
       timeframe = 365 # days
  elif period == "Hourly":
       timeframe = 8760 # hours
  else:
      return print("Error, check inputs")
  df_stats = pd.DataFrame(df.mean(axis=0) * timeframe) # annualized
  \# df_stats = pd.DataFrame((1 + df.mean(axis=0)) ** timeframe - 1) #_U
annualized, this is this true annualized return but we will simply use the
\rightarrowmean
  df stats.columns = ['Annualized Mean']
  df_stats['Annualized Volatility'] = df.std() * np.sqrt(timeframe) #__
\rightarrow annualized
  df_stats['Annualized Sharpe Ratio'] = df_stats['Annualized Mean'] /__

¬df_stats['Annualized Volatility']
  df_cagr = (1 + df['Return']).cumprod()
  cagr = (df_cagr.iloc[-1] / 1) ** (1/(len(df_cagr) / timeframe)) - 1
  df_stats['CAGR'] = cagr
  df_stats[period + ' Max Return'] = df.max()
  df_stats[period + ' Max Return (Date)'] = df.idxmax().values[0]
  df_stats[period + ' Min Return'] = df.min()
  df_stats[period + ' Min Return (Date)'] = df.idxmin().values[0]
  wealth_index = 1000*(1+df).cumprod()
  previous_peaks = wealth_index.cummax()
  drawdowns = (wealth_index - previous_peaks)/previous_peaks
```

```
df_stats['Max Drawdown'] = drawdowns.min()
  df_stats['Peak'] = [previous peaks[col][:drawdowns[col].idxmin()].idxmax()__
→for col in previous_peaks.columns]
  df stats['Bottom'] = drawdowns.idxmin()
  recovery date = []
  for col in wealth index.columns:
      prev_max = previous_peaks[col][:drawdowns[col].idxmin()].max()
      recovery_wealth = pd.DataFrame([wealth_index[col][drawdowns[col]].
→idxmin():]]).T
      recovery_date.append(recovery_wealth[recovery_wealth[col] >= prev_max].
→index.min())
  df_stats['Recovery Date'] = recovery_date
  plan_name = '_'.join(fund_list)
  # Export to excel
  if excel_export == True:
      file = plan_name + "_Summary_Stats.xlsx"
      location = file
      # location = f"{base_directory}/{strategy_name}/{file_name}.xlsx"
      df stats.to excel(location, sheet name="data")
  else:
      pass
  print(f"Summary stats complete for {plan_name}.")
  return df_stats
```

## 2.1.7 plot\_cumulative\_return

```
[11]: def plot_cumulative_return(strat_df):
    # Generate plot
    plt.figure(figsize=(10, 5), facecolor = '#F5F5F5')

# Plotting data
    plt.plot(strat_df.index, strat_df['Cumulative_Return'], label = 'Strategy_
Cumulative Return', linestyle='-', color='green', linewidth=1)

# Set X axis

# x_tick_spacing = 5 # Specify the interval for x-axis ticks
# plt.gca().xaxis.set_major_locator(MultipleLocator(x_tick_spacing))
    plt.gca().xaxis.set_major_locator(mdates.YearLocator())
    plt.gca().xaxis.set_major_formatter(mdates.DateFormatter('%Y'))
    plt.xlabel('Year', fontsize = 9)
    plt.xticks(rotation = 45, fontsize = 7)
# plt.xlim(, )
```

```
# Set Y axis
y_tick_spacing = 0.5 # Specify the interval for y-axis ticks
plt.gca().yaxis.set_major_locator(MultipleLocator(y_tick_spacing))
plt.ylabel('Cumulative Return', fontsize = 9)
plt.yticks(fontsize = 7)
# plt.ylim(0, 7.5)
# Set title, etc.
plt.title('Cumulative Return', fontsize = 12)
# Set the grid & legend
plt.tight_layout()
plt.grid(True)
plt.legend(fontsize=8)
# Save the figure
plt.savefig('03_Cumulative_Return.png', dpi=300, bbox_inches='tight')
# Display the plot
return plt.show()
```

## 2.1.8 plot\_values

```
[12]: def plot_values(strat_df):
          # Generate plot
          plt.figure(figsize=(10, 5), facecolor = '#F5F5F5')
          # Plotting data
          plt.plot(strat df.index, strat df['Total AA $ Invested'], label='Total,
       →Portfolio Value', linestyle='-', color='black', linewidth=1)
          plt.plot(strat_df.index, strat_df['Stocks_AA_$_Invested'], label='Stocks_
       →Position Value', linestyle='-', color='orange', linewidth=1)
          plt.plot(strat_df.index, strat_df['Bonds_AA_$_Invested'], label='Bond_
       →Position Value', linestyle='-', color='yellow', linewidth=1)
          plt.plot(strat_df.index, strat_df['Gold_AA_$_Invested'], label='Gold_
       ⇔Position Value', linestyle='-', color='blue', linewidth=1)
          plt.plot(strat_df.index, strat_df['Cash_AA_$_Invested'], label='Cash_u
       →Position Value', linestyle='-', color='brown', linewidth=1)
          # Set X axis
          # x_tick_spacing = 5 # Specify the interval for x-axis ticks
          # plt.qca().xaxis.set_major_locator(MultipleLocator(x_tick_spacing))
          plt.gca().xaxis.set_major_locator(mdates.YearLocator())
          plt.gca().xaxis.set_major_formatter(mdates.DateFormatter('%Y'))
          plt.xlabel('Year', fontsize = 9)
          plt.xticks(rotation = 45, fontsize = 7)
```

```
# plt.xlim(, )
   # Set Y axis
  y_tick_spacing = 5000 # Specify the interval for y-axis ticks
  plt.gca().yaxis.set_major_locator(MultipleLocator(y_tick_spacing))
  plt.gca().yaxis.set_major_formatter(mtick.FuncFormatter(lambda x, pos: '{:,..
\hookrightarrow 0f}'.format(x))) # Adding commas to y-axis labels
  plt.ylabel('Total Value ($)', fontsize = 9)
  plt.yticks(fontsize = 7)
  # plt.ylim(0, 75000)
  # Set title, etc.
  plt.title('Total Values For Stocks, Bonds, Gold, and Cash Positions and
⇔Portfolio', fontsize = 12)
  # Set the grid & legend
  plt.tight_layout()
  plt.grid(True)
  plt.legend(fontsize=8)
  # Save the figure
  plt.savefig('04_Portfolio_Values.png', dpi=300, bbox_inches='tight')
  # Display the plot
  return plt.show()
```

#### 2.1.9 plot drawdown

```
[13]: def plot_drawdown(strat_df):
                                       rolling_max = strat_df['Total_AA_$_Invested'].cummax()
                                       drawdown = (strat_df['Total_AA_$_Invested'] - rolling_max) / rolling_max *__
                            →100
                                       # Generate plot
                                       plt.figure(figsize=(10, 5), facecolor = '#F5F5F5')
                                       # Plotting data
                                       plt.plot(strat_df.index, drawdown, label='Drawdown', linestyle='-',u
                            ⇔color='red', linewidth=1)
                                       # Set X axis
                                       # x tick spacing = 5 # Specify the interval for x-axis ticks
                                       # plt.gca().xaxis.set_major_locator(MultipleLocator(x_tick_spacing))
                                       plt.gca().xaxis.set_major_locator(mdates.YearLocator())
                                       plt.gca().xaxis.set_major_formatter(mdates.DateFormatter('\( \frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac}\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\frac{\fracc}\frac{\frac{\frac{\frac{\frac
                                       plt.xlabel('Year', fontsize = 9)
                                       plt.xticks(rotation = 45, fontsize = 7)
```

```
# plt.xlim(, )
   # Set Y axis
  y_tick_spacing = 1  # Specify the interval for y-axis ticks
  plt.gca().yaxis.set_major_locator(MultipleLocator(y_tick_spacing))
  # plt.gca().yaxis.set_major_formatter(mtick.FuncFormatter(lambda x, pos: '{:
\hookrightarrow, .0f}'.format(x))) # Adding commas to y-axis labels
  plt.gca().yaxis.set_major_formatter(mtick.FuncFormatter(lambda x, pos: '{:.
\rightarrow 0f '.format(x))) # Adding 0 decimal places to y-axis labels
  plt.ylabel('Drawdown (%)', fontsize = 9)
  plt.yticks(fontsize = 7)
  # plt.ylim(-20, 0)
  # Set title, etc.
  plt.title('Portfolio Drawdown', fontsize = 12)
  # Set the grid & legend
  plt.tight_layout()
  plt.grid(True)
  plt.legend(fontsize=8)
  # Save the figure
  plt.savefig('05_Portfolio_Drawdown.png', dpi=300, bbox_inches='tight')
  # Display the plot
  return plt.show()
```

#### 2.1.10 plot asset weights

```
[14]: def plot_asset_weights(strat_df):
          # Generate plot
          plt.figure(figsize=(10, 5), facecolor = '#F5F5F5')
          # Plotting data
          plt.plot(strat_df.index, strat_df['Stocks_AA_Port_%'] * 100, label='Stocks_
       →Portfolio Weight', linestyle='-', color='orange', linewidth=1)
          plt.plot(strat_df.index, strat_df['Bonds_AA_Port_%'] * 100, label='Bonds_u
       ⇔Portfolio Weight', linestyle='-', color='yellow', linewidth=1)
          plt.plot(strat_df.index, strat_df['Gold_AA_Port_%'] * 100, label='Gold_
       →Portfolio Weight', linestyle='-', color='blue', linewidth=1)
          plt.plot(strat_df.index, strat_df['Cash_AA_Port_%'] * 100, label='Cash_
       →Portfolio Weight', linestyle='-', color='brown', linewidth=1)
          # Set X axis
          # x_tick_spacing = 5  # Specify the interval for x-axis ticks
          # plt.qca().xaxis.set_major_locator(MultipleLocator(x_tick_spacing))
          plt.gca().xaxis.set_major_locator(mdates.YearLocator())
```

```
plt.gca().xaxis.set_major_formatter(mdates.DateFormatter('%Y'))
  plt.xlabel('Year', fontsize = 9)
  plt.xticks(rotation = 45, fontsize = 7)
  # plt.xlim(, )
  # Set Y axis
  y_tick_spacing = 1  # Specify the interval for y-axis ticks
  plt.gca().yaxis.set_major_locator(MultipleLocator(y_tick_spacing))
  # plt.gca().yaxis.set_major_formatter(mtick.FuncFormatter(lambda x, pos: '{:
\hookrightarrow, .0f}'.format(x))) # Adding commas to y-axis labels
  plt.ylabel('Asset Weight (%)', fontsize = 9)
  plt.yticks(fontsize = 7)
  # plt.ylim(14, 36)
  # Set title, etc.
  plt.title('Portfolio Asset Weights For Stocks, Bonds, Gold, and Cash⊔
⇔Positions', fontsize = 12)
  # Set the grid & legend
  plt.tight_layout()
  plt.grid(True)
  plt.legend(fontsize=8)
  # Save the figure
  plt.savefig('07_Portfolio_Weights.png', dpi=300, bbox_inches='tight')
  # Display the plot
  return plt.show()
```

#### 2.1.11 plot\_annual\_returns

```
def plot_annual_returns(return_df):
    # Generate plot
    plt.figure(figsize=(10, 5), facecolor = '#F5F5F5')

# Plotting data
    plt.bar(return_df.index, return_df['Return'] * 100, label='Annual Returns',u
width=0.5) # width adjusted for better spacing

# Set X axis
    x_tick_spacing = 1 # Specify the interval for x-axis ticks
    plt.gca().xaxis.set_major_locator(MultipleLocator(x_tick_spacing))
    # plt.gca().xaxis.set_major_locator(mdates.YearLocator())
    # plt.gca().xaxis.set_major_formatter(mdates.DateFormatter('%Y'))
    plt.xlabel('Year', fontsize = 9)
    plt.xticks(rotation = 45, fontsize = 7)
# plt.xlim(, )
```

```
# Set Y axis
  y_tick_spacing = 1  # Specify the interval for y-axis ticks
  plt.gca().yaxis.set_major_locator(MultipleLocator(y_tick_spacing))
  # plt.gca().yaxis.set_major_formatter(mtick.FuncFormatter(lambda x, pos: '{:
\hookrightarrow, .0f}'.format(x))) # Adding commas to y-axis labels
  plt.ylabel('Annual Return (%)', fontsize = 9)
  plt.yticks(fontsize = 7)
  # plt.ylim(-20, 20)
  # Set title, etc.
  plt.title('Portfolio Annual Returns', fontsize = 12)
  # Set the grid & legend
  plt.tight_layout()
  plt.grid(True)
  plt.legend(fontsize=8)
  # Save the figure
  plt.savefig('08_Portfolio_Annual_Returns.png', dpi=300, bbox_inches='tight')
  # Display the plot
  return plt.show()
```

## 2.2 Import Data

/home/jared/python-virtual-envs/general\_313/lib/python3.13/site-packages/pandas/core/indexes/base.py:7588: FutureWarning: Dtype inference on a pandas object (Series, Index, ExtensionArray) is deprecated. The Index constructor will keep the original dtype in the future. Call `infer\_objects` on the result to get the old behavior.

return Index(sequences[0], name=names)

The last date of data for SPBDU10T\_S&P US Treasury Bond 7-10 Year Total Return

#### Index is:

Close

Date

2024-04-30 579.024

Bloomberg data conversion complete for SPBDU10T\_S&P US Treasury Bond 7-10 Year Total Return Index data

-----

[16]:		${\tt Bonds\_Close}$	Bonds_Daily_Return	Bonds_Total_Return
	Date			
	1990-01-02	99.972	NaN	NaN
	1990-01-03	99.733	-0.002	0.998
	1990-01-04	99.813	0.001	0.998
	1990-01-05	99.769	-0.000	0.998
	1990-01-08	99.681	-0.001	0.997
	•••	•••	•••	•••
	2023-12-22	604.166	-0.001	6.043
	2023-12-26	604.555	0.001	6.047
	2023-12-27	609.355	0.008	6.095
	2023-12-28	606.828	-0.004	6.070
	2023-12-29	606.185	-0.001	6.064

[8527 rows x 3 columns]

/home/jared/python-virtual-envs/general\_313/lib/python3.13/site-packages/pandas/core/indexes/base.py:7588: FutureWarning: Dtype inference on a pandas object (Series, Index, ExtensionArray) is deprecated. The Index constructor will keep the original dtype in the future. Call `infer\_objects` on the result to get the old behavior.

return Index(sequences[0], name=names)

The last date of data for SPXT\_S&P 500 Total Return Index is: Close

Date

2024-04-30 10951.660

Bloomberg data conversion complete for SPXT\_S&P 500 Total Return Index data

-----

/tmp/ipykernel\_16195/1682549453.py:8: FutureWarning: The default
fill\_method='pad' in Series.pct\_change is deprecated and will be removed in a
future version. Either fill in any non-leading NA values prior to calling
pct\_change or specify 'fill\_method=None' to not fill NA values.
 stocks\_data['Stocks\_Daily\_Return'] = stocks\_data['Stocks\_Close'].pct\_change()

[17]:		Stocks_Close	Stocks_Daily_Return	Stocks_Total_Return
	Date			
	1990-01-01	NaN	NaN	NaN
	1990-01-02	386.160	NaN	NaN
	1990-01-03	385.170	-0.003	0.997
	1990-01-04	382.020	-0.008	0.989
	1990-01-05	378.300	-0.010	0.980
	•••	•••	•••	•••
	2023-12-22	10292.370	0.002	26.653
	2023-12-26	10335.980	0.004	26.766
	2023-12-27	10351.600	0.002	26.807
	2023-12-28	10356.590	0.000	26.819
	2023-12-29	10327.830	-0.003	26.745

[8584 rows x 3 columns]

/home/jared/python-virtual-envs/general\_313/lib/python3.13/site-packages/pandas/core/indexes/base.py:7588: FutureWarning: Dtype inference on a pandas object (Series, Index, ExtensionArray) is deprecated. The Index constructor will keep the original dtype in the future. Call `infer\_objects` on the result to get the old behavior.

return Index(sequences[0], name=names)

The last date of data for XAU\_Gold USD Spot is: Close

Date

2024-05-01 2299.310

Bloomberg data conversion complete for XAU\_Gold USD Spot data

-----

```
[18]:
                  Gold_Close Gold_Daily_Return Gold_Total_Return
     Date
      1990-01-02
                     399.000
                                            NaN
                                                               NaN
      1990-01-03
                     395.000
                                         -0.010
                                                             0.990
      1990-01-04
                     396.500
                                          0.004
                                                             0.994
      1990-01-05
                     405.000
                                          0.021
                                                              1.015
      1990-01-08
                     404.600
                                         -0.001
                                                             1.014
                                                             5.146
      2023-12-22
                    2053.080
                                          0.003
      2023-12-26
                    2067.810
                                          0.007
                                                             5.182
                    2077.490
                                          0.005
                                                             5.207
      2023-12-27
      2023-12-28
                    2065.610
                                         -0.006
                                                             5.177
      2023-12-29
                    2062.980
                                         -0.001
                                                             5.170
      [8819 rows x 3 columns]
[19]: # Merge the stock data and bond data into a single DataFrame using their
      ⇔indices (dates)
      perm_port = pd.merge(stocks_data['Stocks_Close'], bonds_data['Bonds_Close'],_u
       ⇔left_index=True, right_index=True)
      # Add gold data to the portfolio DataFrame by merging it with the existing data_
       ⇔on indices (dates)
      perm_port = pd.merge(perm_port, gold_data['Gold_Close'], left_index=True,__
       →right_index=True)
      # Add a column for cash with a constant value of 1 (assumes the value of cash_
       ⇔remains constant at $1 over time)
      perm_port['Cash_Close'] = 1
      # Remove any rows with missing values (NaN) to ensure clean data for further
       ⇔analysis
      perm port.dropna(inplace=True)
      # Display the finalized portfolio DataFrame
      perm_port
Γ197:
                  Stocks_Close Bonds_Close Gold_Close Cash_Close
     Date
      1990-01-02
                       386.160
                                     99.972
                                                399.000
                                                                   1
      1990-01-03
                       385.170
                                     99.733
                                                395.000
                                                                   1
                                     99.813
                                                                   1
      1990-01-04
                       382.020
                                                396.500
      1990-01-05
                       378.300
                                     99.769
                                                405.000
      1990-01-08
                       380.040
                                     99.681
                                                404.600
                                                                   1
```

604.166

604.555

2053.080

2067.810

1

1

2023-12-22

2023-12-26

10292.370

10335.980

2023-12-27	10351.600	609.355	2077.490	1
2023-12-28	10356.590	606.828	2065.610	1
2023-12-29	10327.830	606.185	2062.980	1

[8479 rows x 4 columns]

# [20]: # Check for any missing values in each column missing\_values = perm\_port.isnull().any()

# Display columns with missing values
print(missing\_values)

Stocks\_Close False
Bonds\_Close False
Gold\_Close False
Cash\_Close False

dtype: bool

## [21]: df\_info(perm\_port)

The columns, shape, and data types are:

<class 'pandas.core.frame.DataFrame'>

DatetimeIndex: 8479 entries, 1990-01-02 to 2023-12-29

Data columns (total 4 columns):

#	Column	Non-Null Count	Dtype
0	Stocks_Close	8479 non-null	float64
1	Bonds_Close	8479 non-null	float64
2	${\tt Gold\_Close}$	8479 non-null	float64
3	Cash_Close	8479 non-null	int64

dtypes: float64(3), int64(1)

memory usage: 331.2 KB

None

The first 5 rows are:

	Stocks_Close	${ t Bonds\_Close}$	${ t Gold\_Close}$	${\tt Cash\_Close}$
Date				
1990-01-02	386.160	99.972	399.000	1
1990-01-03	385.170	99.733	395.000	1
1990-01-04	382.020	99.813	396.500	1
1990-01-05	378.300	99.769	405.000	1
1990-01-08	380.040	99.681	404.600	1

The last 5 rows are:

	Stocks_Close	${ t Bonds\_Close}$	${ t Gold\_Close}$	Cash_Close
Date				
2023-12-22	10292.370	604.166	2053.080	1
2023-12-26	10335.980	604.555	2067.810	1
2023-12-27	10351.600	609.355	2077.490	1

## 2.3 Execute Strategy

```
[22]: # List of funds to be used
      fund_list = ['Stocks', 'Bonds', 'Gold', 'Cash']
      # Starting cash contribution
      starting_cash = 10000
      # Monthly cash contribution
      cash_contrib = 0
      strat = strategy(
          fund_list=fund_list,
          starting_cash=starting_cash,
          cash_contrib=cash_contrib,
          close_prices_df=perm_port,
          rebal_month=1,
          rebal day=1,
          rebal per high=0.35,
          rebal_per_low=0.15)
      strat = strat.set_index('Date')
      sum_stats = summary_stats(
          fund_list=fund_list,
          df=strat[['Return']],
          period="Daily",
          excel_export=False)
      strat_pre_1999 = strat[strat.index < '2000-01-01']
      sum_stats_pre_1999 = summary_stats(
          fund_list=fund_list,
          df=strat pre 1999[['Return']],
          period="Daily",
          excel_export=False)
      strat_post_1999 = strat[strat.index >= '2000-01-01']
      sum_stats_post_1999 = summary_stats(
          fund_list=fund_list,
          df=strat_post_1999[['Return']],
          period="Daily",
          excel_export=False)
      strat_post_2009 = strat[strat.index >= '2010-01-01']
```

```
sum_stats_post_2009 = summary_stats(
          fund_list=fund_list,
          df=strat_post_2009[['Return']],
          period="Daily",
          excel_export=False)
     /tmp/ipykernel_16195/15022227.py:104: FutureWarning: Setting an item of
     incompatible dtype is deprecated and will raise an error in a future version of
     pandas. Value '9962.55140947963' has dtype incompatible with int64, please
     explicitly cast to a compatible dtype first.
       df.at[index, 'Total_BA_$_Invested'] = Total_BA_Invested
     /tmp/ipykernel_16195/15022227.py:149: FutureWarning: Setting an item of
     incompatible dtype is deprecated and will raise an error in a future version of
     pandas. Value '9962.55140947963' has dtype incompatible with int64, please
     explicitly cast to a compatible dtype first.
       df.at[index, 'Total_AA_$_Invested'] = Total_AA_Invested
     Strategy complete for Stocks_Bonds_Gold_Cash.
     Summary stats complete for Stocks Bonds Gold Cash.
     Summary stats complete for Stocks_Bonds_Gold_Cash.
     Summary stats complete for Stocks Bonds Gold Cash.
     Summary stats complete for Stocks_Bonds_Gold_Cash.
[23]: all_sum_stats = pd.concat([sum_stats])
      all sum stats = all sum stats.rename(index={'Return': '1990 - 2023'})
      all_sum_stats = pd.concat([all_sum_stats, sum_stats_pre_1999])
      all sum stats = all sum stats.rename(index={'Return': 'Pre 1999'})
      all sum stats = pd.concat([all sum stats, sum stats post 1999])
      all_sum_stats = all_sum_stats.rename(index={'Return': 'Post 1999'})
      all_sum_stats = pd.concat([all_sum_stats, sum_stats_post_2009])
      all_sum_stats = all_sum_stats.rename(index={'Return': 'Post 2009'})
      all_sum_stats
[23]:
                   Annualized Mean Annualized Volatility Annualized Sharpe Ratio \
      1990 - 2023
                             0.083
                                                    0.072
                                                                             1.152
     Pre 1999
                             0.088
                                                    0.060
                                                                             1.453
     Post 1999
                             0.081
                                                    0.077
                                                                             1.063
     Post 2009
                             0.081
                                                    0.073
                                                                             1.115
                   CAGR Daily Max Return Daily Max Return (Date) Daily Min Return \
      1990 - 2023 0.084
                                    0.029
                                                       2020-03-24
                                                                             -0.030
      Pre 1999
                  0.089
                                    0.022
                                                       1999-09-28
                                                                             -0.018
```

2020-03-24

2020-03-24

Peak

-0.154 2008-03-18 2008-11-12

-0.062 1998-07-20 1998-08-31

-0.030

-0.030

Bottom \

0.029

0.029

Daily Min Return (Date) Max Drawdown

2020-03-12

1993-08-05

Post 1999

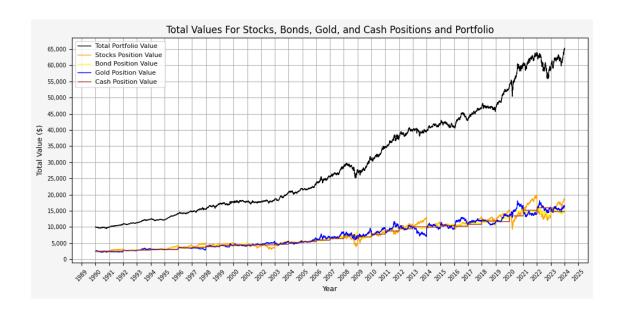
Post 2009

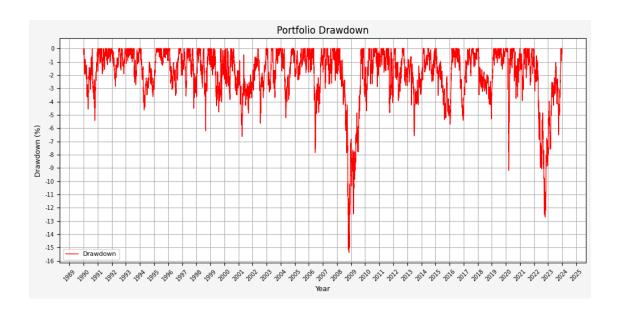
1990 - 2023 Pre 1999 0.082

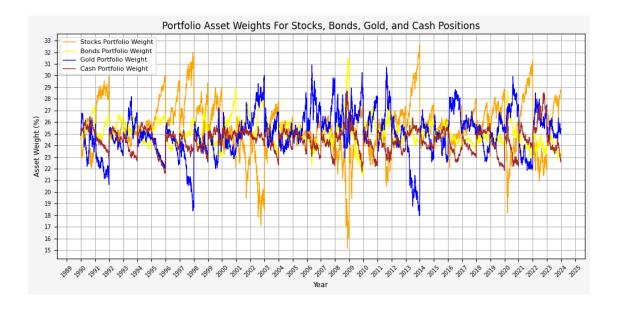
0.082

```
Post 1999
                               2020-03-12
                                                 -0.154 2008-03-18 2008-11-12
      Post 2009
                               2020-03-12
                                                 -0.127 2021-12-27 2022-10-20
                  Recovery Date
      1990 - 2023
                     2009-10-06
     Pre 1999
                     1998-11-05
     Post 1999
                     2009-10-06
     Post 2009
                     2023-12-01
[24]: plot_cumulative_return(strat)
      plot_values(strat)
      plot_drawdown(strat)
      plot_asset_weights(strat)
      # Create dataframe for the annual returns
      strat_annual_returns = strat['Cumulative_Return'].resample('Y').last().
      →pct_change().dropna()
      strat_annual_returns_df = strat_annual_returns.to_frame()
      strat_annual_returns_df['Year'] = strat_annual_returns_df.index.year # Add a__
      →'Year' column with just the year
      strat annual returns df.reset index(drop=True, inplace=True) # Reset the index
      ⇔to remove the datetime index
      # Now the DataFrame will have 'Year' and 'Cumulative Return' columns
      strat_annual_returns_df = strat_annual_returns_df[['Year',__
       → 'Cumulative_Return']] # Keep only 'Year' and 'Cumulative_Return' columns
      strat annual returns df.rename(columns = {'Cumulative Return': 'Return'},
      ⇔inplace=True)
      strat_annual_returns_df.set_index('Year', inplace=True)
      display(strat_annual_returns_df)
      plan_name = '_'.join(fund_list)
      file = plan_name + "_Annual_Returns.xlsx"
      location = file
      strat_annual_returns_df.to_excel(location, sheet_name='data')
      plot_annual_returns(strat_annual_returns_df)
```









/tmp/ipykernel\_16195/2757918626.py:7: FutureWarning: 'Y' is deprecated and will be removed in a future version, please use 'YE' instead.

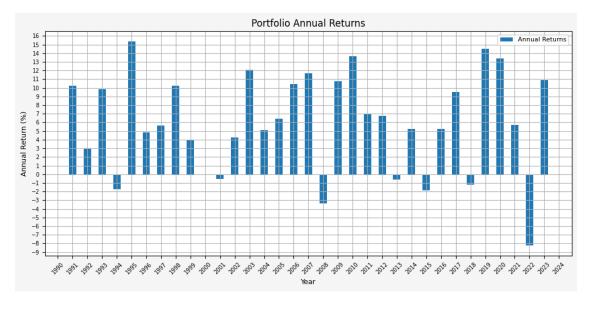
strat\_annual\_returns =

strat['Cumulative\_Return'].resample('Y').last().pct\_change().dropna()

### Return

Year 1991 0.102 1992 0.030 1993 0.099 1994 -0.017

```
1995
       0.153
1996
       0.049
1997
       0.056
1998
       0.102
1999
       0.039
2000
       0.000
2001
      -0.005
2002
       0.043
2003
       0.121
2004
       0.051
2005
       0.064
2006
       0.104
2007
       0.117
2008
      -0.033
2009
       0.107
2010
       0.137
2011
       0.070
2012
       0.068
2013
      -0.006
2014
       0.052
2015
      -0.018
2016
       0.052
2017
       0.095
2018
      -0.012
2019
       0.145
2020
       0.134
2021
       0.057
2022
      -0.082
2023
       0.109
```



#### 2.4 Calculate stats for various rebalance dates

```
[25]: # # List of funds to be used
      # fund_list = ['Stocks', 'Bonds', 'Gold', 'Cash']
      # # Starting cash contribution
      # starting_cash = 10000
      # # Monthly cash contribution
      \# cash contrib = 0
      # months = list(range(1, 13))
      \# days = list(range(1, 28))
      \# stats = pd.DataFrame(columns = ['Rebal_Month', 'Rebal_Day', 'Annualized_
       Mean', 'Annualized Volatility', 'Annualized Sharpe Ratio', 'CAGR',
                                         'Daily Max Return', 'Daily Max Return
       → (Date)', 'Daily Min Return', 'Daily Min Return (Date)', 'Max Drawdown',
                                        'Peak', 'Bottom', 'Recovery Date'])
      # for month in months:
           for day in days:
                strat = strategy(fund_list, starting_cash, cash_contrib, perm_port,_
       ⇔month, day).set_index('Date')
                sum_stats = summary_stats(fund_list, strat[['Return']], 'Daily')
                stats = pd.concat([stats, sum_stats], ignore_index=True)
                stats.loc[stats.index[-1], 'Rebal_Month'] = month
                stats.loc[stats.index[-1], 'Rebal_Day'] = day
                display(stats)
      # plan_name = '_'.join(fund_list)
      # file = plan_name + "_All_Summary_Stats.xlsx"
      # location = file
      # stats.to_excel(location, sheet_name='data')
      # print(f"All summary stats complete for {plan_name}.")
```