IBM Software

Clustering Course Lab 2: Portfolio construction

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Constructing risk efficient portfolios

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Summary

Hello everybody! Welcome to second lab of this series.

We are already on our way to construct our first investment portfolio. Our portfolio will consist of 3 stocks.

The question is, if we decided to invest in these 3 stocks, how we should allocate our capital?

Markowitz portfolio theory has an answer for the problem. We will see how we can implement markowitz portfolio theory to create efficient portfolios.

Let's get started.

1. Downloading historical data for 3 companies and S&P 500 Index

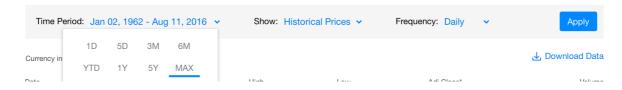
I will open my browser and navigate to finance.yahoo.com

I will download historical data for IBM, Procter & Gamble and Goldman Sachs.

I will type "IBM" in search box and will select IBM.



In resulting page, I will adjust time range by selecting "MAX" and click "Apply" then click "Download".



I will rename file as "IBM.csv"

I will go through same steps and download data for Procter & Gamble and Goldman Sachs as well. I will rename files as "PG.csv" and "GS.csv".

In CAPM, we use market return and for U.S. stock markets we can use S&P 500 Index for that purpose.

I will download historical data for S&P 500 Index and rename it as "sp500.csv"



2. Data preparation

I will open SPSS Modeler and create a new stream, I will save stream as "Lab_2" to my labs folder.

I will add 4 "Var. File" node from "Sources" palette and import downloaded datasets to my stream.



Var. File



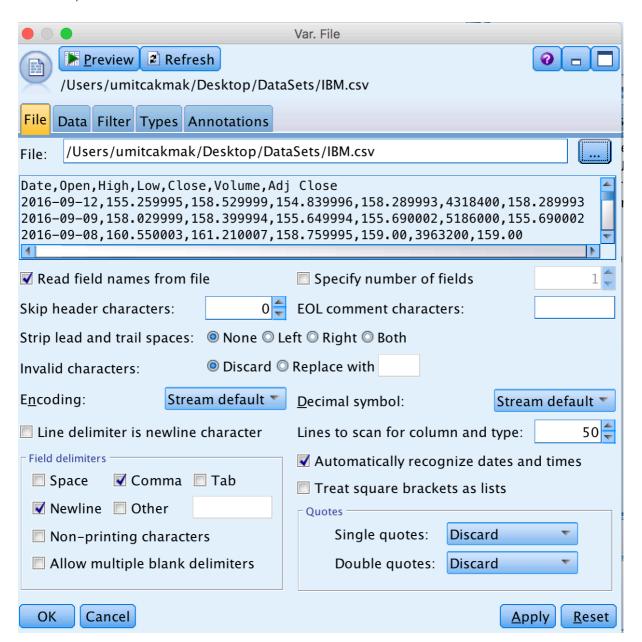
Var. File



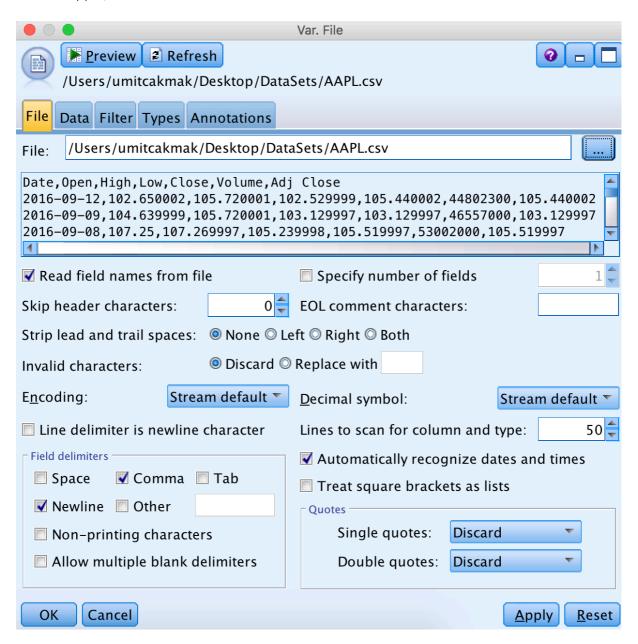
Var. File



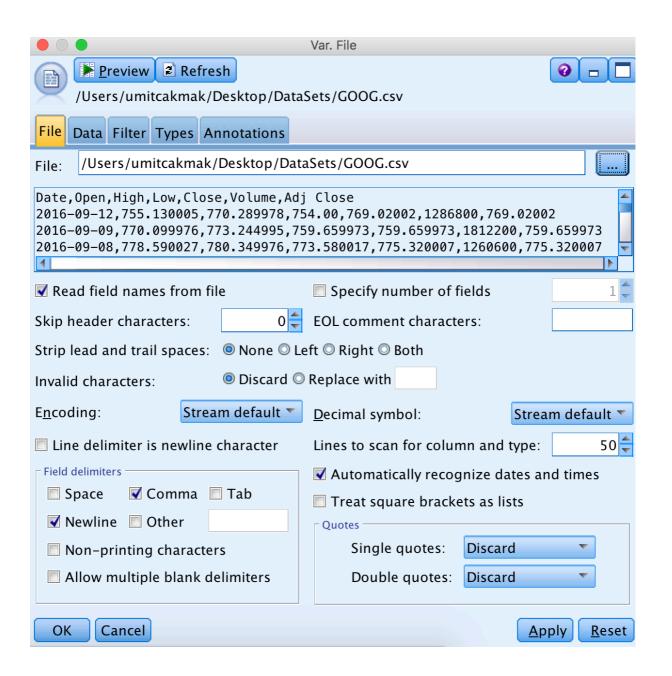
First IBM,

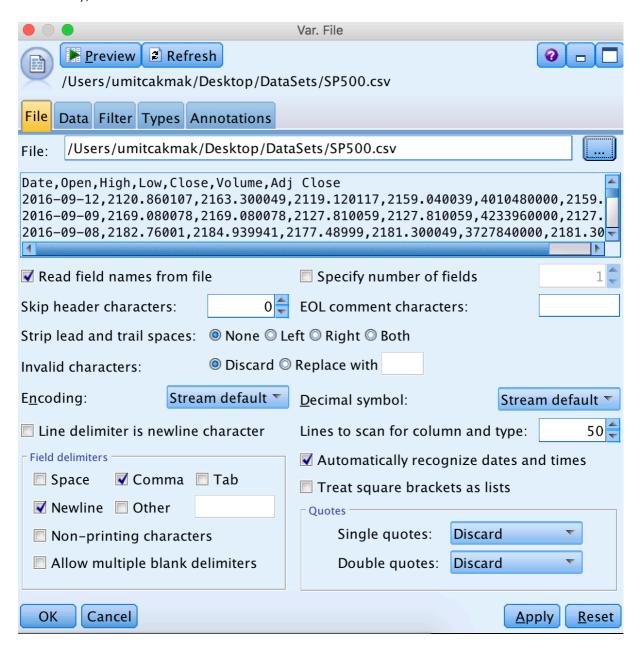


Second Apple,

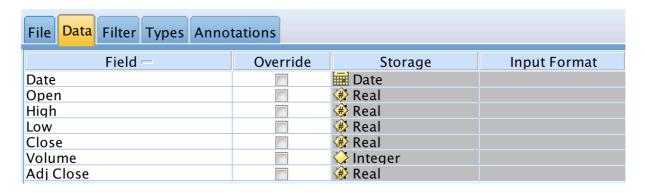


Third Google,

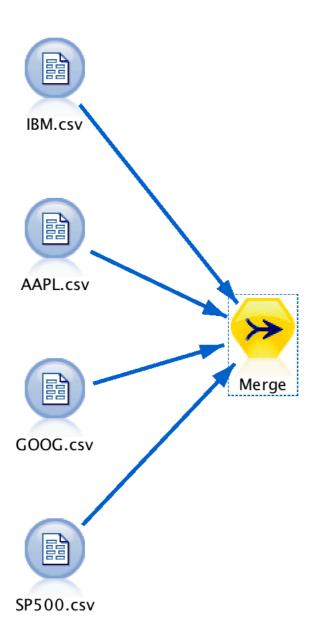




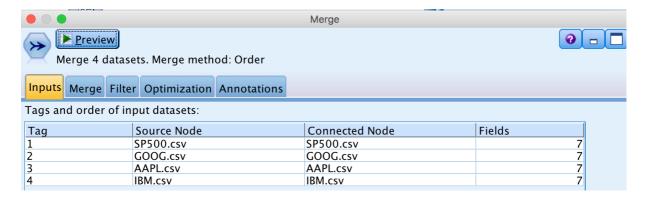
In each "Var. File" node in "Data" tab, you will see that field types are identified correctly



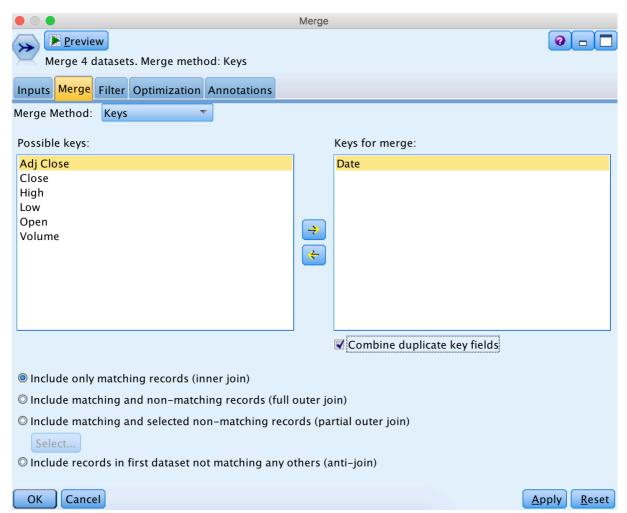
I can now merge these datasets into one by using "Merge" node from "Record Ops" palette.



Double click to open "Merge" node. In first tab, it shows you list of input files.

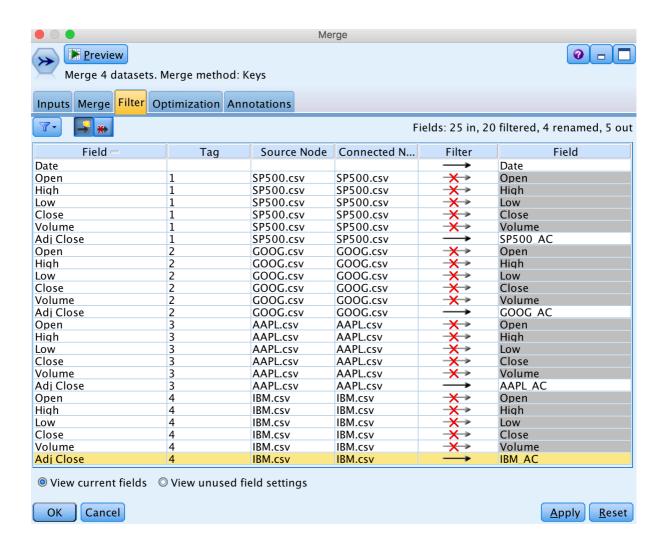


In "Merge" tab, we need to select "Key" for merging these datasets. "Key" should be common in all datasets and it will be used to match records and merge them. In our case, it's "Date" column.

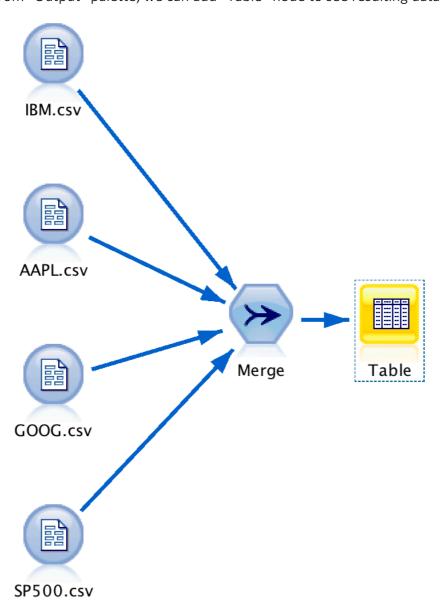


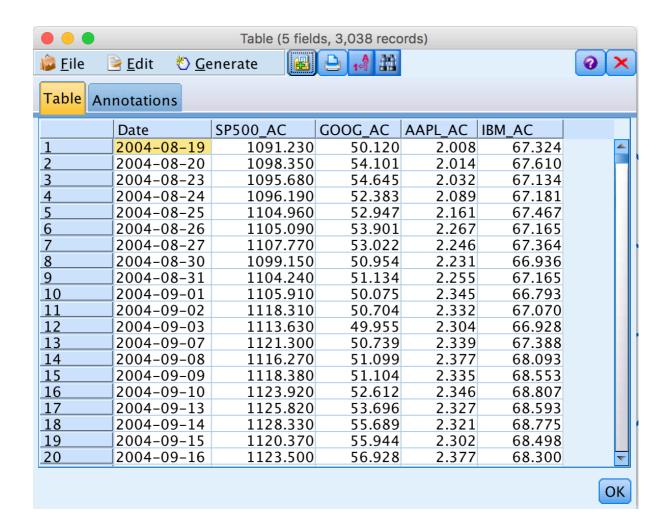
We will use default option "inner join" to merge datasets based on shorter dataset.

In "Filter" tab, we will filter out unnecessary columns and use suitable name, initials in this case, for "Adj Close" field coming from each dataset.



From "Output" palette, we can add "Table" node to see resulting dataset.

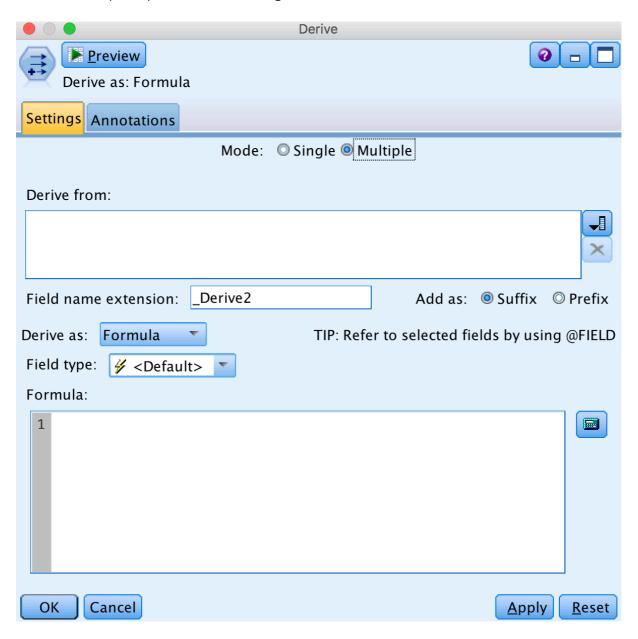




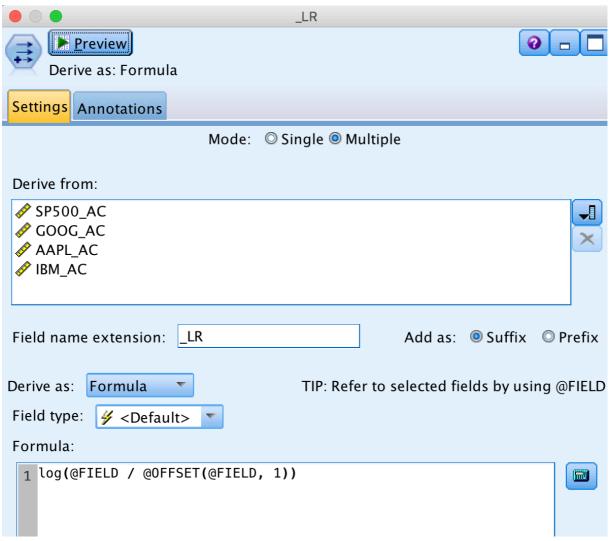
3. Calculating Log Return Series for All Datasets

We will add "Derive" node from "Field Ops" palette

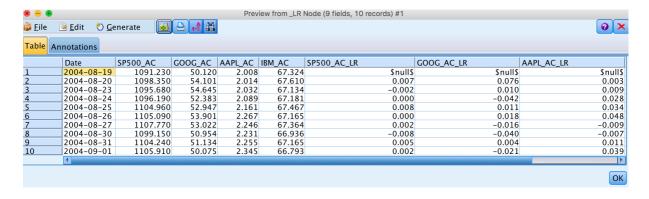
We use "Multiple" option to calculate log return series for all data sets at once.



We will add all series to "Derive from:" section. We will use "log" and "OFFSET" formula to calculate log returns and notice how did we use "@FIELD" to refer to all added fields. We will rename this new field as original field name plus suffix "_Log_Return"

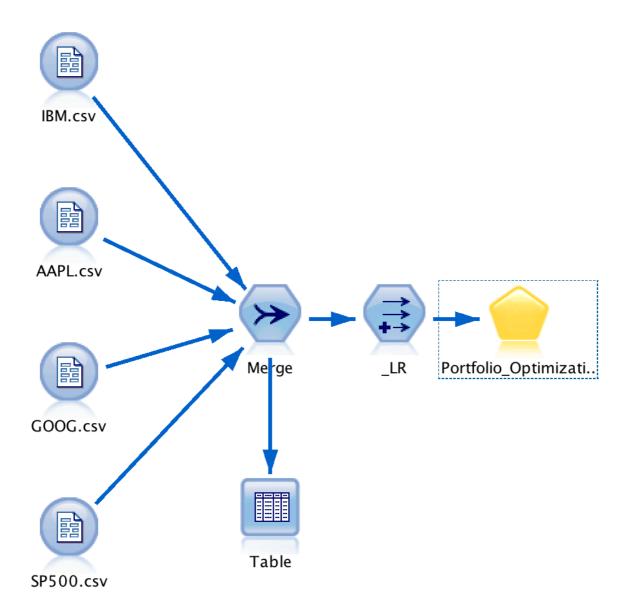


You can click preview to see resulting dataset.

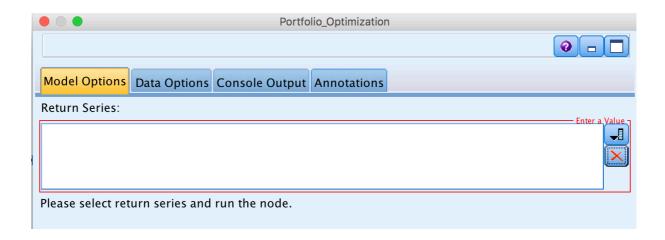


We will need to download following extension from this url (https://github.com/Umit-Mert/Portfolio_Optimization)

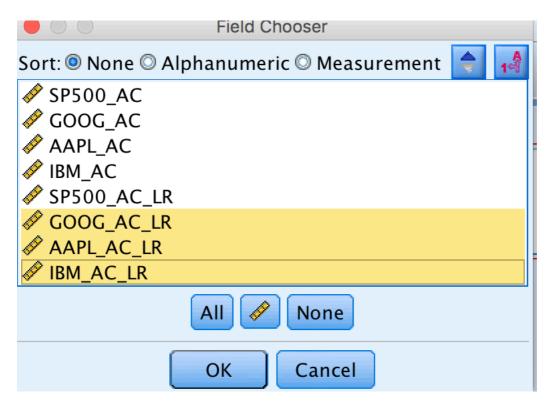
Once download and install finished. We need to add "Portfolio_Optimization" node from "Modeling" palette to our stream.



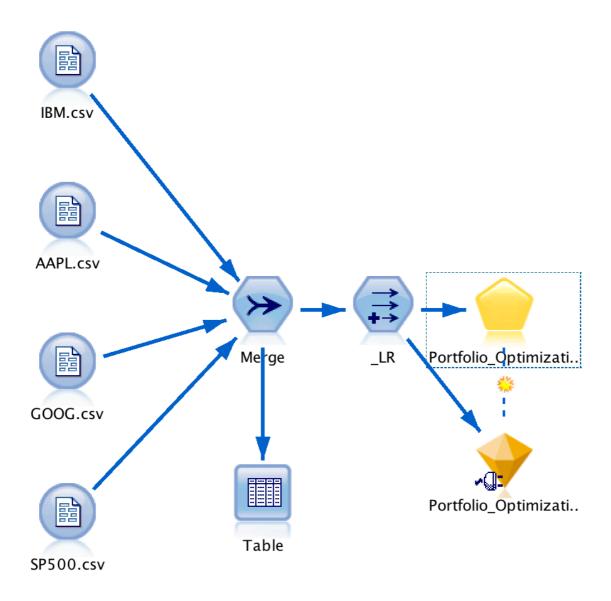
Double click to open that and select return series that we would like to work with.



In our case, return series are "GOOG", "AAPL" and "IBM".

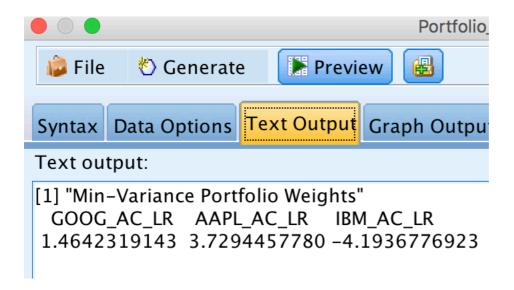


Let's click run and open the model nugget.

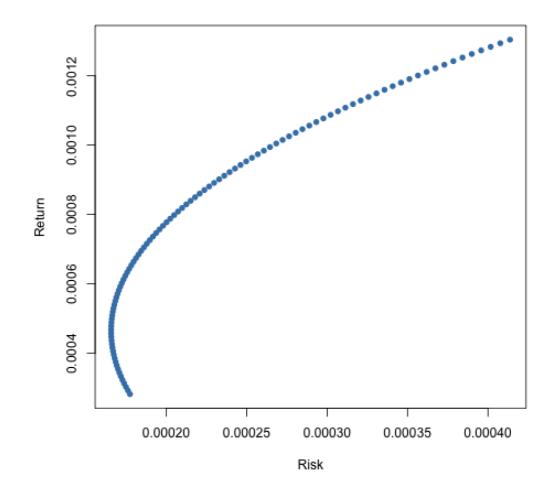


In model nugget, you should click "Text Output" tab to see the result.

We can see proportions in Portfolio weights section that we should invest in stocks "GOOG", "AAPL" and "IBM"



You can also click "Graph Output" tab to see Markowitz bullet.



Our portfolio weights are referring to minimum variance portfolio which is far left point closest to "Return" axis.

Summary

In this lab, you learned how to use "MPT" node to construct risk efficient portfolios. Thank you and hope to see you next time.