

AMS-512 Capital Markets and Portfolio Theory

Final Project

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Set-Up

Packages

```
In[ ]:= Get[FileNameJoin[{NotebookDirectory[], "QuadraticProgramming.m"}]]  
Get[FileNameJoin[{NotebookDirectory[], "MeanCovMissingMLE.m"}]]  
Get[FileNameJoin[{NotebookDirectory[], "FactorFitMLE.m"}]]
```

Calendar Functions

```
In[ ]:= xEndOfMonth[{y_, m_, d_}] := If[m == 12, {y, m, 31}, First@DayPlus[{y, m + 1, 1}, -1]];  
  
In[ ]:= xExpandCalendar[vxCalendar_, vxSeries_] := Sort[  
  Join[  
    vxSeries,  
    Transpose[{#, Array[Missing[] &, Length[#]]}] &[  
      Complement[vxCalendar, First /@ vxSeries]  
    ]  
  ]  
];
```

Candidate Dataset

```
In[ ]:= vsTickers =
  {"EEM", "EFA", "EWJ", "ICF", "IEF", "IEV", "IVV", "RWR", "SHY", "TIP", "TLT", "VTI"}
```

```
Out[ ]:= {EEM, EFA, EWJ, ICF, IEF, IEV, IVV, RWR, SHY, TIP, TLT, VTI}
```

```
In[ ]:= vsNames = FinancialData[#, "Name"] & /@ vsTickers;
```

```
Out[ ]:= $Aborted
```

```
In[ ]:= TableForm[{vsTickers, vsNames}^T, TableHeadings -> {Range[Length[vsTickers]]}]
```

```
Out[ ]:= TableForm=
```

1	EEM	iShares MSCI Emerging Markets ETF
2	EFA	iShares MSCI EAFE ETF
3	EWJ	iShares MSCI Japan ETF
4	ICF	ICF International Inc
5	IEF	iShares 7-10 Year Treasury
6	IEV	iShares Europe ETF
7	IVV	iShares Core S&P 500 ETF
8	RWR	SPDR Dow Jones REIT ETF
9	SHY	iShares 1-3 Year Treasury
10	TIP	iShares TIPS
11	TLT	iShares 20 Year Treasury
12	VTI	Vanguard Total Stock Market ETF

```
In[ ]:= mnDateRange = {{2005, 12, 15}, {2020, 04, 30}};
```

```
In[ ]:= buffer = FinancialData[#, "Close", mnDateRange, Method -> "Legacy"] & /@ vsTickers;
```

```
In[ ]:= Dimensions /@ buffer
```

```
Out[ ]:= {{3618, 2}, {3615, 2}, {3604, 2}, {3416, 2}, {3599, 2}, {3606, 2},
  {3604, 2}, {3604, 2}, {3600, 2}, {3598, 2}, {3600, 2}, {3604, 2}}
```

```
In[ ]:= And@@ (FreeQ[#, _Missing] & /@ buffer)
```

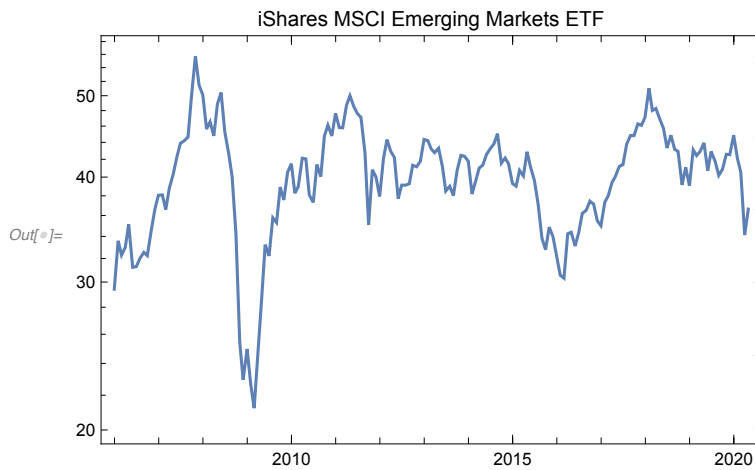
```
Out[ ]:= True
```

```
In[ ]:= monthly = (Last /@ Split[#, (#1[[1, 2]] == #2[[1, 2]]) &]) & /@ buffer;
```

```
In[ ]:= Dimensions /@ monthly
```

```
Out[ ]:= {{173, 2}, {173, 2}, {173, 2}, {164, 2}, {173, 2}, {173, 2},
  {173, 2}, {173, 2}, {173, 2}, {173, 2}, {173, 2}, {173, 2}}
```

```
In[ ]:= DateListLogPlot[monthly[[1]], PlotLabel -> vsNames[[1]]]
```



```
In[ ]:= vmxReturns = {xEndOfMonth /@ Rest[First /@ #],  $\left(\frac{\text{Rest}[\#]}{\text{Most}[\#]} - 1 \& \right) [\text{Last} /@ \#] \}^T \& /@ \text{monthly};$ 
```

```
In[ ]:= Dimensions /@ vmxReturns
```

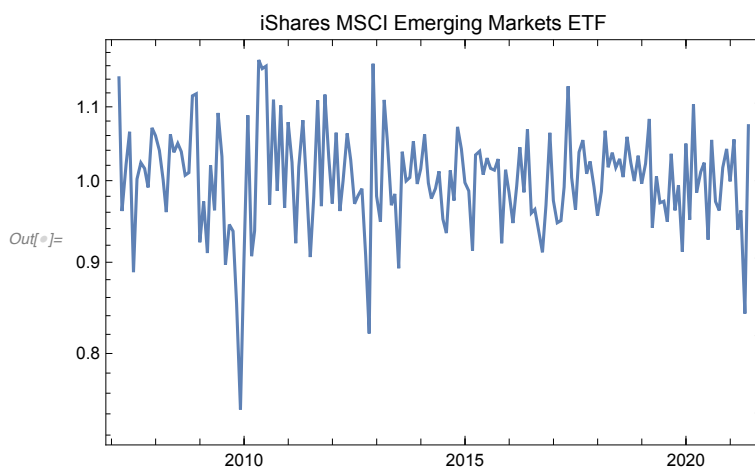
```
Out[ ]:= {{172, 2}, {172, 2}, {172, 2}, {163, 2}, {172, 2}, {172, 2},  
          {172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2}}
```

```
In[ ]:= vxCalendar = Union[Flatten[#[[All, 1]] & /@ vmxReturns, 1]];
```

```
In[ ]:= Length@vxCalendar
```

```
Out[ ]:= 172
```

```
In[ ]:= DateListLogPlot[1 + vmxReturns[[1]], PlotLabel -> vsNames[[1]]]
```



```
In[ ]:= vmxRepaired = xExpandCalendar[vxCalendar, #] & /@ vmxReturns;
```

```
In[ ]:= Dimensions /@ vmxRepaired
```

```
Out[ ]:= {{172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2},  
          {172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2}, {172, 2}}
```

```

In[ ]:= dbReturns = {
    vxCalendar,
    vsTickers,
    Transpose[#[[All, 2]] & /@ vmxRepaired]
};

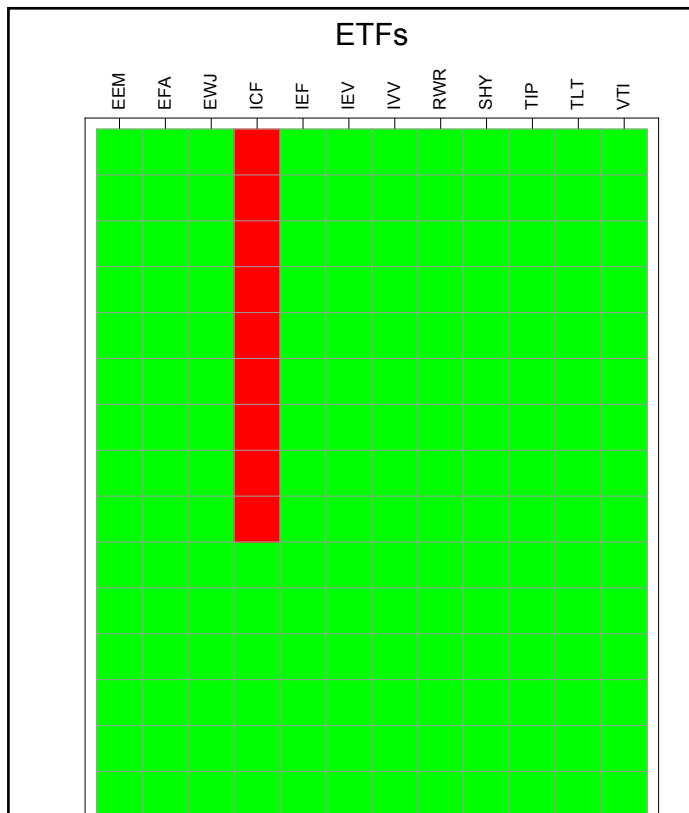
In[ ]:= Export[FileNameJoin[{NotebookDirectory[], "dbReturns.m"}], dbReturns]

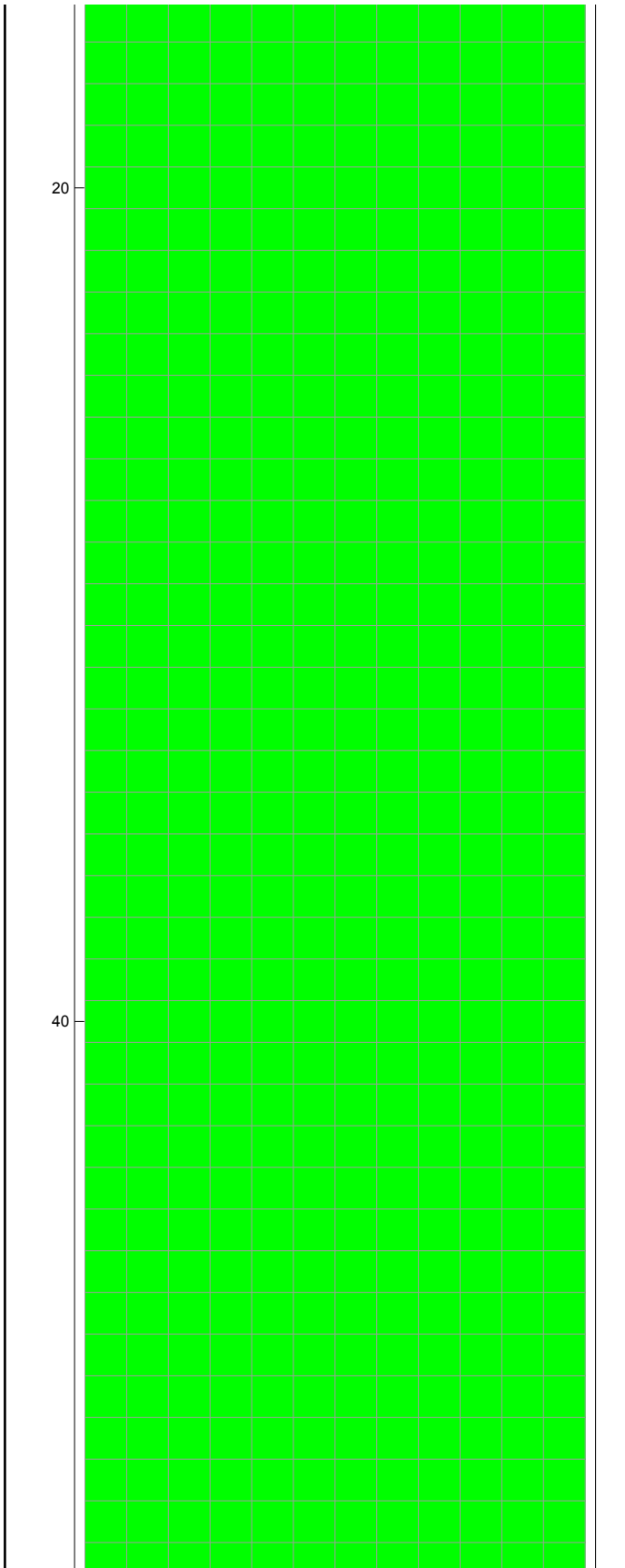
```

Missing Data Analysis

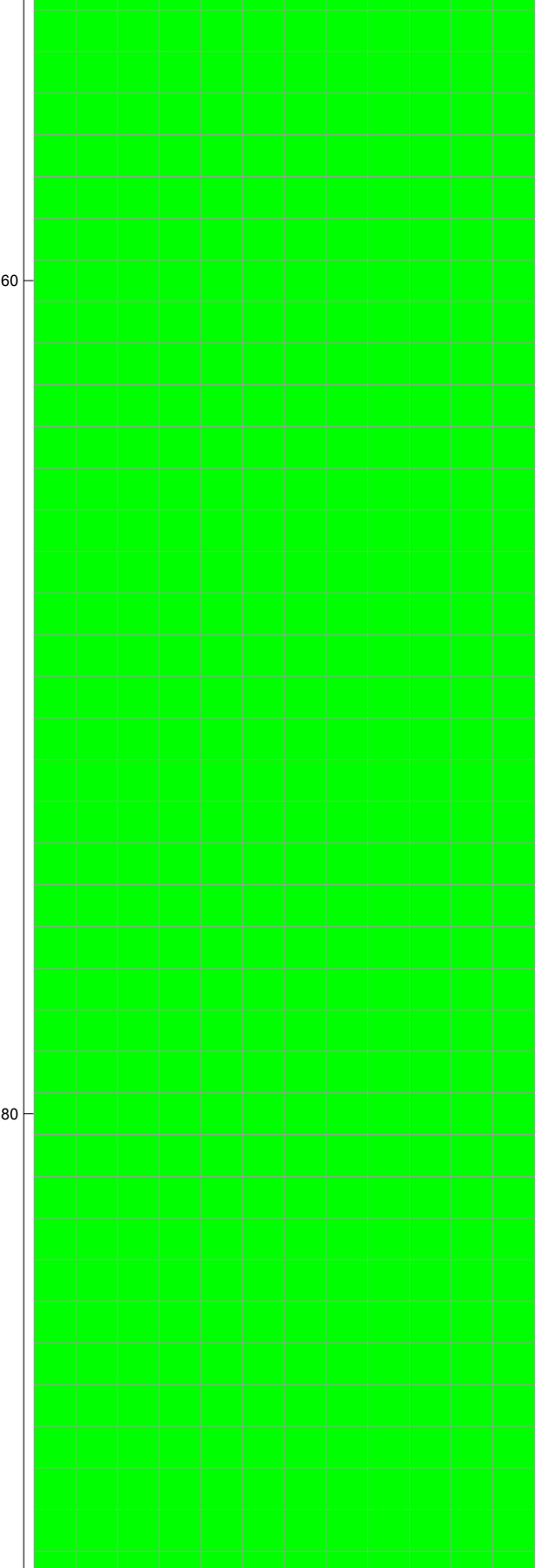
Missing Data Pattern

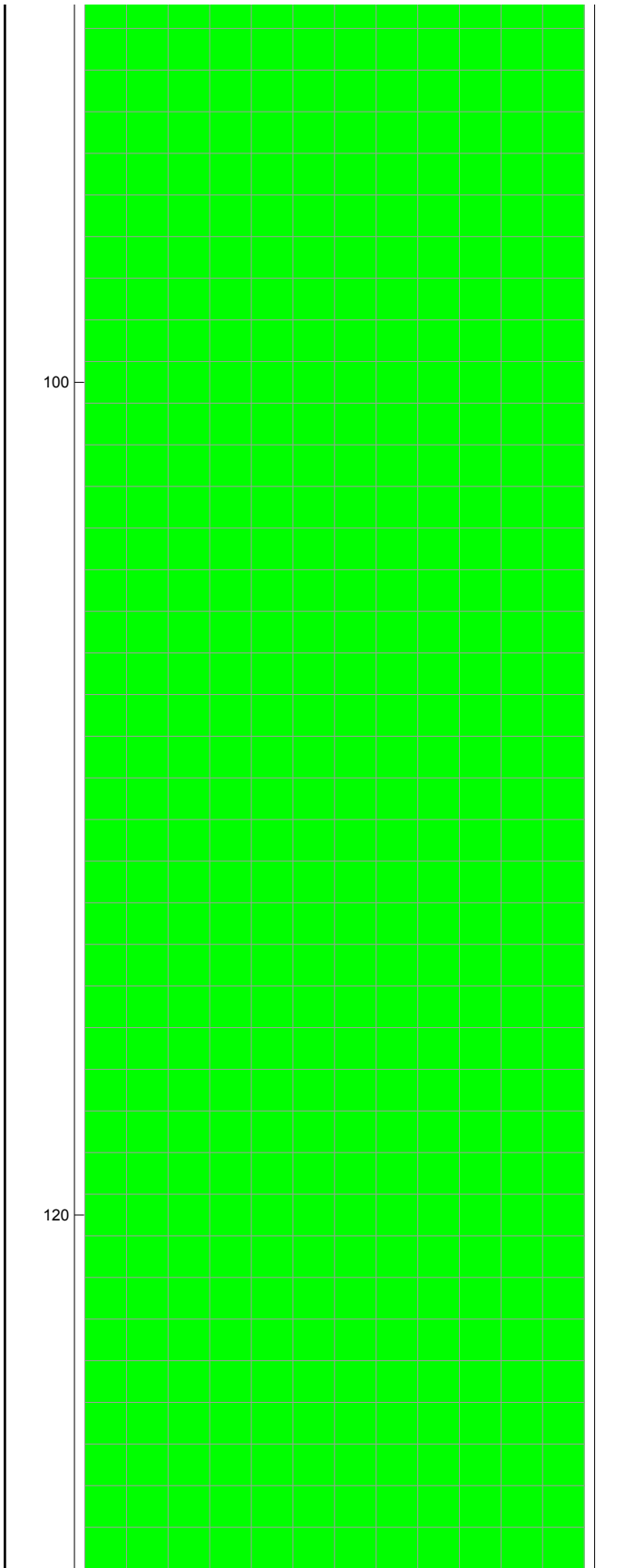
```
In[ ]:= Framed@ArrayPlot[
  Map[Boole@FreeQ[#, _Missing] &, dbReturns[[3]], {2}],
  ImageSize -> 350,
  Mesh -> True,
  ColorFunction -> (If[# == 1, Green, Red] &),
  Frame -> True,
  FrameLabel -> {{ "variable", Style["ETFs", FontSize -> 16]}, {"observation", ""}},
  FrameTicks -> {{Rest@Range[0, Length[dbReturns[[1]], 20], None],
    {None, {Range[Length[#, ], #]^T & [Rotate[#, 90 Degree] & /@ dbReturns[[2]]]}}
```

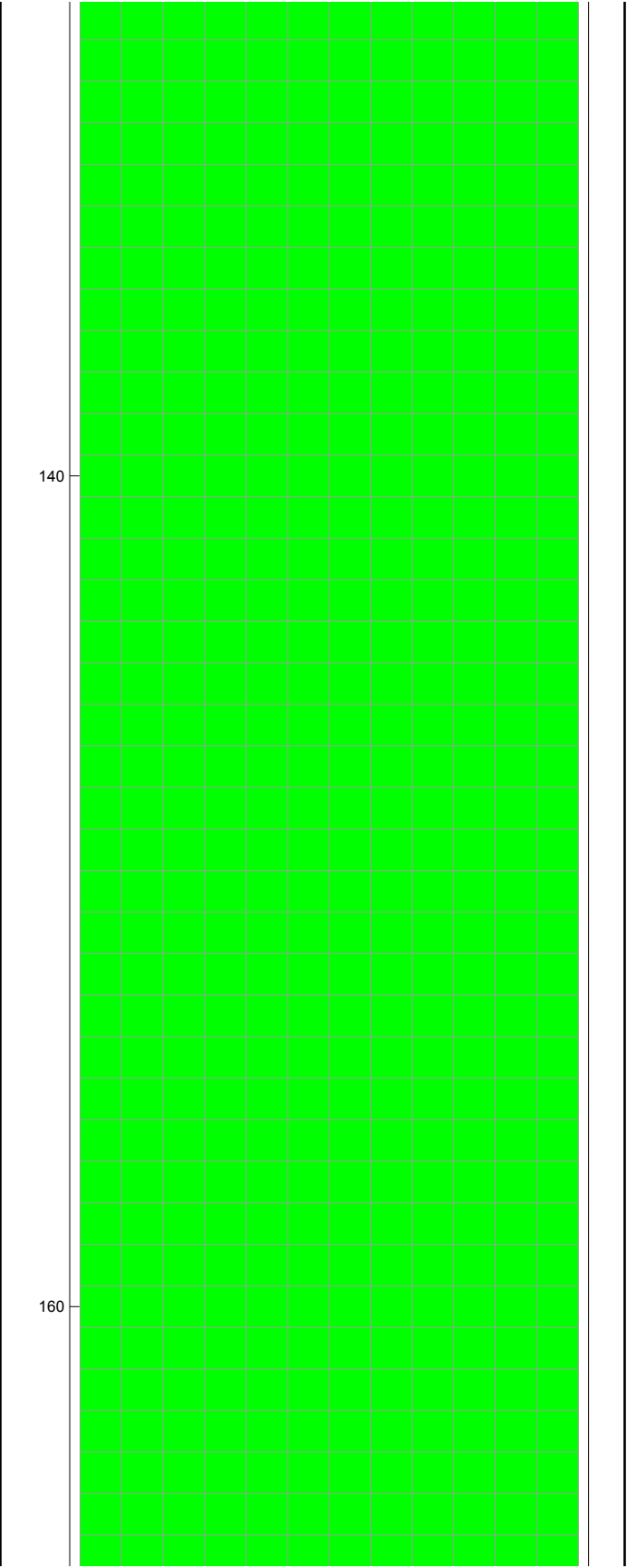


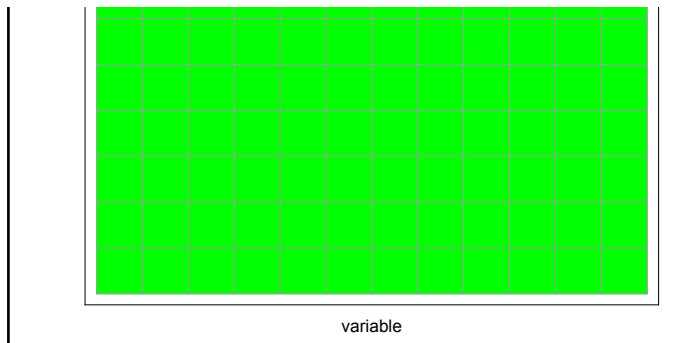


$Out[\#] =$
observation









Mean and Covariance Estimation

In[]:= ? xMeanCovMissingMLE

Symbol

{{Mean, Cov}, {Completed, Cross}, LLHistory, RunID} =
 xMeanCovMissingMLE[dbReferenceReturns, opts] – MLE estimate of mean and cov with missing data.
 Uses XEM–algorithm (extrapolation, expectation, minimization).

dbReferenceReturns – db of reference returns; missing values indicated by Missing[[]].

Option names are strings:

"InitialEstimate" – "Diagonal" (default), "CompleteObs", and "ImputedMean" or {Mean, Cov}.

If a string, then defines how the initial estimate is produced;

otherwise, initial starting values for the mean vector and covariance matrix.

"ToleranceGoal" – Rate of change in log likelihood, criterion for EM termination, default 10^{-6} .

"MaxIterations" – Max number of iterations, criterion for EM termination, default 100.

"Extrapolation" – True to use the X–Step (default); False to use standard EM algorithm.

"CheckPoint" – Frequency at which state sufficient for restart saved; 0 (default) for no check–pointing.

"Verbose" – True to print info on each iteration, default False.

Mean – Mean vector

Cov – Covariance matrix

Completed – Data with missing values replaced by expectations based on regression estimates.

Cross – Matrix for second order correction of cross products involving missing data.

LLHistory – History of log likelihood at each EM iteration.

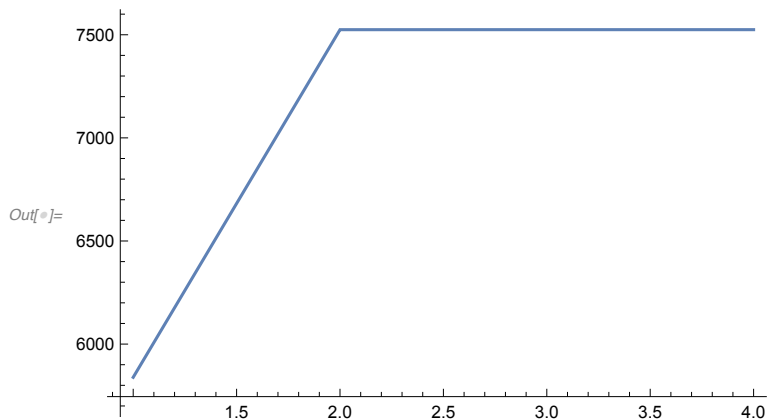
RunID – Unique string representing the environment and time stamp of the run.

In[]:= {{vnMean, mnCovariance}, {mnCompleted, mnCross}, vnLogLik, sRunID} =
 xMeanCovMissingMLE[dbReturns];

In[]:= sRunID

Out[]:= robertjfrey_MacOSX-x86-64_0_20200430_20200504025716

```
In[ ]:= ListPlot[vnLogLik, Joined → True, PlotRange → All]
```



```
In[ ]:= TableForm[{vsTickers, vsNames, vnMean, Mean[Last /@ #] & /@ vmxReturns]^T,
  TableHeadings → {Automatic, {"Tickers", "Names", "Imputed  $\mu$ ", "Observed  $\mu$ "}}
]
```

Out[]//TableForm=

	Tickers	Names	Imputed μ	Observed μ
1	EEM	iShares MSCI Emerging Markets ETF	0.00341402	0.00341402
2	EFA	iShares MSCI EAFE ETF	0.00104696	0.00104696
3	EWJ	iShares MSCI Japan ETF	0.000720955	0.000720955
4	ICF	ICF International Inc	0.0152929	0.0149707
5	IEF	iShares 7-10 Year Treasury	0.00232153	0.00232153
6	IEV	iShares Europe ETF	0.00111804	0.00111804
7	IVV	iShares Core S&P 500 ETF	0.00588997	0.00588997
8	RWR	SPDR Dow Jones REIT ETF	0.00327475	0.00327475
9	SHY	iShares 1-3 Year Treasury	0.000465615	0.000465615
10	TIP	iShares TIPS	0.00110197	0.00110197
11	TLT	iShares 20 Year Treasury	0.00420936	0.00420936
12	VTI	Vanguard Total Stock Market ETF	0.00603939	0.00603939

```
In[ ]:= TableForm[{vsTickers, vsNames,  $\sqrt{\text{Diagonal}[\text{mnCovariance}]}$ ,
  StandardDeviation[Last /@ #] & /@ vmxReturns]^T,
  TableHeadings → {None, {"Tickers", "Names", "Imputed  $\sigma$ ", "Observed  $\sigma$ "}}
]
```

Out[]//TableForm=

	Tickers	Names	Imputed σ	Observed σ
	EEM	iShares MSCI Emerging Markets ETF	0.0648916	0.0650811
	EFA	iShares MSCI EAFE ETF	0.0509951	0.051144
	EWJ	iShares MSCI Japan ETF	0.04377	0.0438978
	ICF	ICF International Inc	0.0918093	0.0922028
	IEF	iShares 7-10 Year Treasury	0.0178814	0.0179336
	IEV	iShares Europe ETF	0.0556086	0.0557709
	IVV	iShares Core S&P 500 ETF	0.0431052	0.043231
	RWR	SPDR Dow Jones REIT ETF	0.0682853	0.0684847
	SHY	iShares 1-3 Year Treasury	0.00359401	0.0036045
	TIP	iShares TIPS	0.0166922	0.016741
	TLT	iShares 20 Year Treasury	0.0387789	0.0388922
	VTI	Vanguard Total Stock Market ETF	0.0447602	0.0448909

Efficient Portfolios and Frontier with Correlation Denoising

Denoising the Correlation Matrix

Restate Covariance Using Denoised Correlation

Factor Model (Denoised Covariance)

Portfolio Optimization

Mean-Variance Efficient Frontier (Using Factor Model Covariance)

Student t Fit of Returns of Efficient Portfolios

Compute Expected Shortfall (CVaR) at 99% Confidence Level

Recommended approach is to write a function

```
xComputeCVaR[ret_, port_, conf_]
```

where `ret` is the return matrix, `port` the portfolio allocations and `conf` the confidence level.

This function then computes the portfolio returns, fits a Student t distribution to them, and estimates a CVaR. In the form

```
xComputeCVaR[dbReturns[[3]], #, 0.99]&
```

it can be mapped over the portfolios in the efficient frontier to produce a vector of CVaRs.

Plot Mean-Variance-CVaR Frontier