Applied Quantitative Investment Management

Lecture 3: Investment Simulation Framework

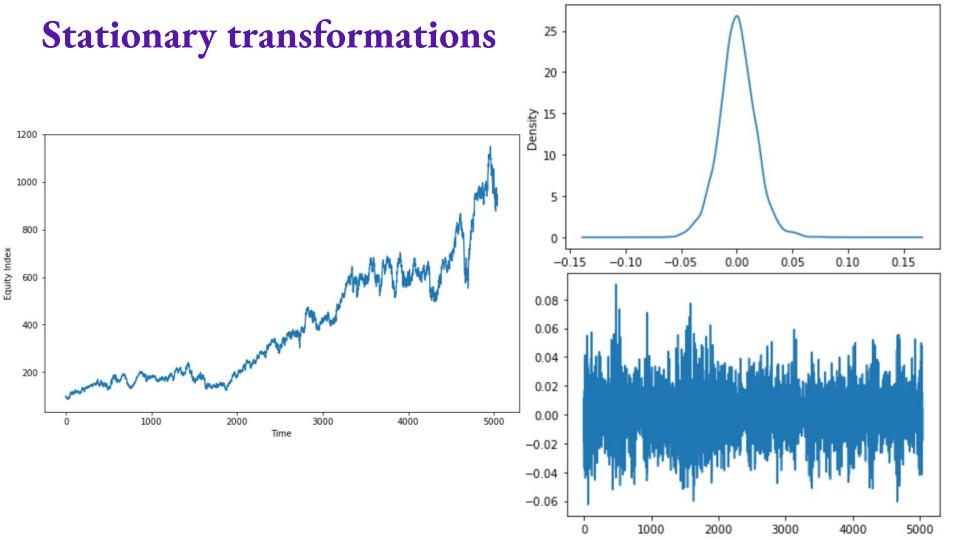
Anton Vorobets

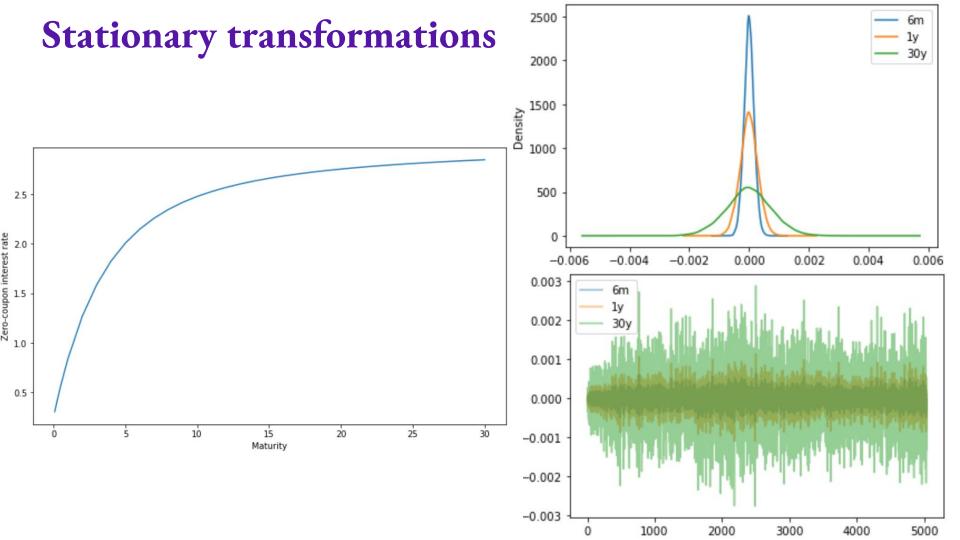
Agenda

- Overview of the multi-asset investment simulation framework
- Stationary transformations (Section 3.1)
- Computing simulated risk factors (Section 3.3)
- Simulation evaluation (Section 3.4)

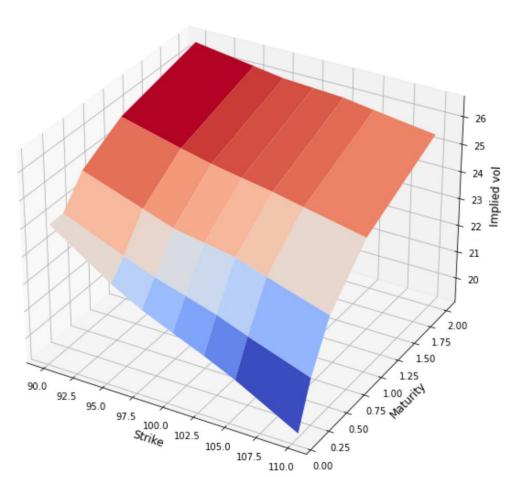
Next lecture:

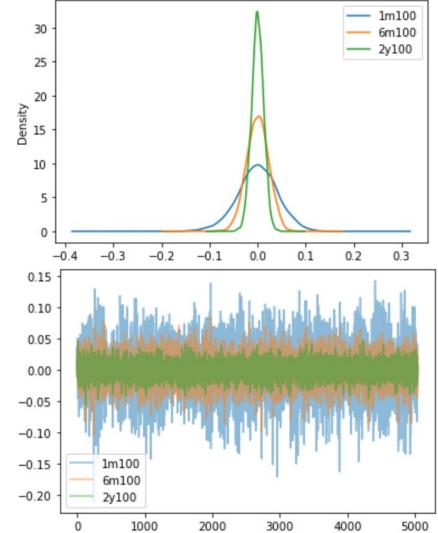
- Projection of stationary transformations (Sections 1.2 and 3.2)
- Better backtesting of CVaR versus variance optimization (Section 3.5)



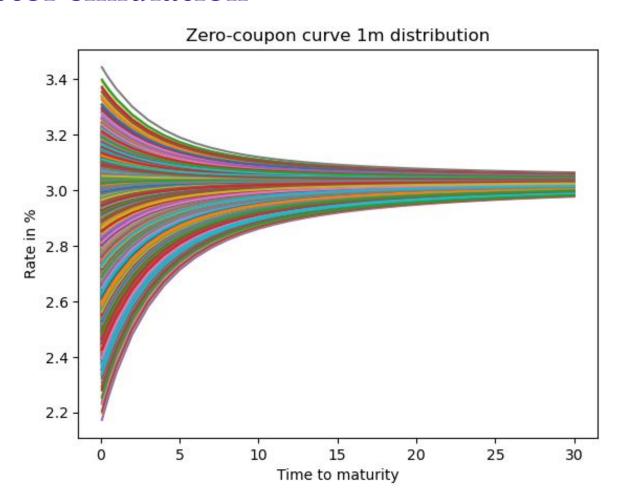


Stationary transformations

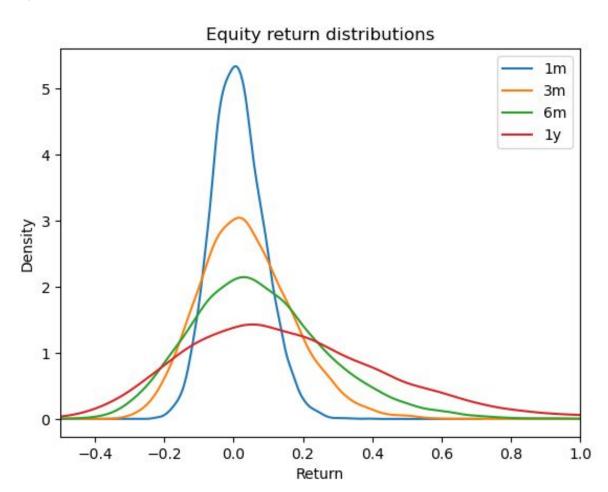




Risk factor simulation



P&L simulation



Market simulation fundamentals

Historical data:

$$D \in \mathbb{R}^{T \times N}$$

• Stationary transformations:

$$D \mapsto ST \in \mathbb{R}^{\tilde{T} \times \tilde{N}}$$

• Simulated stationary transformations:

$$\tilde{ST} \in \mathbb{R}^{S \times \tilde{N} \times H}$$

Simulated data:

$$\tilde{ST} \mapsto \tilde{R} \in \mathbb{R}^{S \times N \times H} \qquad R_h \in \mathbb{R}^{S \times N}$$

Market data

$$D = \begin{pmatrix} D_{1,1} & D_{1,2} & \cdots & D_{1,N} \\ D_{2,1} & D_{2,2} & \cdots & D_{2,N} \\ \vdots & \vdots & \ddots & \vdots \\ D_{T,1} & D_{T,2} & \cdots & D_{T,N} \end{pmatrix} \in \mathbb{R}^{T \times N}$$

100,000000 0.306974 0.415789 0.568395 0.839313 1.269176 1.587154 1.825839 2.007783 2.148673 ... 3.466530 3.387351 3.322905 3.270851 3.228836 3.194803 3.167063 3.144275 3.125390

98.675317 0.315879 0.424345 0.576460 0.846504 1.274975 1.591918 1.829823 2.011171 2.151599 ... 3.487529 3.404820 3.337619 3.283400 3.239669 3.204264 3.175415 3.151722 3.132090 3.115674

916.604273 2.756630 2.768203 2.784355 2.812812 2.857425 2.890037 2.914321 2.932732 2.946936 ... 1.390928 1.657677 1.864124 2.025434 2.152782 2.254410 2.336413 2.403318 2.458506 2.504516

1.592355 1.830189 2.011482 2.151868 ... 3.405938 3.336940 3.280444 3.234637 3.197572 3.167498 3.142957 3.122781 3.106052

... 1.485389 1.736523 1.930706 2.082334 2.201983 2.297435 2.374437 2.437250 2.489056 2.532243

						•		:	•		:					
					$\setminus L$	$T_{.1}$	D	T.2	•••	D	T.N					
					`	- ,-		- ,-			- ,	,				
Equity Index	1m	3m	6m	1у	2у	3у	4y	5у	6у	cr1y	cr2y	cr3y	cr4y	cr5y	cr6y	

					$\setminus D$	T,1	D_{2}	T,2	•	D	T,N)			
Equity Index	1m	3m	6m	1y	2у	3у	4 y	5у	6у	cr1y	cr2y	cr3y	cr4y	cr5y	cr6y

5039 903.256326 2.795153 2.805179 2.819163 2.843773 2.882294 2.910409 2.931322 2.947167 2.959386

						:		:	•••		•				
					$\setminus L$	$O_{T,1}$	L	T,2	• • •	D	T,N				
Equity Index	1m	3m	6m	1у	2y	3у	4y	5у	6y	cr1y	cr2y	cr3y	cr4y	cr5y	cr6y

Stationary transformations

0.000010

0.000014

-0.000029

-0.000007

-0.000047

-0.000051

-0.000013

-0.000083

-0.000082

-0.000100

-0.000026

-0.001022

5034

5035

5036

5037

5038

$$ST = \begin{pmatrix} ST_{1,1} & ST_{1,2} & \cdots & ST_{1,N} \\ ST_{2,1} & ST_{2,2} & \cdots & ST_{2,N} \\ \vdots & \vdots & \ddots & \vdots \\ ST_{T,1} & ST_{T,2} & \cdots & ST_{T,N} \end{pmatrix} \in \mathbb{R}^{\tilde{T} \times \tilde{N}}$$

0.001666

-0.000325

-0.000353

0.002495

-0.000662

-0.000351

-0.000381

-0.000380

ST =	$ST_{2,1}$	$ST_{2,2}$	• • •	$ST_{2,N}$	$\subset \mathbb{T} \tilde{T} imes ilde{N}$
SI =	:	:	٠.	÷	$\in \mathbb{R}$
	$\setminus ST_{T,1}$	$ST_{T,2}$		$ST_{T,N}$	$\in \mathbb{R}^{ ilde{T} imes ilde{N}}$

Equity Index 1m 3m 6m 1y 2y 3y 4y 5y 6y cr1y cr2y cr3y cr4y cr5y cr6y	_			0.000001	0.000010			0.0004.10						0.0000.10		0.000500	0.000540		
		Equity Index	1m	3m	6m	1у	2у	3у	4y	5у	6у		cr1y	cr2y	cr3y	cr4y	cr5y	сгбу	
						1				$ST_{T,}$	2	• •	•	ST_T	r,N)			
								•		•		•		•					

0.000038

0.000054

-0.000147

-0.000112

-0.000029

-0.000183

-0.001137

		S	ST =		ST	T,1	S'	\vdots $T_{T,2}$	··.		$\vdots \ ST_{T,}$	N /	\in	\mathbb{R}^{1}	× IV	
ity Index	1m	3m	6m	1v	2v	3v	4v	5v	6v	cr1v	cr2v	cr3v	cr4v	cr5v	cr6v	

0.000057

-0.000156

-0.000119

-0.000030

-0.000194

-0.001206

-0.000162

-0.001249

-0.001589

... -0.000135

-0.000146

0.001035

-0.000225

-0.000285

-0.000309

-0.000581

-0.000670 -0.001118 -0.001416 -0.001614

Stationary transformation suggestions

• For indices, stocks, currencies, and implied volatilities, log changes:

$$ST_{t,n} = \ln D_{t,n} - \ln D_{t-1,n}$$

• For **interest rates and spreads** with maturity *m*, log changes in constant maturity zero-coupon bonds:

$$ST_{t,n} = m \left(D_{t-1,n} - D_{t,n} \right)$$

 "Filtering" of persistent time series with long-run values also a possibility, subsequently using the residuals for simulation

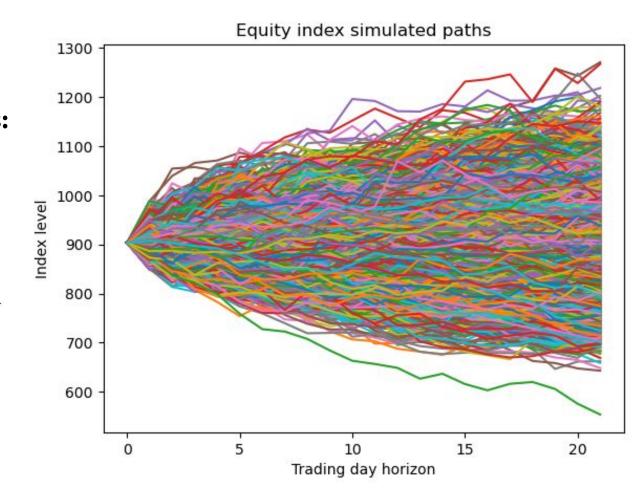
Simulated paths

Stationary transformations:

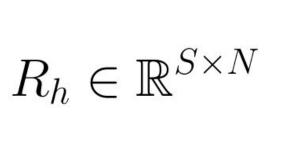
$$\tilde{ST} \in \mathbb{R}^{S \times \tilde{N} \times H}$$

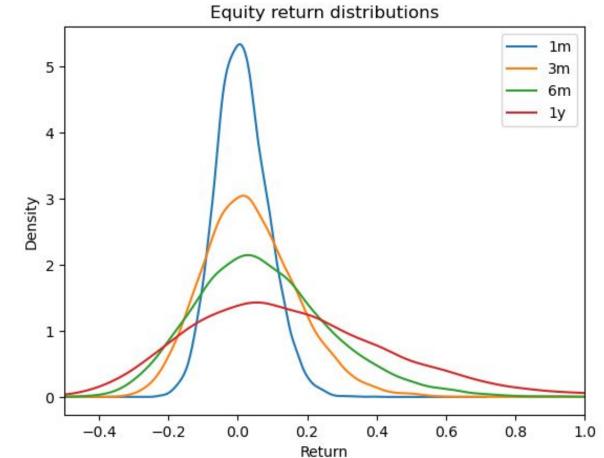
Risk factor paths:

$$\tilde{ST} \mapsto \tilde{R} \in \mathbb{R}^{S \times N \times H}$$



Specific horizons





Computing simulated risk factors and P&L

- "Inverse" of the stationary transformation
- Often involves cumulative sums of log change, for example, computing the discrete return on a one-month horizon using daily log changes
- For interest rates and credit spreads, it is important to first compute the initial discount factors. The cumulative sums of our log changes are in those
- See Chapter 3.3 Python code for all the inverse risk factors computations

Simulation evaluation

- Objective: simulate future paths that generalize out-of-sample and capture both the cross-sectional and time series dependencies
- **Challenge:** only one real-world observation for each time step to evaluate *S* cross-sectional simulations
- Reality: No single metric to assess if the simulation is "good" for fully general time series data
- **Proposal:** Use a mix of time series and cross-sectional metrics
- Sanity check: visualize market paths for prices, interest rates, portfolios, etc. to see if they fall within the simulated bands

Simulation evaluation

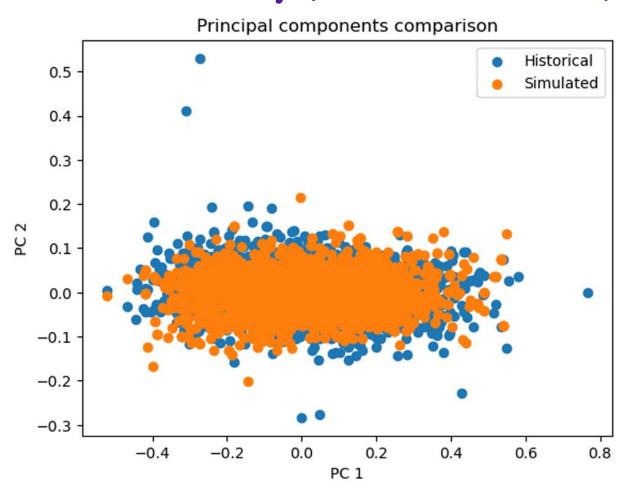
	Mean	Vol	Skew	Kurt
(0, Equity Index)	-0.009	0.048	-0.021	-0.092
(5, 2y)	-0.001	0.001	0.008	-0.031
(13, 10y)	0.000	0.000	0.008	-0.031
(33, 30y)	0.000	0.001	0.007	-0.031
(37, 1m100)	-0.003	-0.012	-0.049	0.118
(51, 6m100)	0.015	-0.008	0.018	0.065
(65, 2y100)	0.000	0.011	0.009	-0.004
(69, cr1y)	-0.001	0.003	-0.032	0.025
(73, cr5y)	-0.001	0.006	-0.031	0.027

Table 3.1: Difference between prior and simulation statistics using one state variable.

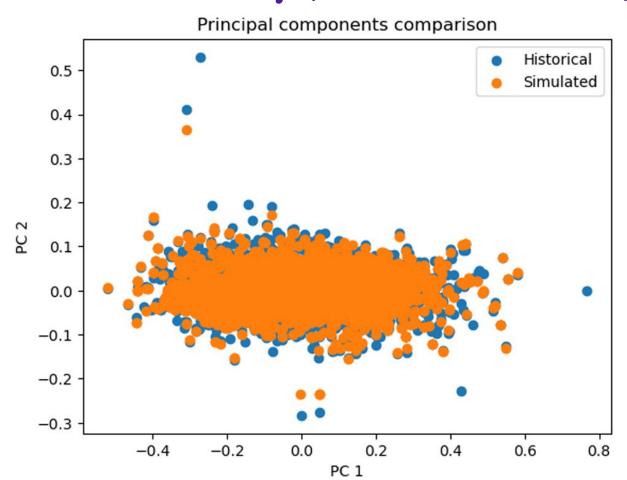
	Mean	Vol	Skew	Kurt
(0, Equity Index)	-0.004	0.013	-0.031	0.117
(5, 2y)	-0.001	0.001	-0.020	-0.008
(13, 10y)	-0.001	0.001	-0.019	-0.009
(33, 30y)	-0.001	0.001	-0.019	-0.009
(37, 1m100)	-0.022	-0.021	-0.021	-0.039
(51, 6m100)	-0.016	-0.002	0.024	-0.078
(65, 2y100)	-0.016	-0.011	0.036	-0.135
(69, cr1y)	-0.001	0.001	-0.023	0.264
(73, cr5y)	-0.001	0.000	-0.023	0.252

Table 3.2: Difference between prior and simulation statistics using two state variables.

Cross-sectional summary (one state variable)



Cross-sectional summary (two state variables)



Summary and next lecture

- Overview of the multi-asset investment simulation framework for fully general multi-asset data
- Understanding of the relationship between stationary transformations and risk factors

Next lecture:

- A good idea to get an introduction to Entropy Pooling (pages 70-73)
- Projection of stationary transformations (Sections 1.2 and 3.2)
- Better backtesting of CVaR versus variance optimization (Section 3.5)