

## Example 2: portfolio modeling

Suppose that you're trying to construct a portfolio: that is, to decide how to allocate your wealth among  $D$  financial assets. Things you want might to track include:

- the expected value of your portfolio at some point in the future (e.g. when you retire).
- the variance of your portfolio's value at some point in the future.
- the probability of losing some specific amount of money (10K, 20% of total value, etc)
- some measure of “tail risk,” i.e. what a bad week/month/year might look like.

Key idea: **use the bootstrap to simulate portfolio performance.**

# Example 2: portfolio modeling

Notation:

- Let  $T$  be our investing horizon (e.g.  $T = 20$  days,  $T = 40$  years, etc), and let  $t$  index discrete time steps along the way.
- Let  $X_{t,j}$  be the value of asset  $j = 1, \dots, D$  at time period  $t$ .
- Let  $R_{t,j}$  be the *return* of asset  $j$  in period  $t$ , so that we have the following recursive update:

$$X_{t,j} = X_{t-1,j} \cdot (1 + R_{t,j})$$

# Example 2: portfolio modeling

Notation:

- A portfolio is a set of investment weights over assets:  $(w_{t1}, w_{t2}, \dots, w_{tD})$ . Note: these weights might be fixed, or they might change over time.
- The value of your portfolio is the weighted sum of the value of your assets:

$$W_t = \sum_{j=1}^D w_{t,j} X_{t,j}$$

## Example 2: portfolio modeling

We care about  $W_T$ : the random variable describing your terminal wealth after  $T$  investment periods.

Problem: this random variable is a super-complicated, nonlinear function of  $T \times D$  individual asset returns:

$$W_T = f(R) \quad \text{where} \quad R = \{R_{t,j} : t = 1, \dots, T; j = 1, \dots, D\}$$

## Example 2: portfolio modeling

If we knew the asset returns, we could evaluate this function recursively, starting with initial wealth  $W_0$  at time  $t = 0$  and sweeping through time  $t = T$ :

Starting with initial wealth  $X_{1,j}^{(i)}$  in each asset, we sweep through from  $t = 1$  to  $t = T$ :

$$X_{t,j}^{(f)} = X_{t,j}^{(i)} \cdot (1 + R_{t,j}) \quad (\text{Update each asset})$$

$$W_t = \sum_{j=1}^D w_{t,j} X_{t,j}^{(f)} \quad (\text{Sum over assets})$$

$$X_{t+1,j}^{(i)} = w_{t+1,j} \cdot W_t \quad (\text{Rebalance})$$

## Example 2: portfolio modeling

But of course, we don't know the asset returns! This suggests that we should use a Monte Carlo simulation, where we repeat the following `for` loop many times.

For  $t = 1, \dots, T$ :

1. Simulate  $R_t = (R_{t1}, R_{t2}, \dots, R_{tD})$  from the joint probability distribution of asset returns at time  $t$ .
2. Use these returns to update  $X_{j,t}$ , the value of your holdings in each asset at step  $t$ .
3. Rebalance your portfolio to the target allocation.

The precise math of the update and rebalance steps are on the previous slide.

## Example 2: portfolio modeling

The difficult step here is (I): simulate a high-dimensional vector of asset returns from its joint probability distribution.

- very complicated correlation structure
- probably not something simple like a Gaussian!

In general, using simple parametric probability models (e.g. multivariate Gaussian) to describe high-dimensional joint distributions is a very dicey proposition.

# A simple approach: bootstrap resampling

Suppose we have  $M$  past samples of the asset returns, stacked in a matrix:

$$R = \begin{pmatrix} R_{11} & R_{12} & \cdots & R_{1D} \\ R_{21} & R_{22} & \cdots & R_{2D} \\ \vdots & & & \\ R_{M1} & R_{M2} & \cdots & R_{MD} \end{pmatrix}$$

where  $R_{tj}$  is the return of asset  $j$  in period  $t$ .



# A simple approach: bootstrap resampling

The key idea of bootstrap resampling is the following:

- We may not be able to describe what the joint distribution  $P(R_1, \dots, R_D)$  is.
- But we *do know that every row of this  $R$  matrix is a sample from this joint distribution.*
- So instead of sampling from some theoretical joint distribution, we will sample from the sample—i.e. we will bootstrap the past data.
- Thus every time we need a new draw from the joint distribution  $P(R_1, \dots, R_D)$ , we randomly sample (with replacement) a single row of  $R$ .

# A simple approach: bootstrap resampling

Thus our Monte Carlo simulation looks like the following at each draw.

For  $t = 1, \dots, T$ :

1. Simulate  $R_t = (R_{t1}, R_{t2}, \dots, R_{tD})$  by drawing a whole row, with replacement, from our matrix of past returns.
2. Use these returns to update  $X_{j,t}$ , the value of your holdings in each asset at step  $t$ .
3. Rebalance your portfolio to the target allocation.

# Example

Let's go to the R code! See `portfolio.R` on the website.

# Key discussion question

**Why do we draw an entire row of  $R$  at a time?**