### 6.1 The Horse Race

**Assumption:** Let m horses run in a race. Let the ith horse win with probability  $p_i$ .

### Payoff:

- If horse *i* wins, the payoff is  $o_i$  for 1.
- An investment of one dollar on horse *i* yields:

$$\begin{cases} o_i \text{ dollars,} & \text{if horse } i \text{ wins,} \\ 0 \text{ dollars,} & \text{if horse } i \text{ loses.} \end{cases}$$

• The gambler distributes **all** her money on the m possible bets:  $b_1, \ldots, b_m, \ b_i \geq 0, \ \sum_i b_i = 1$ 

# Some intuitive possibilities

- Put all of the money on the horse with the highest return.
- Put all of the money on the horse with the highest probability.
- Risky: most probable horse might still lose.
- Better to hedge.

### Gambler's Wealth After n Races

Let  $S_n$  be the gambler's wealth after n races. Then

$$S_n = \prod_{i=1}^n S(X_i),$$

where S(X) = b(X) o(X) is the factor by which the gambler's wealth is multiplied when horse X wins.

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# Constant rebalanced porfolios

- Use a fixed distribution of the stocks  $\vec{b}$
- Not the same as buy and hold.
- Example:  $\vec{b} = (1/2, 1/2)$

**Iter:** 1,2,3,4,5,...

**Cash:** 1,1,1,1,1,...

**Stock:** 1,2,1,2,1,...

• Wealth:  $1, \frac{3}{2}, \frac{3}{4} + \frac{3}{4} \frac{1}{2} = \frac{9}{8}, \frac{9}{8} \frac{3}{2}, \dots$ 

• Wealth increases by a factor of  $\frac{9}{8}$  every two iterations.

# Kelly's rule

- Suppose that the returns are drawn from a fixed and known distribution.
- The optimal strategy, in terms of rate of increase of log wealth, is a constant rebalanced portfolio.
- But which portfolio?

### What Are Universal Portfolios?

- **Concept:** Introduced by Thomas M. Cover, a universal portfolio is an investment strategy that asymptotically achieves the same growth rate of wealth as the best rebalanced portfolio in hindsight—without knowing the future in advance.
- **Key Idea:** Rather than fix a single strategy, the universal portfolio effectively averages over all possible rebalanced portfolios and updates weights based on observed performance.
- Goal: Leverage the law of large numbers—type property so that, over time, the universal portfolio tracks the growth rate of the best static rebalancing strategy.

# Model Setup

- Consider a market with *m* assets (e.g. stocks).
- Trading occurs in **discrete time**: t = 1, 2, ..., n.
- Let  $x_t \in \mathbb{R}^m$  be the gross returns of the m assets between time t-1 and t.
  - For example, if the *j*-th asset goes up by 2%, then  $x_t^{(j)} = 1.02$ .
- A portfolio vector  $b_t \in \mathbb{R}^m$  specifies how one's capital is allocated among the m assets at time t.
  - The entries of  $b_t$  sum to 1 and are nonnegative:  $\sum_{j=1}^m b_t^{(j)} = 1$ ,  $b_t^{(j)} \geq 0$ .

### Wealth Evolution

- Let  $S_t$  denote the wealth at time t.
- Given a portfolio  $b_t$  at time t, the wealth is updated by

$$S_{t+1} = S_t \left( b_t \cdot x_{t+1} \right),$$

where  $b_t \cdot x_{t+1}$  is the dot product of  $b_t$  and  $x_{t+1}$ .

• The goal is to choose  $\{b_t\}_{t=1}^n$  to maximize the final wealth  $S_{n+1}$  or equivalently  $\log S_{n+1}$ .

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# Cover's Universal Portfolio (Informal Definition)

- Consider all constant-rebalanced portfolios. A constant-rebalanced portfolio (CRP) is one that keeps the same fraction in each asset at every time step.
- ② Assign a prior. Treat each CRP as an element in the simplex of possible weights  $b \in \Delta^m$ . Typically, one uses the *Dirichlet* (or uniform) prior over the simplex.
- Update posterior. After each period, update this distribution (the "mixture") over all CRPs based on how well each CRP performed.
- **The second of the second of**

$$b_{t+1} = \int b \, d\mu_t(b),$$

where  $\mu_t$  is the posterior over the simplex after observing t periods.

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# Key Result

- **Asymptotic Optimality:** Cover showed that the growth rate of the *universal portfolio* will, in the limit, approach the growth rate of the best *single* constant-rebalanced portfolio in hindsight.
- Formally, let  $b^*$  be the CRP that maximizes  $\log S_n$  in hindsight. Then the ratio of the universal portfolio's wealth  $U_n$  to the wealth of  $b^*$  (both starting at 1) grows sub-exponentially in n.

$$\frac{U_n}{S_n(b^*)} \geq \exp(-o(n)) \quad \text{as } n \to \infty.$$

• This means that the universal portfolio is *universally* good, without prior knowledge of which CRP is best.

### **Practical Construction**

- Objective the Simplex:
  - In practice, the integral over all b in the simplex is approximated by a finite grid or sampling.
- Recompute Weights:

$$w_{t+1}(b) = \frac{w_t(b) \cdot (b \cdot x_{t+1})}{\int w_t(u) \cdot (u \cdot x_{t+1}) du}$$

where  $w_t(b)$  is the "weight" or "posterior" for CRP b.

Ompute New Investment:

$$b_{t+1} = \int b w_{t+1}(b) db \quad \approx \sum_{b \in \mathcal{B}} b w_{t+1}(b).$$

**Q** Rebalance Accordingly: Actually execute  $b_{t+1}$  in the market at time t+1.

# Simple Example (2 Assets)

- Suppose there are 2 assets, so  $b \in [0,1]$  with  $b_1 = b$ ,  $b_2 = 1 b$ .
- Uniform Prior: Start with  $w_0(b) = 1$  for  $b \in [0, 1]$ .
- **Observations:** If the asset returns over first period are  $(x_1, x_2)$ , then after seeing that outcome, the weight function updates:

$$w_1(b) \propto b x_1 + (1-b) x_2.$$

• **Next Step:** The universal strategy at t = 1 invests

$$b_1 = \int_0^1 b \, \frac{b \, x_1 + (1-b) \, x_2}{Z} \, db,$$

where Z is a normalization constant to ensure the posterior integrates to 1.

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### Numerical Illustration

- By discretizing  $b \in [0,1]$  into many small steps, you can numerically approximate the integrals.
- Over time, the algorithm will put more weight on the "best" fraction b\* that maximizes growth, but it still accounts for uncertainty and adapts as the environment changes.

# Summary

- Cover's Universal Portfolio is a powerful idea that uses mixture methods over all possible constant-rebalanced portfolios.
- Guarantees: It asymptotically matches the best constant-rebalanced strategy in hindsight.
- Implementation: Although conceptually elegant, the naive integral approach can be computationally expensive. Practical approximations are used (discretization, sampling, etc.).
- Significance: This method bridges information theory and portfolio choice, illustrating how "universal" strategies can learn from the market without predictions.

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#### References



T. M. Cover, Universal Portfolios, Math. Finance, Vol. 1, No. 1 (1991), pp. 1–29.



T. M. Cover and J. A. Thomas, *Elements of Information Theory*, Wiley, 2nd edition, 2006.



T. M. Cover, *Universal Portfolios*, in The Kelly Capital Growth Investment Criterion: Theory and Practice (2011), World Scientific.

### Blocks of Highlighted Text

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#### **Block**

Sample text

#### Alertblock

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### Examples

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