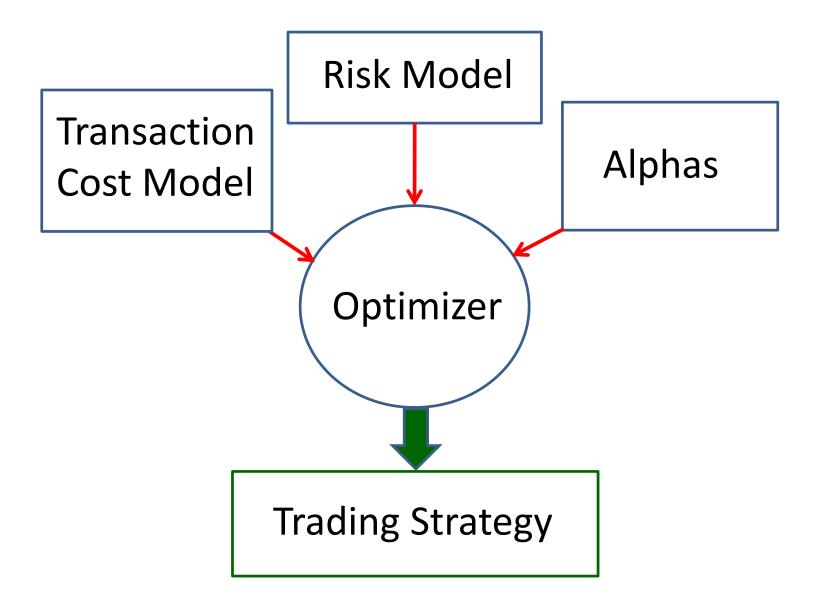
# MGMT 237M2 Statistical Arbitrage Lecture 07: Portfolio Optimization Professor Olivier Ledoit

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## **Overall Structure**



## What We've Seen So Far

 Risk Model: shrinkage estimator of the covariance matrix of stock returns

Transaction Cost Model: 1bp + ½ bid-ask spread

 Alpha: Weighted blend of various standardized, windsorized alphas

## Optimizer

#### Inputs:

- position as of close of business on day t-1
- alphas using data observed up to day t-1
- -t-costs using data observed up to day t-1
- -risk model using data observed up to day t-1
- constraints using data observed up to day t-1
- Output: trade to be executed on day t

```
final position(t+1) = final position(t) + trade (t)
```

## Timeline

• Day t-1: most recent available data

<u>Day t:</u> trade gets executed

• Day t+1: returns start to be earnt

### **Backtest Code**

- Load all necessary data into memory
- Create the alphas
- Start from portfolio with zero dollar invested
- Loop over all days in backtest period
  - Every day: call optimizer to find optimal rebalancing trade given initial position
  - End-of-day position becomes initial position of next day
- Compute P&L

## **Notation**

- x: (n × 1) vector of desired portfolio weights
- w: (n × 1) vector of initial portfolio weights
- $\Sigma$ : (n × n) covariance matrix of stock returns
- $\alpha$ : (n × 1) vector of aggregate alphas
- $\beta$ : (n × 1) vector of historical betas
- $\tau$ : (n × 1) vector of transaction costs

## Objectives and Constraints

- Minimize risk:  $x' \Sigma x$
- Maximize exposure to alpha:  $\alpha$  ' x
- Neutralize exposure to beta:  $\beta' x = 0$
- Minimize transaction costs: τ ' | x-w |
- Other constraints:
  - maximum trade size
  - maximum position size
  - maximum industry and country exposure

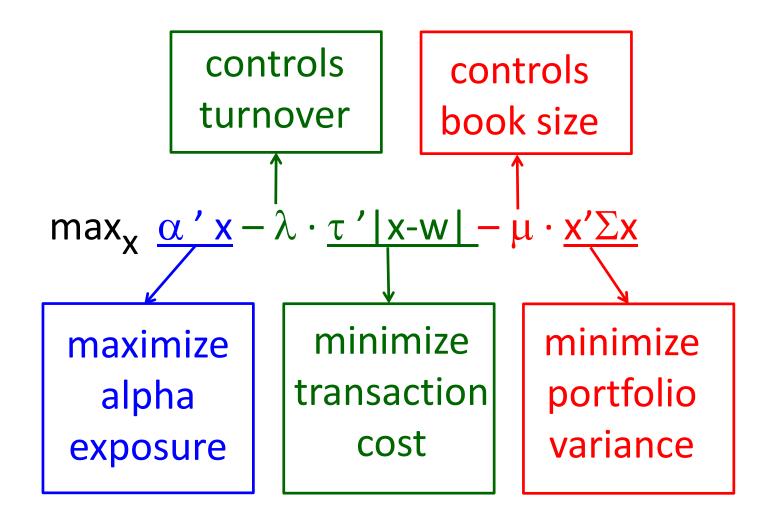
## **Optimization Problem**

$$\max_{\mathbf{X}} \alpha' \mathbf{x} - \lambda \cdot \tau' |\mathbf{x} - \mathbf{w}| - \mu \cdot \mathbf{x}' \Sigma \mathbf{x}$$
subject to:
$$\beta' \mathbf{x} = \mathbf{0}$$

and other constraints:

- maximum trade size
- maximum position size
- maximum industry exposure
- maximum country exposure

## **Objective Function**



## Maximum Trade Size

1% of Average Daily Volume (ADV)
 Can go to 2% if necessary (big book)

- Capped so liquid stocks do not dominate
- Example: cap at \$150K
   Can go higher if necessary (big book)

## Maximum Position Size

- Multiple of maximum trade size
- I want to be able to liquidate every position in how many days?
- 10 days  $\Rightarrow$  10 × max trade size
- Can be big relative to book size
- Keep balance between liquid and illiquid stocks

min(10 × max trade size, 2.5% of long side of book)

## Merge the 2 Constraints

• Max trade size for i<sup>th</sup> stock:  $\theta_i$ 

$$\Rightarrow$$
  $w_i - \theta_i \le x_i \le w_i + \theta_i$ 

• Max position size for i<sup>th</sup> stock:  $\pi_i$ 

$$\Rightarrow -\pi_{i} \leq x_{i} \leq \pi_{i}$$

Enforce both constraints simultaneously:

$$\max(w_{i} - \theta_{i}, -\pi_{i}) \leq x_{i} \leq \min(w_{i} + \theta_{i}, \pi_{i})$$

$$\gamma_{i} \leq x_{i} \leq \delta_{i}$$

## **Industry Constraints**

- Sectors are a factor of risk
- Difficult to time sector performance
- Constrain industry exposure
- But not to zero (too much transaction cost)
- For \$50 × 50M book size: r\* = \$300,000 limit

## **Industry Dummy**

- ρ industries
- Boolean matrix R of dimension (n  $\times \rho$ )
- R(i,j) = 1 if i<sup>th</sup> stock belongs to j<sup>th</sup> industry
- R(i,j) = 0 if i<sup>th</sup> stock is outside j<sup>th</sup> industry
- Every row of matrix R has exactly one entry equal to 1; all other entries are equal to 0
- Constraint:  $-r^* \cdot 1 \le R'x \le r^* \cdot 1$ where 1 = vector of ones of the right dimension

## **Country Constraints**

- Countries are a factor of risk
- Difficult to time country performance
- Constrain country exposure
- But not to zero (too much transaction cost)
- For \$50 × 50M book size: f\* = \$100,000 limit
- Tighter than industry exposure

## **Country Dummy**

- φ countries
- Boolean matrix F of dimension (n  $\times \phi$ )
- F(i,j) = 1 if i<sup>th</sup> stock belongs to j<sup>th</sup> country
- F(i,j) = 0 if i<sup>th</sup> stock does not belong to j<sup>th</sup> country
- Every row of matrix F has exactly one entry equal to 1; all other entries are equal to 0
- Constraint:  $-f^* \cdot 1 \le F'x \le f^* \cdot 1$

#### Overall Problem

$$\text{max}_{\textbf{X}} \ \alpha \ ' \ \textbf{x} - \lambda \cdot \tau \ ' \ | \ \textbf{x-w} \ | \ - \mu \cdot \textbf{x}' \Sigma \textbf{x}$$

#### Subject to:

- beta neutrality:  $\beta' x = 0$
- max trade and position:  $\gamma \le x \le \delta$
- industry constraint:  $-r^* \cdot 1 \le R'x \le r^* \cdot 1$
- country constraint:  $-f^* \cdot 1 \le F'x \le f^* \cdot 1$

Is this standard Quadratic Programming?

## **Quadratic Programming**

- Quadratic programming (QP) is fast, efficient and guaranteed to converge
- Excellent off-the-shelf software
- Matlab optimization toolbox

 Problem: the absolute value in the transaction cost term is not standard quadratic programming: τ '|x-w|

# Split Variables

- Classic solution: split each variable into 2
- Drawback: twice as many variables
- Advantage: no need to use nonlinear programming
- Define:
  - $\rightarrow$  y = max(x-w,0)
  - $\geq$  z = max(w-x,0)
- Then  $y \ge 0$ ,  $z \ge 0$ , x = w + y z and |x-w| = y+z

## Indeterminacy?

- Initial problem strictly convex
   ⇒ unique solution in x
- Twice as many variables: solution still unique in y and z?
- Replace y by y+1 and z by z+1
   ⇒ x = w + y z remains unchanged!

Still OK because |x-w| = y+z penalized

## **New Formulation**

```
\max_{y,z} \alpha'(w+y-z) - \lambda \cdot \tau'(y+z) - \mu \cdot (w+y-z)' \Sigma(w+y-z)
Subject to:
```

- beta neutrality:  $\beta'(w+y-z) = 0$
- max trade and position:  $\gamma \leq w+y-z \leq \delta$
- industry constraint:  $-r^* \cdot 1 \le R' (w+y-z) \le r^* \cdot 1$
- country constraint:  $-f^* \cdot 1 \le F'$  (w+y-z)  $\le f^* \cdot 1$

Very close to standard Quadratic Programming

# Standard Quadratic Programming

$$min_u 0.5 u' H u + g' u$$

#### Subject to:

- A u ≤ b
- C u = d
- LB  $\leq$  u  $\leq$  UB

## Rewrite Optimization Problem

$$\begin{aligned} \min_{\mathbf{y},\mathbf{z}} - \alpha'(\mathbf{y}-\mathbf{z}) + \lambda \cdot \tau'(\mathbf{y}+\mathbf{z}) + 2\mu \cdot \mathbf{w}' \Sigma(\mathbf{y}-\mathbf{z}) \\ + \mu \cdot (\mathbf{y}-\mathbf{z})' \Sigma(\mathbf{y}-\mathbf{z}) + \text{constant} \end{aligned}$$

#### Subject to:

- beta neutrality:  $\beta'(y-z) = -\beta'w$
- max trade and position:  $\gamma w \le y z \le \delta w$
- industries:  $-r^* \cdot 1 R'w \le R'(y-z) \le r^* \cdot 1 R'w$
- countries:  $-f^* \cdot 1 F'w \le F'(y-z) \le f^* \cdot 1 F'w$

Maps into standard Quadratic Programming

# Mapping Objective Function

• 
$$u = \begin{bmatrix} y \\ z \end{bmatrix}$$

• H = 2 
$$\mu$$
  $\begin{bmatrix} \Sigma & -\Sigma \\ -\Sigma & \Sigma \end{bmatrix}$ 

• g = 
$$\begin{bmatrix} 2\mu \Sigma w - \alpha + \lambda \tau \\ -2\mu \Sigma w + \alpha + \lambda \tau \end{bmatrix}$$

## Mapping Inequality Constraints

$$\bullet A = \begin{pmatrix} R' & -R' \\ -R' & R' \\ F' & -F' \\ -F' & F' \end{pmatrix}$$

# **Mapping Equality Constraints**

• 
$$C = \begin{bmatrix} \beta' & -\beta' \end{bmatrix}$$
 •  $d = -\beta' w$ 

# Bounds on Optimization Variables

Lower bound:
 LB = vector of zeros of dimension (2n × 1)

Upper bound:

UB = 
$$\left( \max(0, \min(\theta, \pi - w)) \right)$$
$$\max(0, \min(\theta, \pi + w))$$

## Matlab Quadratic Optimizer

- quadprog.m
- No starting point needed
- options = optimset('Algorithm','interior-pointconvex')
- options = optimset(options,'Display','iter')

[u,fval,exitflag,output] = quadprog(H,g,A,b,C,d,LB,UB,[],options)

## Other Good Optimizers

- IBM: CPLEX
- FICO: Xpress
- Sunset: XA
- Stanford: QPOPT, SQOPT and MINOS
- Roger Fletcher: BQPD
- KNITRO

Not cheap!

# Required Readings for Next Lecture

- Cristi A. Gleason and Charles M. C. Lee.
   Analyst forecast revisions and market price discovery. The Accounting Review, 78(1):pp. 193–225, 2003.
- Narasimhan Jegadeesh, Joonghyuk Kim, Susan D. Krische, and Charles M. C. Lee. Analyzing the analysts: When do recommendations add value? The Journal of Finance, 59(3):1083–1124, 2004.