

# Back to bonds: ① Convexity and Duration

② Swap Payoff =  $\sum_{i=1}^n \tau N(t) (T_{i-1}, T_i) - S_0) p(0, T_i)$

$P_{\text{Swap}} = \frac{1 - p(0, T_n)}{\tau \sum_{i=1}^n p(0, T_i)} \rightarrow p(0, T_n) + \sum_{i=1}^n \tau T_i p(0, T_i) = 1$

$P_{\text{Swap}} = N(1 - S_0 \tau \sum_{i=1}^n p(0, T_i) - p(0, T_n))$  净现值 = 面值 = 1

③ CMT:  $S_{t_0} = \frac{\sum p(t_0, t_i) f_{t_0}(t_{i-1}, t_i)}{\sum_{i=1}^n p(t_0, t_i)}$ , 远期利率加权  
 $\downarrow$   
 $"r(0, t_n)"$

①  $p = \sum C e^{-rt} + F e^{-rT}$

$D = \frac{\sum_{t=1}^T t C e^{-rt} + T F e^{-rT}}{p}$ , 加权回本时间

$D_{\text{mod}} = \frac{1}{p} \cdot \frac{dp}{dr}$ , Convexity =  $\frac{1}{p} \cdot \frac{d^2 p}{dr^2}$

$= \frac{1}{p} \cdot (\sum C e^{-rt} \cdot (-t) + (-T) \cdot F e^{-rT})$

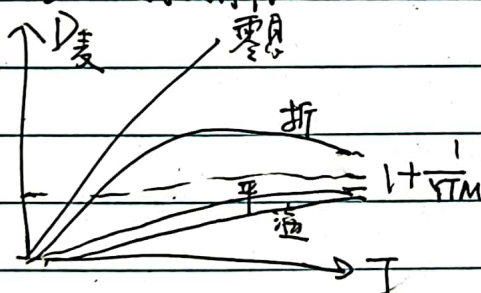
$= -D$ , 如果是离散  $D_{\text{mod}} = \frac{D}{1+r} \rightarrow$  付息区间, 不一定为年

$D = \frac{\sum_{t=1}^{nm} t \cdot \frac{C/m}{(1+\frac{y}{m})^{tm}} + n \cdot \frac{F}{(1+\frac{y}{m})^{nm}}}{\sum_{t=1}^{nm} \frac{C/m}{(1+\frac{y}{m})^{tm}} + \frac{F}{(1+\frac{y}{m})^{nm}}} = \frac{\sum t x_t + n y}{\sum x_t + y}$

$\frac{(1+\frac{y}{m})}{y} - \frac{1+\frac{y}{m} + n(\frac{C}{m} \frac{y}{m})}{\frac{C}{m}((1+\frac{y}{m})^n - 1) + y} \cdot \frac{-t}{T}$

$n$ : 剩余期数  
 $m$ : 每次间隔

修正久期没有单位



Convexity =  $\frac{1}{p} \cdot (\sum t^2 C e^{-rt} + T^2 F e^{-rT})$

$\Delta p = -D_{\text{mod}} \cdot p \cdot \Delta r + \frac{1}{2} \text{convexity} \cdot p \cdot (\Delta r)^2$

$N$ : 半年一次, 总付次数

CPT  $\rightarrow$  PV

CF  $\rightarrow$  (2nd + CE/C 清除)

I/Y: 年利率

[500] [+/-] [enter]

PMT: 每次付

DATA F5 STAT

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


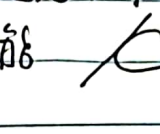
F1: 终值, 2.75

FNPV

$\rightarrow$  [CPT]



扫描全能王 创建

回到最初的问题, 切点是什么? 如果风险资产组合, 得到   
 加入无风险, 得到 , 在无风险为0权重时, 二者相交,  
 同时, 加入无风险的组合优化的结果“更优”, 不然无风险权重取0, 则为“C”上点  
 所以不可能出现 , 只可能  相切, 切点处  $W_{rf}=0$

代入该点得到二个线的斜率相同(麻烦)

该点就是有无风险的情况下, 风险资产的最优解  $\rightarrow$  市场组合

反过来, 一基金定理通过  $r_f$  和  $r_M$  可以画出

$\rightarrow$  权重固定  $\rightarrow$  一般假设市场加权

## 六. Beta of Portfolio

$$\begin{aligned}\beta^P &= \frac{\text{COV}(r_P, r_M)}{\text{Var}(r_M)} = \frac{\text{COV}(\sum W_i r_i, r_M)}{\text{Var}(r_M)} \\ &= \frac{\sum W_i \text{COV}(r_i, r_M)}{\text{Var}(r_M)} = \sum W_i \beta_i\end{aligned}$$

$$\text{SML: } r_i = r_f + \beta_i(r_M - r_f)$$

$$t. \text{ Risk: } \sigma_i^2 = \beta_i^2 \sigma_m^2 + \text{Var}(\varepsilon_i), \quad r_i - r_f = \alpha + \beta_i(r_M - r_f) + \varepsilon_i$$

$$\begin{aligned}&\downarrow \\ &\frac{\sum \text{COV}(r_i, r_M)}{\sigma_m^2 (\text{Var}(r_M))} \\ &\sigma_{ij}^2 = \beta_i \beta_j \sigma_m^2\end{aligned}$$

$$\text{COV}(r_i - r_f, r_j - r_f) = \begin{bmatrix} \beta_i & S_i & h_i \end{bmatrix} \begin{bmatrix} \sigma_{r_M - r_f}^2 & \sigma_{r_M - r_f, \text{SMB}} & \sigma_{r_M - r_f, \text{HML}} \end{bmatrix} \begin{bmatrix} \beta_j \\ S_j \\ h_j \end{bmatrix} + \text{COV}(\varepsilon_i, \varepsilon_j)$$





(CML与有效前沿的切点?) 就是市场组合

五. CAPM ~~★~~ Think again, Minimize  $\frac{1}{2} w' \Sigma w$

Such that  $r_f + w'(F - r_f I) - \bar{r}_p = 0$

$$w_0 + w'I = 1$$

Then we got  $\Sigma w + \lambda(F - r_f I) = 0$   ~~$\bar{r}_p - r_f = 0$~~

$$w'(F - r_f I) = \bar{r}_p - r_f$$

$$-\frac{1}{\lambda} = \frac{w'(F - r_f I)}{w' \Sigma w} \quad F - r_f I = \frac{-\Sigma w}{\lambda}$$

$$1 = -w' \quad = -\Sigma w \cdot \frac{w'(F - r_f I)}{w' \Sigma w}$$

$$\text{COV}(r_i, w'r) = \text{COV}(r_i, \sum_{j=1}^n w_j r_j) = \sum_{j=1}^n w_j \sigma_{ij}$$

$$\Sigma w = \begin{bmatrix} \sum_{j=1}^n w_j \sigma_{1j} \\ \vdots \\ \sum_{j=1}^n w_j \sigma_{nj} \end{bmatrix} = \begin{bmatrix} \text{COV}(r_1, w'r) \\ \vdots \\ \text{COV}(r_n, w'r) \end{bmatrix}$$

$$F - r_f I = - \begin{bmatrix} \text{COV}(r_1, w'r) \\ \vdots \\ \text{COV}(r_n, w'r) \end{bmatrix} \cdot \frac{w'(F - r_f I)}{w' \Sigma w}$$

$$\begin{bmatrix} \bar{r}_1 - r_f \\ \bar{r}_2 - r_f \\ \vdots \end{bmatrix}$$

for each row:  $\bar{r}_i - r_f = \frac{\text{COV}(r_i, w'r)}{w' \Sigma w} \cdot w'(F - r_f I)$

$$w' \Sigma w = \sum \sum w_i w_j \cdot \sigma_{ij}^2 = \sum \sum w_i w_j \text{COV}(r_i, r_j) = \frac{\text{COV}(w'r, w'r)}{\text{Var}(w'r)} \cdot w'(F - r_f I)$$

$$\text{Var}(w'r) = \text{Var}(\sum w_i r_i) = \sum w_i^2 \text{Var}(r_i) + \sum w_i w_j \text{COV}(w_i r_i, w_j r_j) = w' \Sigma w$$

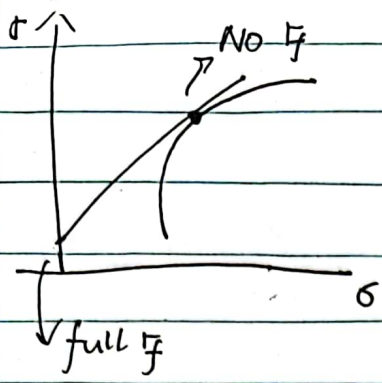
拆开有协方差

取市场组合,  $r_i - r_f = \frac{\text{COV}(r_i, r_M)}{\text{Var}(r_M)} \cdot (\bar{r}_M - r_f)$ , 此处  $w'I + w_0 = 1$

$$\downarrow$$
$$\beta_i$$



# 四. Capital Market Line



$$r_p = r_f + \frac{E(r_M) - r_f}{\sigma_M} \cdot \sigma_p$$

Tangency point is "Market portfolio"

直觉上,  $r_p = E(r_M)$  时,  $\sigma_M = \sigma_p$ ,

引入无风险资产 (此时我们完全不必其形态公式),

$$r_p = w_M E(r_M)$$

Minimize  $\frac{1}{2} W' \Sigma W$  ( $w_0$  不影响方差)

Such that  $\frac{1-w_0}{1-w_0} r_f + w' (r - r_f I) = r_p \rightarrow r_f + (w' r - I' w r_f) = r_p$   
 $w_0 + I' w = 1$

$$L(w, \lambda) = \frac{1}{2} w' \Sigma w + \lambda (r_f + w' (r - r_f I) - r_p)$$

$$\frac{\partial L}{\partial w} = \Sigma w + \lambda (r - r_f I)' = 0$$

$$w = \lambda \Sigma^{-1} (r - r_f I)$$

$$\frac{\partial L}{\partial \lambda} = r_f + w' (r - r_f I) = r_p \Rightarrow (r - r_f I)' \cdot \lambda \Sigma^{-1} (r - r_f I) = r_p - r_f$$

$$\lambda = (r - r_f I)' \cdot (r_p - r_f) (r - r_f I)' \cdot \Sigma$$

$$= (r - r_f I)' \cdot (r_p - r_f) \frac{1}{\Sigma^{-1} (r - r_f I)}$$

$$= \frac{r_p - r_f}{(r - r_f I)' \Sigma^{-1} (r - r_f I)}$$

注意没有逆矩阵

$$w_0 = 1 - I' w, \quad w = \frac{(r_p - r_f) \Sigma^{-1} (r - r_f I)}{(r - r_f I)' \Sigma^{-1} (r - r_f I)} = \frac{(r_p - r_f) \Sigma^{-1} (r - r_f I)}{(r - r_f I)' \Sigma^{-1} (r - r_f I)}$$

基金: 无风险  $\rightarrow$  基金  $w_0 = 1 - I'$   $\sigma_p^2 = w' \Sigma w = (r_p - r_f)^2 \cdot (k) \rightarrow$  常数

$\sigma_p = (r_p - r_f) \cdot \sqrt{k}$ , 斜率  $\sqrt{k}$ , 可知是一条直线

此时风险组合的权重比例不变, 同比变动



Market Model:  $r_i = \alpha + \beta r_m + \epsilon_i$   $\alpha = r_f(1-\beta)$  CAPM

Part I: effective portfolio (from mean-variance)

① 系统性风险

$$F_p = \sum W_i F_i, \quad \sigma_p^2 = \sum W_i^2 \sigma_i^2 + \sum_{i \neq j} W_i W_j \sigma_{ij}$$

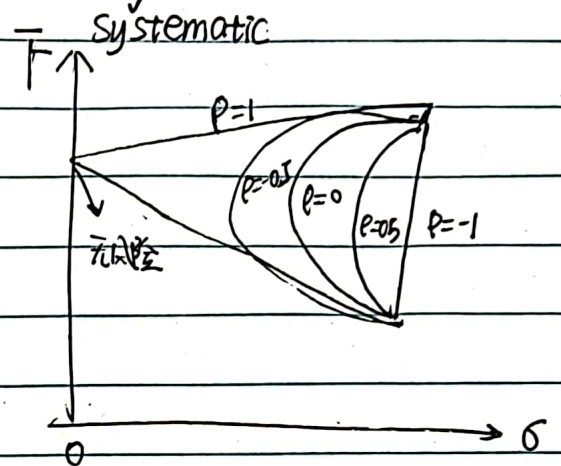
Suppose  $W_i = \frac{1}{N}$ ,  $\sigma_p^2 = \frac{\sum \sigma_i^2}{N^2} + \frac{1}{N^2} \sum_{i \neq j} \sigma_{ij} = \frac{\sum \sigma_i^2}{N^2} + \frac{1}{N^2} \cdot C_{N \times 2} \times \overline{\sigma_{ij}^2}$

$$N \rightarrow \infty \quad 0 + \frac{N-1}{N} \overline{\sigma_{ij}^2} \Rightarrow \overline{\sigma_{ij}^2}$$

②  $\rho=1$ ,  $\sigma_p = W_A \sigma_A + W_B \sigma_B$ , 无法分散

$0 < \rho < 1$ ,  $\rho \downarrow$ , 风险分散越好

$\rho < 0$ , 可以构建无风险



③ 如果不考虑  $F_p$  给定

$$L(W, \lambda) = W' \Sigma W + \lambda (1 - I'W)$$

$$\frac{\partial L}{\partial W} = 2\Sigma W - \lambda I = 0, \quad \Sigma W = \frac{\lambda}{2} I, \quad \lambda = 2\Sigma W \cdot I'$$

$$W = \frac{I' \Sigma^{-1} I}{I' \Sigma^{-1} I} I$$

I 没有逆矩阵

$$\begin{cases} I'W = 1 \\ \lambda = 2\Sigma W \cdot I' \end{cases} \Rightarrow \frac{\lambda}{2} I' \Sigma^{-1} I = I'W = 1$$

$$I' \Sigma^{-1} \frac{\lambda}{2} I \cdot \frac{1}{\lambda} = I'W = 1$$

$$\frac{\lambda}{2} = \frac{2}{I' \Sigma^{-1} I}$$

$$W = \frac{1}{2} \Sigma^{-1} \cdot \lambda I$$

$$= \frac{1}{2} \Sigma^{-1} \cdot \frac{2}{I' \Sigma^{-1} I} \cdot I$$

$$= \frac{\Sigma^{-1} I}{I' \Sigma^{-1} I} \quad (\Sigma W = \frac{I}{I' \Sigma^{-1} I})$$

