

T1.17)

1 2 3 ... T

↑
t

$$l_t = SPL(t) = \left[\begin{array}{c} \\ \\ \end{array} \right] \Bigg\}^T$$

$$l_1 = \begin{bmatrix} 1 \\ -(1+r) \\ 0 \\ \vdots \end{bmatrix}$$

$$l_2 = \begin{bmatrix} 0 \\ 1 \\ -(1+r) \\ \vdots \end{bmatrix} \dots$$

$$l_{T-1} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ -(1+r) \end{bmatrix}$$

$$C = \left\{ \begin{array}{l} C_1 = \begin{bmatrix} 1 \\ -(1+r) \\ 0 \\ \vdots \end{bmatrix} \quad C_2 = \begin{bmatrix} 0 \\ (1+r) \\ -(1+r)^2 \\ \vdots \end{bmatrix} \end{array} \right.$$

$$C_{T-1} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ (1+r)^{T-2} \\ -(1+r)^{T-1} \end{bmatrix}$$

$$C_T = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ (1+r)^{T-1} \end{bmatrix}$$

$$C = C_1 + C_2 + C_3 \dots + C_{T-1} + C_T = l_1 + l_2 \cdot (1+r) + l_3 \cdot (1+r)^2 \dots + l_{T-1} \cdot (1+r)^{T-2} + l_T \cdot (1+r)^{T-1}$$

=

$$2.5) \quad \psi: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$\psi(1,0) = 1 \quad \psi(1,-2) = 2$$

$$\psi(x) = a^T x + b$$

$$= [a_1 \ a_2] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + b$$

$$a_1 \cdot 1 + a_2 \cdot 0 + b = 1$$

$$a_1 \cdot 1 - a_2 \cdot 2 + b = 2$$

$$2a_2 = -1 \quad a_2 = -\frac{1}{2}$$

$$a_1 = 1 - b$$

$$(a) \quad \psi(1,-1) = \cancel{\text{wrong}} \\ [a_1 \ a_2] \begin{bmatrix} 1 \\ -1 \end{bmatrix} + b$$

$$= a_1 - a_2 + b = 1 - a_2$$

$$= 1 - (-\frac{1}{2}) = \boxed{\frac{3}{2}}$$

$$(b) \quad \psi(1,-2) = 2a_1 - 2a_2 + b$$

$$= 1 + b + 2(1 - b)$$

$$= 3 - b \quad \text{but idk } b$$

so ~~UNDETERMINATE~~
UNDETERMINATE

3.25
minim

$$a) \mu_p = \theta \cdot \mu + (1-\theta) \cdot \mu^{rf}$$

$$\sigma_p = \sigma_1 + \sigma_2 = |\theta| \sigma$$

where $\sigma_1 = |\theta| \sigma$ and $\sigma_2 = 0$

$$\text{when } \theta = 0 : \mu_p = \mu^{rf} \quad \sigma_p = 0 \quad \checkmark$$

$$\theta = 1 \quad \mu_p = \mu \quad \sigma_p = |\theta| \sigma = \sigma \quad \checkmark$$

$$b) \sigma_p = \sigma^{tar} = |\theta| \sigma \rightarrow |\theta| = \frac{\sigma^{tar}}{\sigma}$$

$$\theta = \pm \frac{\sigma^{tar}}{\sigma} \quad \text{as } \theta \uparrow, \sigma_p \uparrow \text{ but } \mu_p \uparrow$$

$$\text{as } \theta \downarrow, \sigma_p \downarrow \text{ but } \mu_p \downarrow$$

$$\theta = \pm \sigma^{tar} / \sigma$$

if $\mu > \mu^{rf}$ then σ^{tar} / σ

$\mu < \mu^{rf}$ then $\theta = -\sigma^{tar} / \sigma$ optimal.

c) we use leverage aka $\theta > 1$ when $\sigma^{tar} > \sigma$, or when target risk is greater than asset risk.

we use shorting when $\theta < 0$.

- in the case of (b) where

$$\theta = \frac{\sigma^{tar}}{\sigma}, \text{ we never short the stock.}$$

we hedge when $0 < \theta < 1$, or when $\sigma^{tar} < \sigma$ aka target risk is less than asset risk

4.3

Let us take a look at the set of vectors that lie on the bisecting perpendicular bisector of z_1 and z_2 .

Aka the set of all points equidistant from z_1 and z_2

if $\|x - z_1\|^2 = \|x - z_2\|^2$, $\|x - z_1\|^2 - \|x - z_2\|^2 = 0$.

Now, for the equation:

$$d = \|x - z_1\|^2 - \|x - z_2\|^2$$

as we saw before, if $d=0$, x is equidistant from both z_1 and z_2

if $\lambda < 0$, x is closer to z_1

$\delta > 0$. x is closer to z_2 .

If x is closer to z_1 , it will fall into G_1 after k-means.

closer to \mathbb{Z}_2 " " G_2 after k weeks.

$$\begin{aligned} \|x - z_1\|^2 &= (x - z_1)^T (x - z_1) = x^T x - x^T z_1 - z_1^T x + z_1^T z_1 = \cancel{\|x\|^2} - x^T z_1 - z_1^T x + \|z_1\|^2 \\ \|x - z_2\|^2 &= \quad " \quad \quad \quad " \quad \quad \quad = \cancel{\|x\|^2} - x^T z_2 - z_2^T x + \|z_2\|^2 \end{aligned}$$

$$\begin{aligned} \|x - z_1\|^2 - \|x - z_2\|^2 &= x^T z_2 - x^T z_1 + z_2^T x - z_1^T x + \|z_1\|^2 - \|z_2\|^2 \\ &= x^T (z_2 - z_1) + (z_2 - z_1)^T x + \|z_1\|^2 - \|z_2\|^2 \\ &= 2 \cdot (z_2 - z_1)^T x + \|z_1\|^2 - \|z_2\|^2 \end{aligned}$$

OOPS my function $d = \|x - z_1\|^2 - \|x - z_2\|^2$ actually does < 0 for G_1 and > 0 for G_2

It is supposed to do the opposite I think.

REVERSING SIGNS $\Rightarrow (\underline{2z}_1, -\underline{2z}_2)^T \cdot \underline{x} + \|\underline{z}_2\|^2 - \|\underline{z}_1\|^2 = \|\underline{x} - \underline{z}_2\|^2 - \|\underline{x} - \underline{z}_1\|^2$

$$w = 2(z_1 - z_2) \quad \text{and} \quad v = \|z_2\|^2 - \|z_1\|^2$$

Problem A 1.10:

```
digits = digits(:, 1:10000);
group = randi(20, 1, 10000);
Z = zeros(784, 20);
Jprev = -1;

while 1
    % compute reps (Z)
    for i = 1:20
        I = find(group == i);
        Z(:, i) = mean(digits(:, I), 2)';
    end
    J = 0;
    for i = 1:size(digits, 2)
        mingroup = -1;
        minval = 100000;
        for j = 1:20
            tv = (norm(Z(:, j) - digits(:, i)))^2;
            if (tv < minval)
                minval = tv;
                mingroup = j;
            end
        end
        J = J + minval;
        group(i) = mingroup;
    end
    if (abs(J - Jprev) <= J / 100000)
        break;
    end
    Jprev = J;
end

for k=1:20
    subplot(4,5,k)
    imshow(reshape(Z(:,k), 28, 28));
end
```

9	7	5	9	8
3	4	6	3	0
9	2	1	0	8
1	6	2	6	7