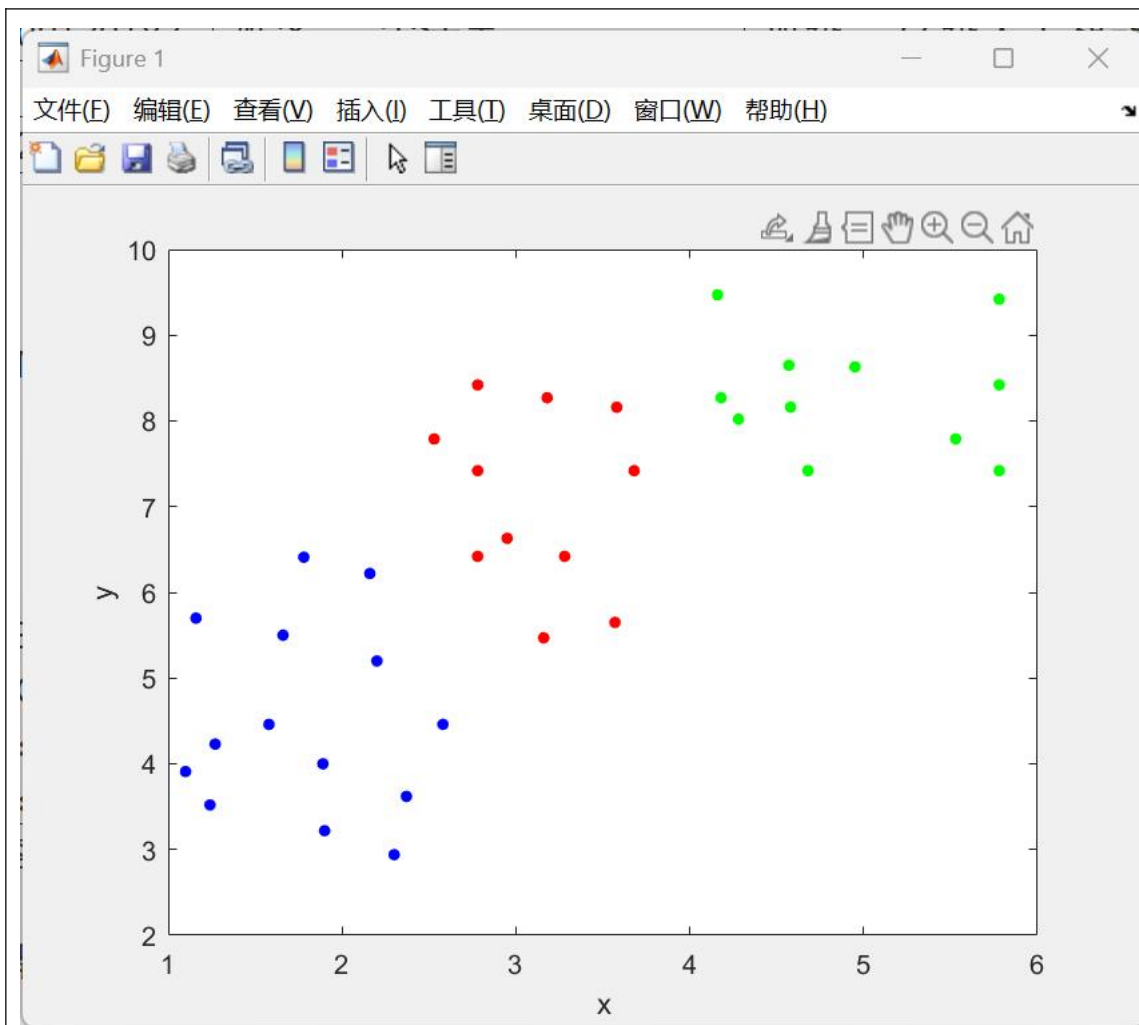


山东大学\_\_\_\_\_计算机科学与技术\_\_\_\_\_学院

\_\_\_\_\_机器学习与模式识别\_\_\_\_\_课程实验报告

学号：202300130183	姓名： 宋浩宇	班级：23 级人工智能班
实验题目：Linear Discriminant Analysis		
实验学时：2	实验日期：2025/3/11	
<p>实验环境：</p> <p>软件环境：</p> <p>系统：Windows 11 家庭中文版 23H2 22631.4317</p> <p>计算软件：MATLAB 版本: 9.8.0.1323502 (R2020a)</p> <p>Java 版本: Java 1.8.0_202-b08 with Oracle Corporation Java HotSpot(TM) 64-Bit Server VM mixed mode</p> <p>硬件环境：</p> <p>CPU：13th Gen Intel(R) Core(TM) i9-13980HX 2.20 GHz</p> <p>内存：32.0 GB (31.6 GB 可用)</p> <p>磁盘驱动器：NVMe WD_BLACKSN850X2000GB</p> <p>显示适配器：NVIDIA GeForce RTX 4080 Laptop GPU</p>		
<p>1. 实验内容</p> <p>In this exercise, you need to implement Linear Discriminant Analysis(LDA)</p> <p>2. 实验步骤</p> <p>(1) 获取实验使用的数据。</p>		



(2) 构造二分类 LDA 模型:

用  $\sum_0$  和  $\sum_1$  分别表示  $\sum_{x \in X_0} (x - \mu_0)(x - \mu_0)^T$  和  $\sum_{x \in X_1} (x - \mu_1)(x - \mu_1)^T$ , 并令  $S_w = \sum_0 + \sum_1$ .

$$w = S_w^{-1}(\mu_0 - \mu_1)$$

(3) 用 matlab 代码实现并进行计算。

(4) 完成结果的可视化。

(5) 构造多分类的 LDA 模型:

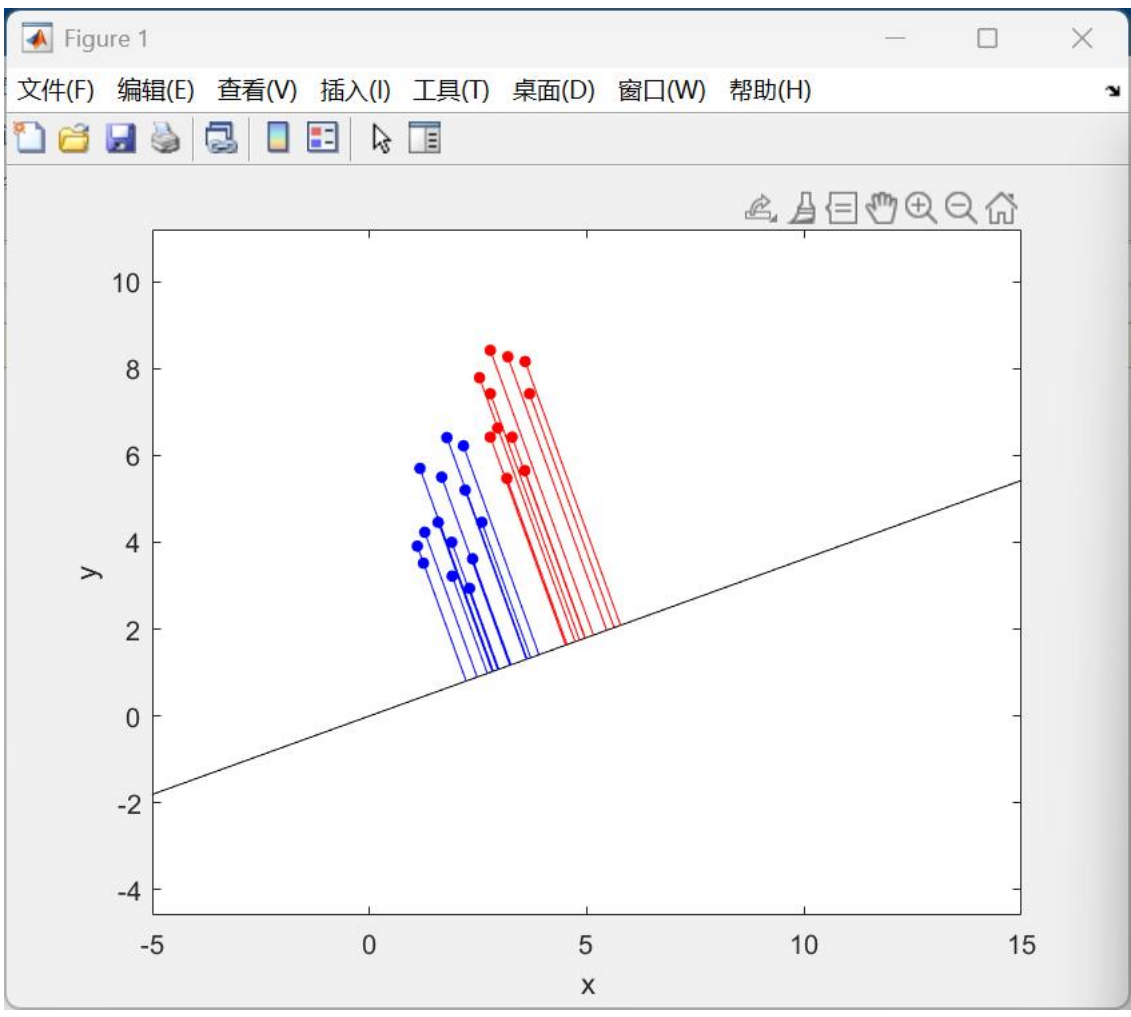
$$\begin{aligned} S_w &= \sum_{i=1}^N S_{w_i} \\ &= \sum_{i=1}^N \sum_{x \in X_i} (x - \mu_i)(x - \mu_i)^T \end{aligned}$$

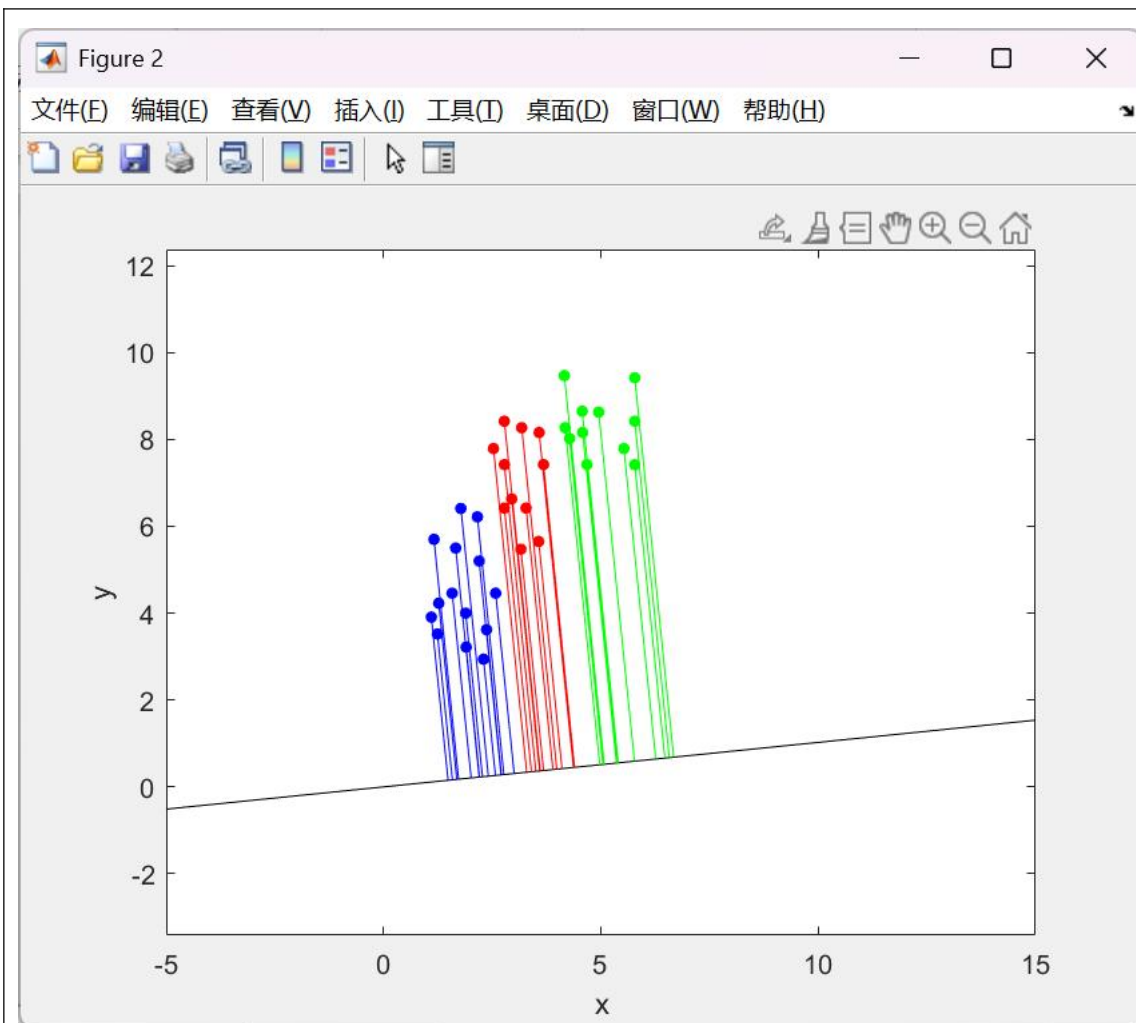
$$\begin{aligned}
S_b &= S_t - S_w \\
&= \sum_{i=1}^N \sum_{j=1}^{m_i} (x_j - \mu)(x_j - \mu)^T - \sum_{i=1}^N \sum_{x \in X_i} (x - \mu_i)(x - \mu_i)^T \\
&= \sum_{i=1}^N \sum_{j=1}^{m_i} (x_j - \mu)(x_j - \mu)^T - \sum_{i=1}^N \sum_{j=1}^{m_i} (x_j - \mu_i)(x_j - \mu_i)^T \\
&= \sum_{i=1}^N \sum_{j=1}^{m_i} [(x_j - \mu)(x_j - \mu)^T - (x_j - \mu_i)(x_j - \mu_i)^T] \\
&= \sum_{i=1}^N \sum_{j=1}^{m_i} [(x_j x_j^T - x_j \mu^T - \mu x_j^T + \mu \mu^T) - (x_j x_j^T - x_j \mu_i^T - \mu_i x_j^T + \mu_i \mu_i^T)] \\
&= \sum_{i=1}^N [\sum_{j=1}^{m_i} x_j (-\mu^T + \mu_i^T) + (-\mu + \mu_i) \sum_{j=1}^{m_i} x_j^T + \sum_{j=1}^{m_i} (\mu \mu^T - \mu_i \mu_i^T)] \\
&= \sum_{i=1}^N [m_i \mu_i (-\mu^T + \mu_i^T) + (-\mu + \mu_i) m_i \mu_i^T + m_i (\mu \mu^T - \mu_i \mu_i^T)] \\
&= \sum_{i=1}^N m_i [-\mu_i \mu^T + \mu_i \mu_i^T - \mu \mu_i^T + \mu \mu^T] \\
&= \sum_{i=1}^N m_i (\mu_i - \mu)(\mu_i - \mu)^T
\end{aligned}$$

而W的闭式解就是\*\* $S_w^{-1} S_b$ 的 $d'$ 个最大非零广义特征值所对应的特征向量组成的矩阵\*\*， $d' \leq N - 1$ 。

- (6) 用 matlab 代码实现并进行计算。
- (7) 完成结果的可视化。

### 3. 测试结果





#### 4. 附录：实现源代码

```
%% 清空环境变量
clear;
clc;

%% 加载数据
redPoints = load('ex3Data/ex3red.dat'); % 红色点
bluePoints = load('ex3Data/ex3blue.dat'); % 蓝色点
greenPoints = load('ex3Data/ex3green.dat'); % 绿色点
%% 绘制数据点
figure;
plot(redPoints(:, 1), redPoints(:, 2), 'r.', 'MarkerSize', 15);
axis equal;
hold on;
plot(bluePoints(:, 1), bluePoints(:, 2), 'b.', 'MarkerSize', 15);
% plot(greenPoints(:, 1), greenPoints(:, 2), 'g.', 'MarkerSize', 15);
xlabel('x');
ylabel('y');
%% 二分类 LDA
% disp(redPoints);
```

```

mu0 = mean(redPoints);
mu1 = mean(bluePoints);
sum0 = cov(redPoints - mu0);
sum1 = cov(bluePoints - mu1);
sum_w = sum0+sum1;
w = inv(sum_w)*(mu0-mu1)';
% disp(w);
x = linspace(-5,15);
y = x * (w(2)/w(1));
plot(x,y,'k-');
for i = 1:size(redPoints, 1)
    m = w(2) / w(1);
    b = 0;
    x0 = redPoints(i, 1);
    y0 = redPoints(i, 2);
    x1 = (m * (y0 - b) + x0) / (m^2 + 1);
    y2 = (m^2 * y0 + m * x0 + b) / (m^2 + 1);
    plot([x0 x1], [y0 y2], 'r-');
end
for i = 1:size(bluePoints, 1)
    m = w(2) / w(1);
    b = 0;
    x0 = bluePoints(i, 1);
    y0 = bluePoints(i, 2);
    x1 = (m * (y0 - b) + x0) / (m^2 + 1);
    y2 = (m^2 * y0 + m * x0 + b) / (m^2 + 1);
    plot([x0 x1], [y0 y2], 'b-');
end
%% 多分类 LDA
figure;
plot(redPoints(:, 1), redPoints(:, 2), 'r.', 'MarkerSize', 15);
axis equal;
hold on;
plot(bluePoints(:, 1), bluePoints(:, 2), 'b.', 'MarkerSize', 15);
plot(greenPoints(:, 1), greenPoints(:, 2), 'g.', 'MarkerSize', 15);
xlabel('x');
ylabel('y');
mu = mean([bluePoints;redPoints;greenPoints]);
disp(mu);
sb = size(bluePoints, 1) * cov(mean(bluePoints) - mu) + size(greenPoints, 1) * cov(mean(greenPoints) - mu) + size(redPoints, 1) * cov(mean(redPoints) - mu);
sw = cov(bluePoints - mean(bluePoints)) + cov(greenPoints - mean(greenPoints)) + cov(redPoints - mean(redPoints));
S = inv(sw) * sb;

```

```

[eigvecs, eigvals] = eig(S);
disp(eigvecs);
disp(eigvals);
max_eigval = max(diag(eigvals));
[~, idx] = max(diag(eigvals));
max_eigvec = eigvecs(:, idx);
disp(max_eigval);
disp(max_eigvec);
unit_max_eigvec = max_eigvec / norm(max_eigvec);
x = linspace(-5,15);
y = x * (unit_max_eigvec(2)/unit_max_eigvec(1));
plot(x,y,'k-');
for i = 1:size(redPoints, 1)
    m = unit_max_eigvec(2) / unit_max_eigvec(1);
    b = 0;
    x0 = redPoints(i, 1);
    y0 = redPoints(i, 2);
    x1 = (m * (y0 - b) + x0) / (m^2 + 1);
    y2 = (m^2 * y0 + m * x0 + b) / (m^2 + 1);
    plot([x0 x1], [y0 y2], 'r-');
end
for i = 1:size(bluePoints, 1)
    m = unit_max_eigvec(2) / unit_max_eigvec(1);
    b = 0;
    x0 = bluePoints(i, 1);
    y0 = bluePoints(i, 2);
    x1 = (m * (y0 - b) + x0) / (m^2 + 1);
    y2 = (m^2 * y0 + m * x0 + b) / (m^2 + 1);
    plot([x0 x1], [y0 y2], 'b-');
end
for i = 1:size(greenPoints, 1)
    m = unit_max_eigvec(2) / unit_max_eigvec(1);
    b = 0;
    x0 = greenPoints(i, 1);
    y0 = greenPoints(i, 2);
    x1 = (m * (y0 - b) + x0) / (m^2 + 1);
    y2 = (m^2 * y0 + m * x0 + b) / (m^2 + 1);
    plot([x0 x1], [y0 y2], 'g-');
end

```