实验八曲线拟合

实验目的

- 1. 掌握曲线拟合的最小二乘法原理
- 2. 理解超定方程组的最小二乘法原理
- 3. 通过联系掌握实现最小二乘法曲线拟合的编程技巧

实验环境

- 1. 计算机
- 2. MATLAB 集成环境

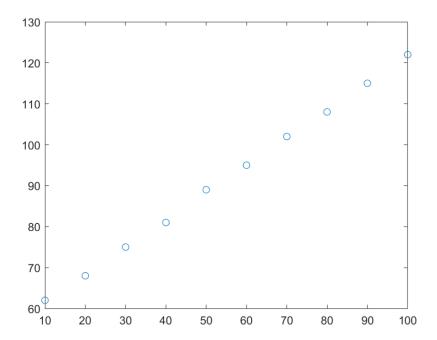
实验内容与代码

某车间计划加工一批飞机零部件,为了规定工时定额,需要确定加工零件所花费的时间,为此进行了**10**次实验,收集数据如下:

```
\begin{bmatrix} x \\ y \end{bmatrix} \begin{bmatrix} 10 & 20 & 30 & 40 & 50 & 60 & 70 & 80 & 90 & 100 \\ 62 & 68 & 75 & 81 & 89 & 95 & 102 & 108 & 115 & 122 \end{bmatrix}
```

- 1. 画出散点图
- **2**. 用最小二乘法拟合直线 $y = a \cdot x + b$

```
clc;clear;
x = [10 \ 20 \ 30 \ 40 \ 50 \ 60 \ 70 \ 80 \ 90 \ 100]
x = 1 \times 10
           20 30 40
                              50 60
                                           70
                                                       90
                                                            100
    10
                                                 80
y = [62 68 75 81 89 95 102 108 115 122]
y = 1 \times 10
    62
           68 75 81 89 95 102
                                                108
                                                            122
                                                      115
plot(x, y, 'o')
```



```
p = polyFit(x, y, 1)
```

p = 1×2 0.6685 54.9333

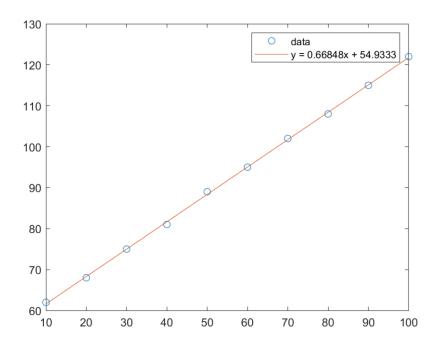
res = "
$$y$$
 = " + $p(1)$ + " x + " + $p(2)$

res = "y = 0.66848x + 54.9333"

f = polyval(p,x)

f = 1×10 61.6182 68.3030 74.9879 81.6727 88.3576 95.0424 101.7273 ...

plot(x, y, 'o', x, f, '-')
legend({'data',res})

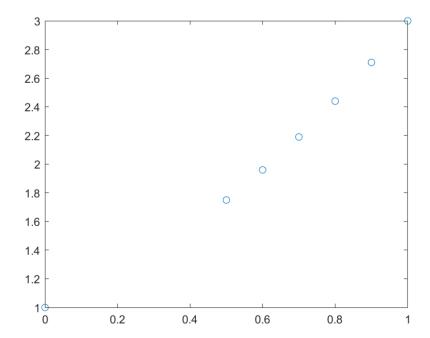


给定一组观察数据,试用最小二乘法拟合这组数据的多项式

 $\begin{bmatrix} x & 0.0 & 0.50 & 0.60 & 0.70 & 0.80 & 0.90 & 1.00 \\ f(x) & 1.0000 & 1.75 & 1.96 & 2.19 & 2.44 & 2.71 & 3.00 \end{bmatrix}$

- 1. 画出拟合数据点的图形
- 2. 确定用几次的多项次拟合这组数据
- 3. 求 f(x) 的最小二乘拟合函数

```
clc;clear;
x = [0 \ 0.5 \ 0.6 \ 0.7 \ 0.8 \ 0.9 \ 1]
x = 1 \times 7
           0
               0.5000
                            0.6000
                                        0.7000
                                                    0.8000
                                                               0.9000
                                                                           1.0000
y = [1 \ 1.75 \ 1.96 \ 2.19 \ 2.44 \ 2.71 \ 3]
                                                                           3.0000
                                        2.1900
                                                    2.4400
                                                               2.7100
     1. 0000 1. 7500
                           1.9600
plot(x, y, 'o')
```



p2 = polyFit(x, y, 2)

 $p2 = 1 \times 3$

1.0000 1.0000

1.0000

p3 = polyFit(x, y, 3)

 $p3 = 1 \times 4$

0.0000 1.0000

1.0000

1.0000

p4 = polyFit(x, y, 4)

 $p4 = 1 \times 5$

0.0000 -0.0000

1.0000

1.0000

1.0000

x1 = 0:0.01:1

 $x1 = 1 \times 101$

0

0.0100

0.0200

0.0300

0.0400

0.0500

0.0600 ...

f = polyval(p2,x1)

 $f = 1 \times 101$

1.0000 1.0101

1.0204

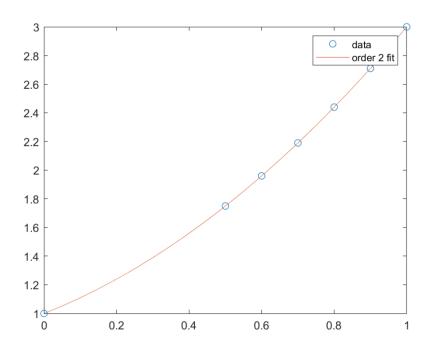
1.0309

1.0416

1.0525

1.0636 ...

```
plot(x, y, 'o', x1, f, '-')
legend({'data','order 2 fit'})
```



由上述结果可知:该函数从2次开始拟合结果相同

则可知, 拟合函数为: $f(x) = x^2 + x + 1$

此实时脚本中使用的函数:

```
function [p] = polyFit(x,y,n)
     %POLYFIT Fit polynomial to data.
         P = POLYFIT(X,Y,N) finds the coefficients of a polynomial P(X) of
         degree N that fits the data Y best in a least-squares sense. P is a
         row vector of length N+1 containing the polynomial coefficients in
     %
     %
         descending powers, P(1)*X^N + P(2)*X^{(N-1)} + ... + P(N)*X + P(N+1).
     %
     %
         Example: simple linear regression with polyfit
     %
     %
           % Fit a polynomial p of degree 1 to the (x,y) data:
     %
             x = 1:50;
     %
             y = -0.3*x + 2*randn(1,50);
     %
             p = polyFit(x,y,1);
     %
     %
           % Evaluate the fitted polynomial p and plot:
     %
             f = polyval(p,x);
             plot(x,y,'o',x,f,'-')
     %
             legend('data','linear fit')
     %
     %
     %
        Class support for inputs X,Y:
            float: double, single
     %
     %
     x = x(:);
     y = y(:);
     outputClass = superiorfloat(x,y);
     % Construct the Vandermonde matrix V = [x.^n ... x.^2 \times ones(size(x))]
     V(:,n+1) = ones(length(x),1,class(x));
     for j = n:-1:1
         V(:,j) = x.*V(:,j+1);
     end
     % Convert y to the same class as V
     y1 = cast(full(y), class(V));
     % Solve least squares problem p = V \setminus y to get polynomial coefficients p.
     [ORfactor, tau, perm, ~] =
matlab.internal.decomposition.builtin.qrFactor(V, -2);
     % use nonzero diagonal entries to determin rank for qrSolve.
     rV = sum(abs(getDiag(QRfactor)) ~= 0);
     % QR solve with rank = rV.
```

```
p = matlab.internal.decomposition.builtin.qrSolve(QRfactor, tau, perm, y1,
rV);
    % Get correct output class
     p = cast(p, outputClass);
     p = p.'; % Polynomial coefficients are row vectors by convention.
end
function d = getDiag(X)
    % get diagonal entries of X.
    if isvector(X)
         if isempty(X)
             d = X(:);
         else
             d = X(1);
         end
    else
        d = diag(X);
        d = d(:); %handle diag([])
     end
end
```

实验小结

通过此次实验,掌握了曲线拟合的最小二乘法原理,理解了超定方程组的最小二乘法原理,并通过练习掌握了实现最小二乘法曲线拟合的编程技巧