

In[]:= (*导入数据，第一列为x坐标，第二列为y坐标*)

Clear[data];

清除

data = N[Import["E:\\study_materials\\MachineLearning\\HW1\\data.txt", "Data"], 15];

导入

自然常数

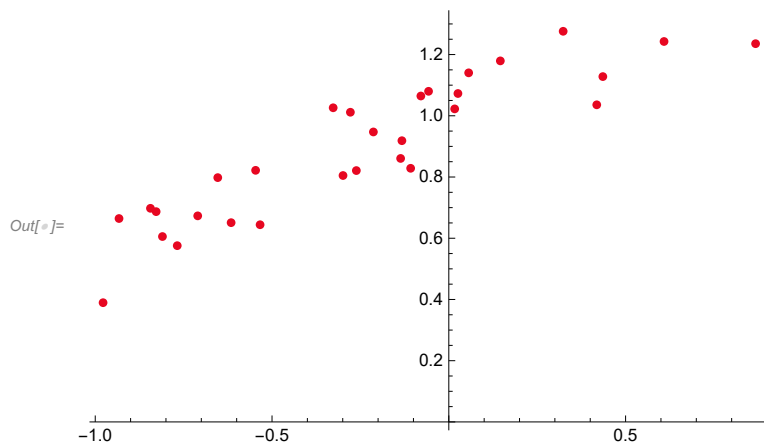
(*绘制原始数据点图*)

img = ListPlot[data, PlotStyle -> ColorData[3, "ColorList"]]

绘制点集

绘图样式

颜色数据



In[]:= (*提取x,y坐标*)

x = data[[All, 1]];

全部

y = data[[All, 2]];

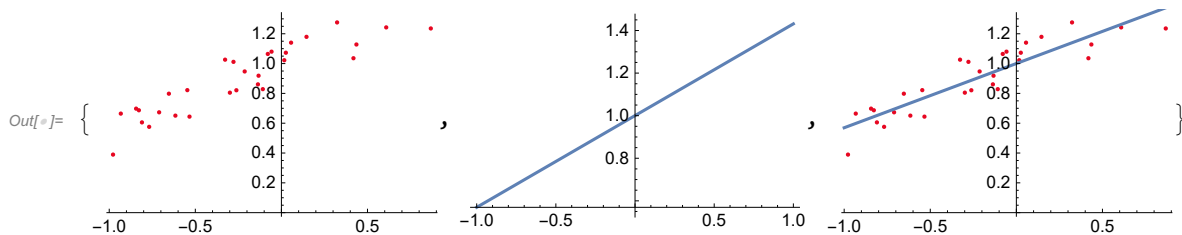
全部

```

In[ ]:= (*线性回归*)
(*生成系数矩阵*)
A1 = Table[{1, data[[i, 1]]}, {i, 1, Length[data]};
      |表格 |长度
(*求解系数*)
wlinear = Inverse[Transpose[A1].A1].(Transpose[A1].y);
      |逆 |转置 |转置
(*绘制模型图像*)
img1 = Plot[wlinear[[1]] + wlinear[[2]] * x, {x, -1, 1}];
      |绘图
(*计算二范数误差*)
y1 = wlinear[[1]] + wlinear[[2]] * x;
error1 = Sqrt[Sum[(y1[[i]] - y[[i]])^2, {i, 1, Length[y]}]];
      |... |求和 |长度
(*输出相关结果*)
Print["二范数误差error=", NumberForm[error1, 15], "\n", "w0=",
      |打印 |数值近似
      NumberForm[wlinear[[1]], 15], "\n", "w1=", NumberForm[wlinear[[2]], 15]]
      |数值近似 |数值近似
{img, img1, Show[img, img1]}
      |显示

二范数误差error=0.531194876522673
w0=1.00038680662317
w1=0.430838344703966

```

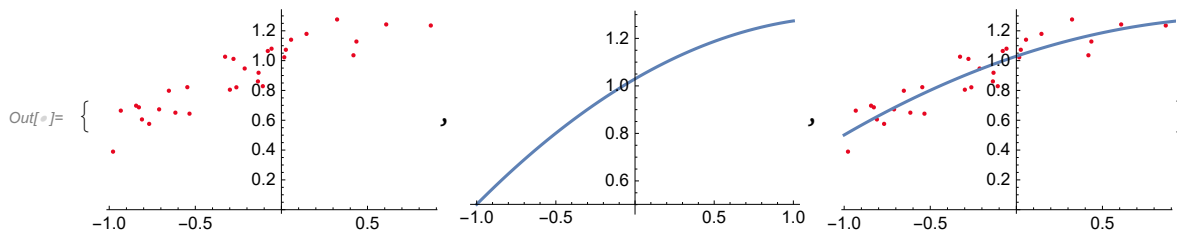


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In[ ]:= (*二次回归*)
(*生成系数矩阵*)
A1 = Table[{1, data[[i, 1]], data[[i, 1]]^2}, {i, 1, Length[data]};
      |表格                                     |长度
(*求解系数*)
wquard = Inverse[Transpose[A1].A1].(Transpose[A1].y);
      |逆      |转置      |转置
(*绘制模型图像*)
img2 = Plot[wquard[[1]] + wquard[[2]] * x + wquard[[3]] * x^2, {x, -1, 1}];
      |绘图
(*计算二范数误差*)
y2 = wquard[[1]] + wquard[[2]] * x + wquard[[3]] * x^2;
error2 = Sqrt[Sum[(y2[[i]] - y[[i]])^2, {i, 1, Length[y]}]];
      |... |求和      |长度
(*输出相关结果*)
Print["二范数误差error=", NumberForm[error2, 15],
      |打印      |数值近似
      "\n", "w0=", NumberForm[wquard[[1]], 15], "\n", "w1=",
      |数值近似
      NumberForm[wquard[[2]], 15], "\n", "w2=", NumberForm[wquard[[3]], 15]]
      |数值近似      |数值近似
{img, img2, Show[img, img2]}
      |显示

```

二范数误差error=0.492436111120306
w0=1.02956837465647
w1=0.386143334032323
w2=-0.142151113086161



可以看到，二次模型更好的回归拟合出更贴近原数据，从二范数的误差结果来看，也是二次模型更为适合一些。