

# 随机模拟方法与应用导论作业七

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## 7.5 (Hubble's Law)

In 1929 Edwin Hubble investigated the relationship between distance and velocity of celestial objects. Knowledge of this relationship might give clues as to how the universe was formed and what may happen in the future. Hubble's Law is

$$\text{Recession Velocity} = H_0 \times \text{Distance}$$

where  $H_0$  is Hubble's constant. This model is a straight line through the origin with slope  $H_0$ . Data that Hubble used to estimate the constant  $H_0$  are given on the DASL web at <http://lib.stat.cmu.edu/DASL/Datafiles/Hubble.html>. Use the data to estimate Hubble's constant by simple linear regression.

首先读取储存有数据的文件 `hubble.xls`, 数据如下

```
hubble = read.table('hubble.txt',header = TRUE)
```

```
hubble
```

##	distance	recession_velocity
## 1	0.032	170
## 2	0.034	290
## 3	0.214	-130
## 4	0.263	-70
## 5	0.275	-185
## 6	0.275	-220
## 7	0.450	200
## 8	0.500	290
## 9	0.500	270
## 10	0.630	200
## 11	0.800	300
## 12	0.900	-30
## 13	0.900	650
## 14	0.900	150
## 15	0.900	500
## 16	1.000	920
## 17	1.100	450
## 18	1.100	500

```
## 19    1.400            500
## 20    1.700            960
## 21    2.000            500
## 22    2.000            850
## 23    2.000            800
## 24    2.000           1090
```

然后调用函数`lm`，用过原点的一条直线拟合上面的数据，展示拟合直线的斜率和图像

```
M1 = lm(hubble$recession_velocity ~ 0 + hubble$distance)
```

```
M1
```

```
##
```

```
## Call:
```

```
## lm(formula = hubble$recession_velocity ~ 0 + hubble$distance)
```

```
##
```

```
## Coefficients:
```

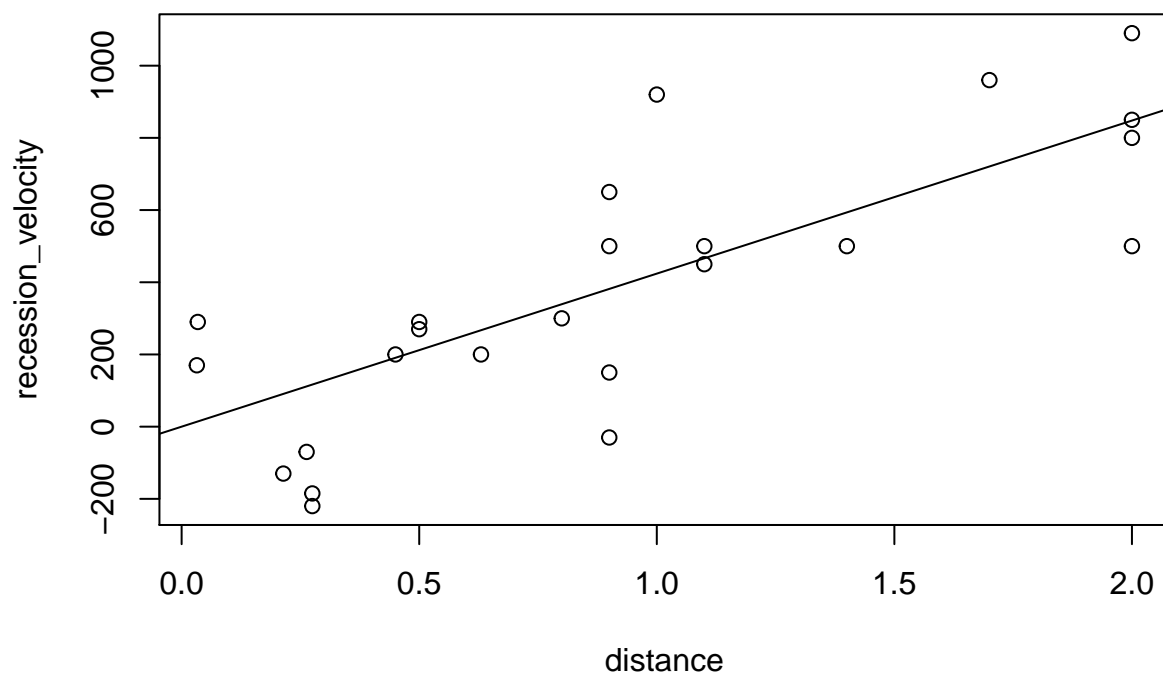
```
## hubble$distance
```

```
##           423.9
```

```
plot(hubble,main = 'The Relationship Between Distance and Velocity of Celestial Objects')
```

```
abline(M1)
```

## The Relationship Between Distance and Velocity of Celestial Object



由此可见，哈勃常数（拟合直线的斜率） $H_0 = 432.9$

$$\text{Recession Velocity} = 432.9 \times \text{Distance}$$

这意味着天体距离每增加1Mpc，其推行速度增加432.9km/s。

## 7.10 (lunatics data)

Refer to the “lunatics” data in Example 7.8. Repeat the analysis, after deleting the two counties that are offshore islands, NANTUCKET and DUCKS counties. Compare the estimates of slope and intercept with those obtained in Example 7.8. Construct the plots and analyze the residuals as in Example 7.8.

首选读取储存有数据的文件lunatics.txt，数据如下

```
lunatics = read.table('lunatics.txt',header = TRUE)
lunatics
```

##	COUNTY	NBR	DIST	POP	PDEN	PHOME
## 1	BERKSHIRE	119	97	26.656	56	77
## 2	FRANKLIN	84	62	22.260	45	81
## 3	HAMPSHIRE	94	54	23.312	72	75
## 4	HAMPDEN	105	52	18.900	94	69
## 5	WORCESTER	351	20	82.836	98	64
## 6	MIDDLESEX	357	14	66.759	231	47
## 7	ESSEX	377	10	95.004	3252	47
## 8	SUFFOLK	458	4	123.202	3042	6
## 9	NORFOLK	241	14	62.901	235	49
## 10	BRISTOL	158	14	29.704	151	60
## 11	PLYMOUTH	139	16	32.526	91	68
## 12	BARNSTABLE	78	44	16.692	93	76
## 13	NANTUCKET	12	77	1.740	179	25
## 14	DUKES	19	52	7.524	46	79

去除COUNTY为NANTUCKET和DUKES的行

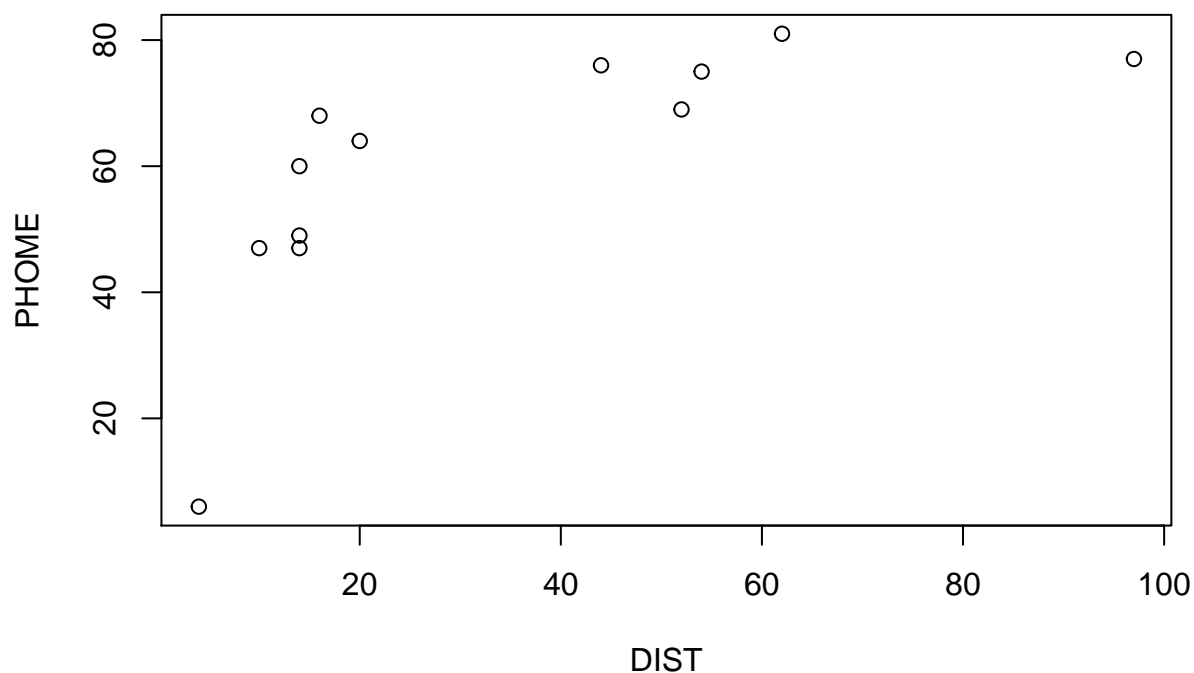
```
lunatics = lunatics[which(lunatics$COUNTY != 'NANTUCKET' & lunatics$COUNTY != 'DUKES'),]
lunatics
```

##	COUNTY	NBR	DIST	POP	PDEN	PHOME
## 1	BERKSHIRE	119	97	26.656	56	77
## 2	FRANKLIN	84	62	22.260	45	81
## 3	HAMPSHIRE	94	54	23.312	72	75
## 4	HAMPDEN	105	52	18.900	94	69
## 5	WORCESTER	351	20	82.836	98	64

```
## 6  MIDDLESEX 357  14  66.759  231  47
## 7      ESSEX 377  10  95.004 3252  47
## 8  SUFFOLK 458   4 123.202 3042   6
## 9  NORFOLK 241  14  62.901  235  49
## 10 BRISTOL 158  14  29.704  151  60
## 11 PLYMOUTH 139  16  32.526   91  68
## 12 BARNSTABLE 78  44  16.692   93  76
```

重复例7.8的分析过程，首先绘制在家照料的疯人百分比PHOME到最近的精神卫生中心的距离DIST的散点图并计算两者的相关系数

```
attach(lunatics)
plot(DIST,PHOME)
```

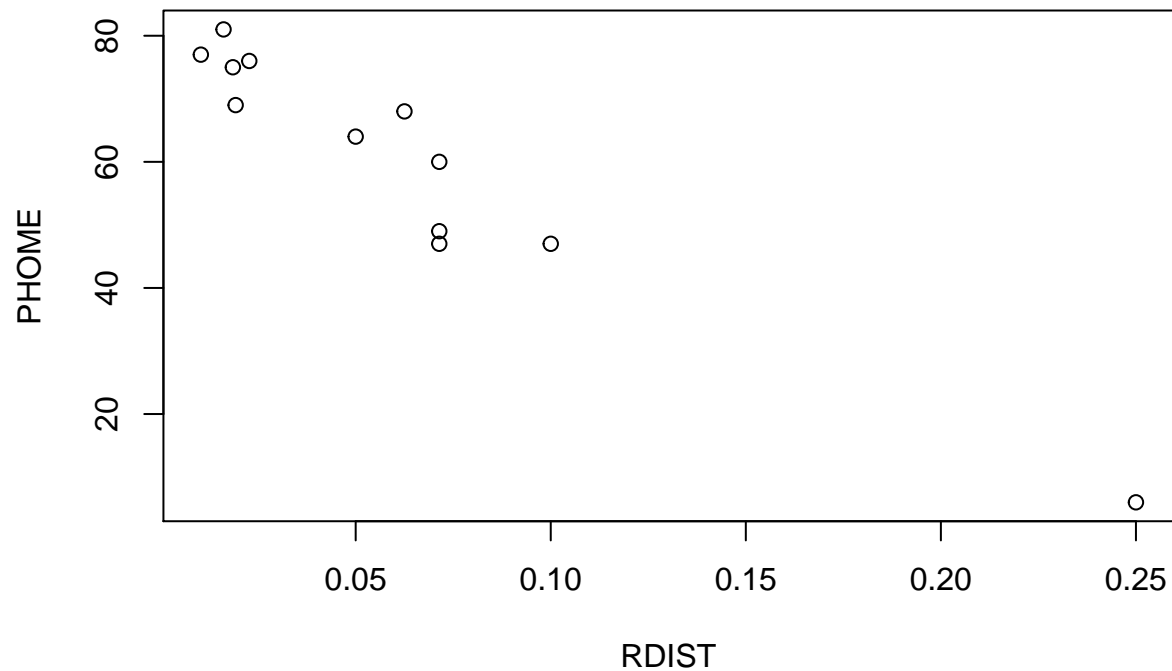


```
cor(DIST,PHOME)
```

```
## [1] 0.6963078
```

由此可见，PHOME和DIST之间的关系似乎并非线性，而更可能是双曲线。因此，取DIST的倒数RDIST，绘制PHOME关于RDIST的散点图，并计算两者之间的相关系数。

```
RDIST = 1 / DIST
plot(RDIST,PHOME)
```



```
cor(RDIST,PHOME)
```

```
## [1] -0.963456
```

在这里 $|\text{cor}(\text{RDIST}, \text{PHOME})| > |\text{cor}(\text{DIST}, \text{PHOME})|$ ，说明PHOME和RDIST之间相对于PHOME和DIST之间具有更强的线性相关性。因此，我们用简单的线性回归模型

$$\text{PHOME}_i = \beta_0 + \beta_1 \text{RDIST}_i + \varepsilon_i$$

来拟合两者之间的关系，调用函数`lm`计算相关的参数并绘制相应的拟合线

```
M2 = lm(PHOME ~ RDIST)
```

```
M2
```

```
##
```

```
## Call:
```

```
## lm(formula = PHOME ~ RDIST)
```

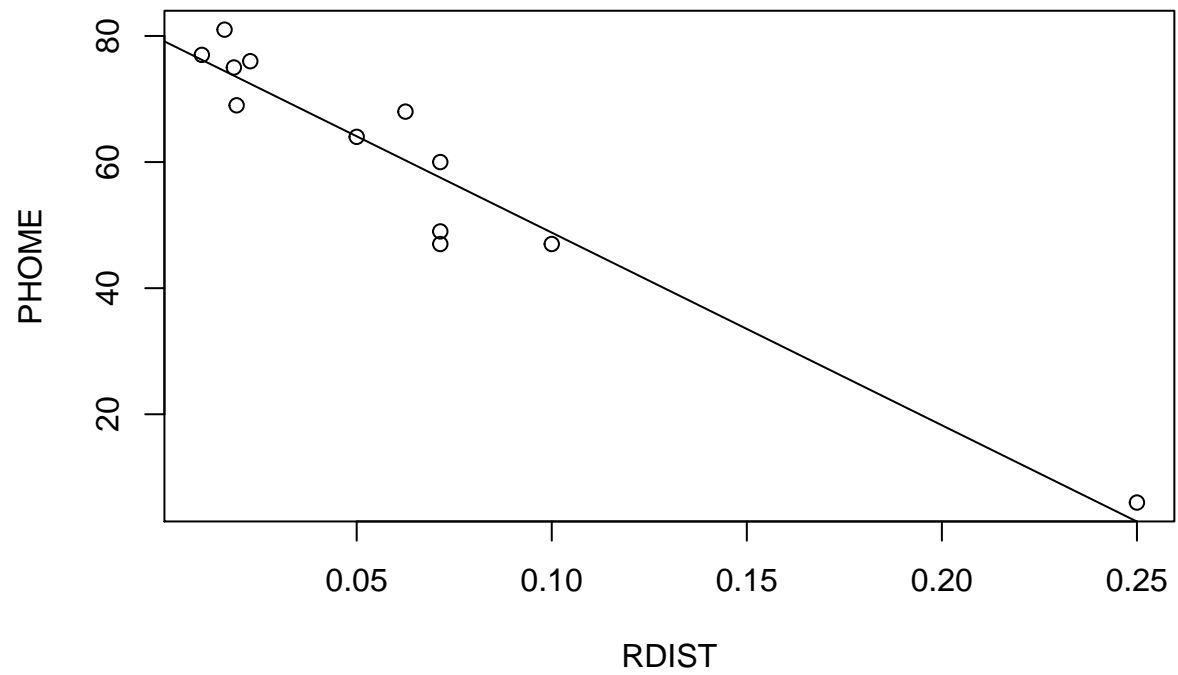
```
##
```

```
## Coefficients:
```

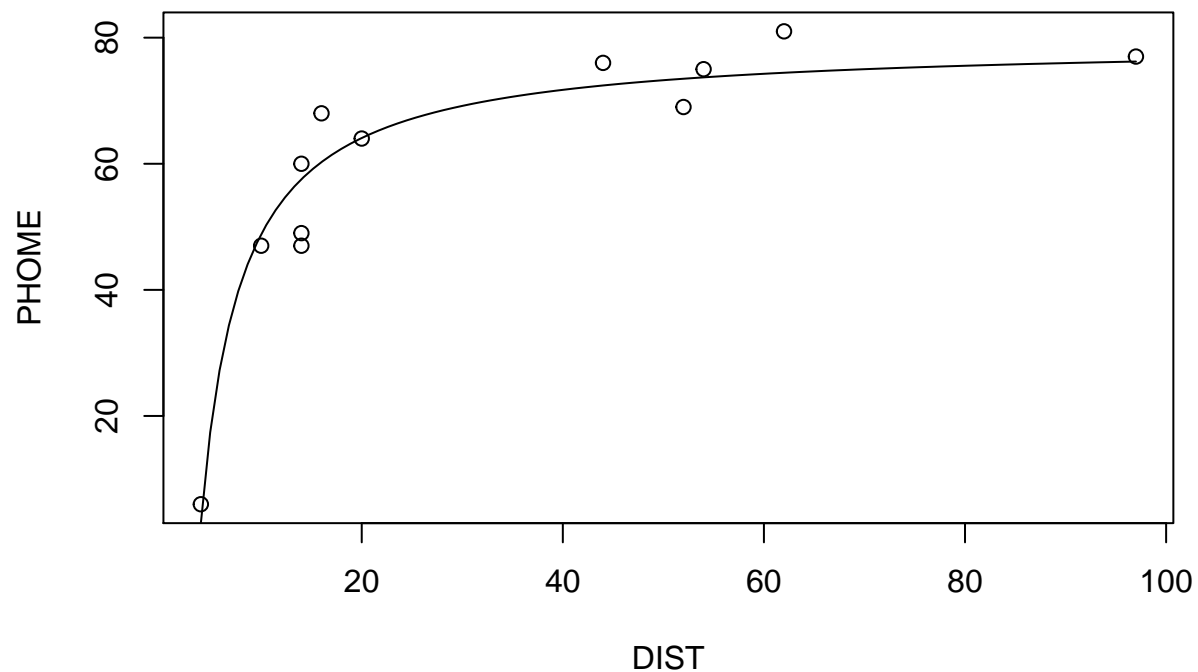
```
## (Intercept)      RDIST
```

```
##          79.36      -305.52
```

```
plot(RDIST,PHOME)  
abline(M2)
```



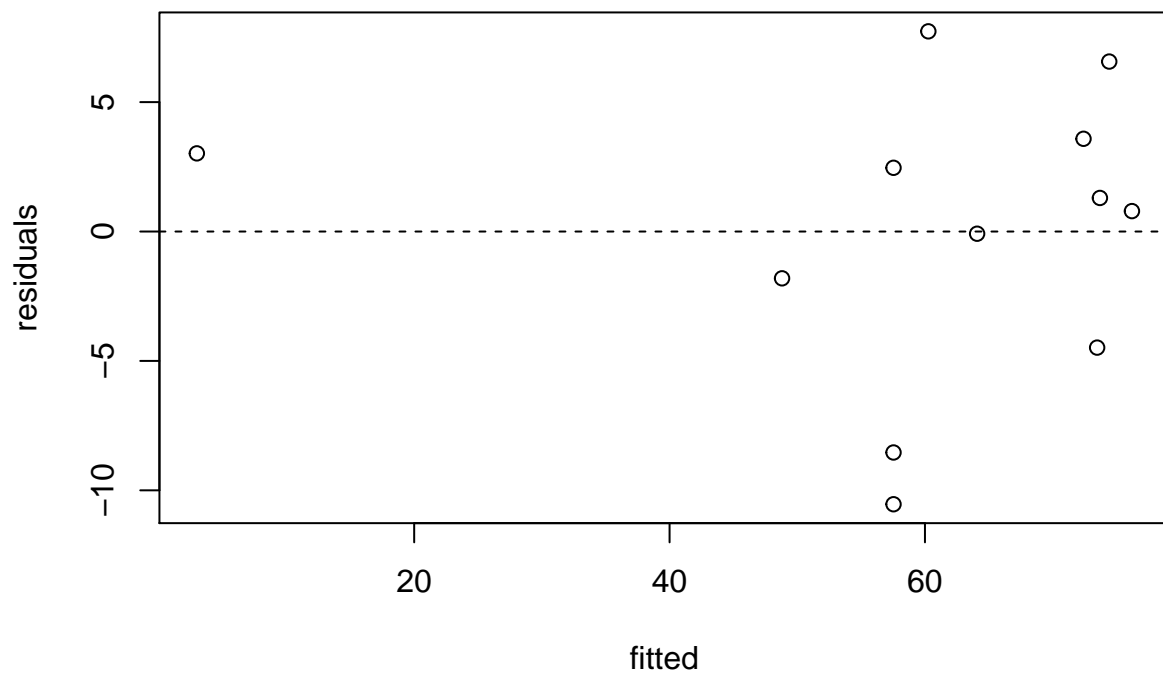
```
plot(DIST,PHOME)  
curve(M2$coef[1] + M2$coef[2] / x,add = TRUE)
```



与例7.8中斜率73.93，截距-266.32相比，当去除NANTUCKET和DUCKS两个点后，获得的截距79更大，截距-305.52更小，由图可见，拟合得到的曲线更加贴合数据的走势。

仿照例7.8，绘制残差图

```
plot(M2$fitted.values,M2$residuals,xlab = 'fitted',ylab = 'residuals')
abline(h = 0,lty = 2)
```



相比例7.8，当去除NANTUCKET和DUCKS两个点后，残差范围更小，说明拟合效果更优。