

```
> plot(OwnPrice,Sales)
> plot(OtherPrice,Sales)
> plot(OwnPrice,OtherPrice)
> plot(Prices.lm$resid,sales1.lm$resid,xlab="Residuals from
OtherPrice on OwnPrice",ylab="Residuals from Sales on OwnPrice")
```

Multiple linear regression of Sales on OwnPrice and OtherPrice:

```
> sales.lm <- lm(Sales ~ OwnPrice + OtherPrice)
> sales.lm
Call:
lm(formula = Sales ~ OwnPrice + OtherPrice)
```

```
Coefficients:
(Intercept)  OwnPrice  OtherPrice
  115.7172  -97.65737   108.7999
```

```
Degrees of freedom: 100 total; 97 residual
Residual standard error: 28.41801
```

Simple linear regression of Sales on OwnPrice:

```
> sales1.lm <- lm(Sales ~ OwnPrice)
> sales1.lm
Call:
lm(formula = Sales ~ OwnPrice)
```

```
Coefficients:
(Intercept)  OwnPrice
  211.1646  63.71296
```

```
Degrees of freedom: 100 total; 98 residual
Residual standard error: 223.4006
```

Understanding the effect of adding one predictor: Simple linear regression of OtherPrice on OwnPrice.

```
> Prices.lm<-lm(OtherPrice ~ OwnPrice)
> Prices.lm
Call:
lm(formula = OtherPrice ~ OwnPrice)
```

```
Coefficients:
```

```
(Intercept) OwnPrice
0.8772747 1.483184
```

```
Degrees of freedom: 100 total; 98 residual
Residual standard error: 2.036807
```

Just checking that the instructor is correct! Regress the residuals.

```
> lm(sales1.lm$resid ~ Prices.lm$resid)
Call:
lm(formula = sales1.lm$resid ~ Prices.lm$resid)
```

```
Coefficients:
(Intercept) Prices.lm$resid
-2.01794e-14      108.7999
```

```
Degrees of freedom: 100 total; 98 residual
Residual standard error: 28.27265
```

Then the slope is  $\beta_2 = 108.8$ , the coefficient associated with OtherPrice in the multiple regression.

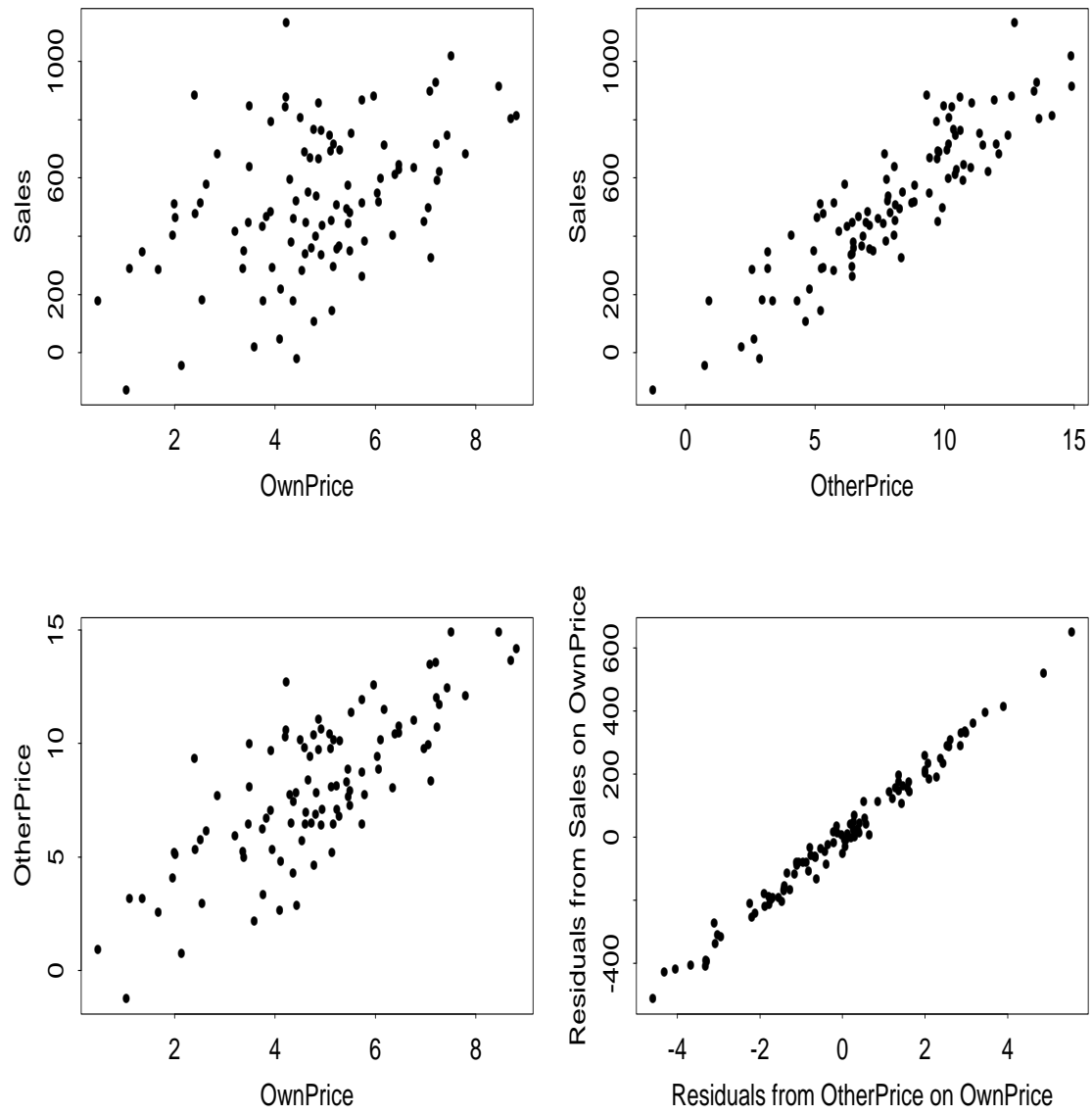


Figure 1: Explaining multiple regression