

The Salary Survey

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<https://xuejunma.github.io/>



Outline

1 The Salary Survey

2 Interaction effect

The Salary Survey

- The objective of the survey was to identify and quantify those variables that determine salary differentials
- The response variable: salary (S)
- The predictors:
 - experience (X), measured in years
 - education (E),
 - 1: high school (H.S.)
 - 2: bachelor degree (B.S.)
 - 3: advanced degree
 - management (M),
 - 1: management
 - 0: otherwise

```
1 > rm(list=ls())
2 > dat <- read.table("p130.txt", head=TRUE)
3 > attach(dat)
4 > head(dat)
5           S X E M
6 1 13876 1 1 1
7 2 11608 1 3 0
8 3 18701 1 3 1
```

- Note that when using indicator variables to represent a set of categories, the number of these variables required is one less than the number of categories.
- For example, in the case of the education categories above, we create two indicator variables $E1$ and $E2$, where

$$E_{i1} = \begin{cases} 1 = & \text{if } i\text{th person is in the H.S. category} \\ 0 = & \text{otherwise} \end{cases}$$

and

$$E_{i2} = \begin{cases} 1 = & \text{if } i\text{th person is in the B.S. category} \\ 0 = & \text{otherwise} \end{cases}$$

- stated above, these two variables taken together uniquely represent the three groups.
 - H.S., $E1 = 1$, $E2 = 0$;
 - B.S., $E1 = 0$, $E2 = 1$;
 - advanced degree, $E1 = 0$, $E2 = 0$.

$$S = \beta_0 + \beta_1 X + \gamma_1 E_1 + \gamma_2 E_2 + \delta_1 M + \varepsilon$$

Table: Regression Equations for the Six Categories of Education and Management

Category	E	M	Regression Equation
1	1	0	$S = \beta_1 X + \varepsilon + (\beta_0 + \gamma_1)$
2	1	1	$S = \beta_1 X + \varepsilon + (\beta_0 + \gamma_1 + \delta_1)$
3	2	0	$S = \beta_1 X + \varepsilon + (\beta_0 + \gamma_2)$
4	2	1	$S = \beta_1 X + \varepsilon + (\beta_0 + \gamma_2 + \delta_1)$
5	3	0	$S = \beta_1 X + \varepsilon + \beta_0$
6	3	1	$S = \beta_1 X + \varepsilon + (\beta_0 + \delta_1)$

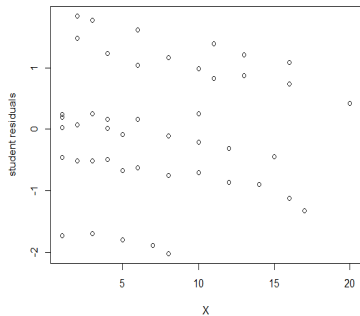
```

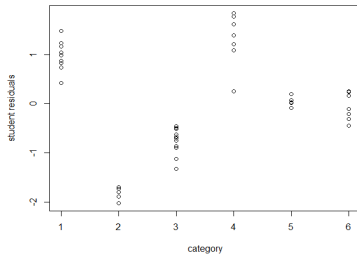
1 > E1 <- as.numeric(dat$E == 1)
2 > E2 <- as.numeric(dat$E == 2)
3 > fit <- lm(S~X+E1+E2+M, data=dat)
4 > summary(fit)
5
6           Estimate Std. Error t value Pr(>|t|)
7 (Intercept) 11031.81    383.22  28.787  < 2e-16 ***
8 X            546.18     30.52  17.896  < 2e-16 ***
9 E1          -2996.21    411.75  -7.277 6.72e-09 ***
10 E2           147.82     387.66   0.381   0.705
11 M           6883.53     313.92  21.928  < 2e-16 ***

```

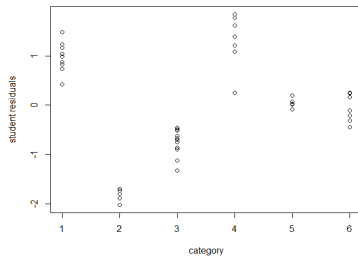
$$S = \beta_0 + \beta_1 X + \gamma_1 E_1 + \gamma_2 E_2 + \delta_1 M + \varepsilon$$

- The coefficient of X is 546.16. That is, each additional year of experience is estimated to be worth an annual salary increment of 546.
- The coefficient of the management indicator variable, δ_1 , is estimated to be 6883.50. Interpret this amount to be the average incremental value in annual salary associated with a management position.
- For the education variables,
 - γ_1 : measures the salary differential for the H.S. category relative to the advanced degree category
 - γ_2 : measures the differential for the B.S. category relative to the advanced degree category.
 - $\gamma_2 - \gamma_1$: measures the differential salary for the H.S. category relative to the B.S. category.
 - An advanced degree is worth 2996 more than a high school diploma,
 - A B.S. is worth 148 more than an advanced degree (this differential is not statistically significant, $t = 0.38$),
 - A B.S. is worth about 3144 more than a high school diploma.





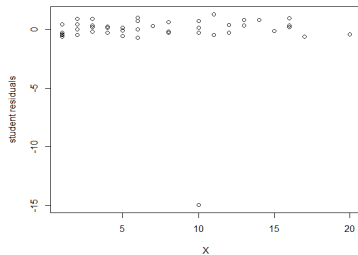
- The graph that the residuals cluster by size according to their education-management category.
- The combinations of education and management have not been satisfactorily treated in the model.
- Within each of the six groups, the residuals are either almost **totally positive** or **totally negative**.
- This behavior implies that the model does not adequately explain the relationship between salary and experience, education, and management variables.
- The graph points to some hidden structure in the data that has not been explored.
- The graphs **strongly** suggest that the effects of education and management status on salary determination are not additive.

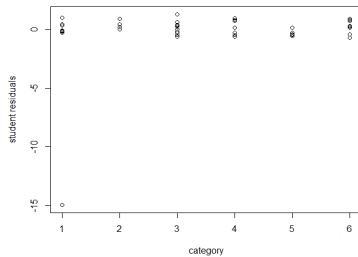


$$S = \beta_0 + \beta_1 X + \gamma_1 E_1 + \gamma_2 E_2 + \delta_1 M + \alpha_1 E_1 \bullet M + \alpha_2 E_2 \bullet M + \varepsilon$$

1 Coefficients:

2		Estimate	Std. Error	t value	Pr(> t)
3	(Intercept)	11203.434	79.065	141.698	< 2e-16 ***
4	X	496.987	5.566	89.283	< 2e-16 ***
5	E1	-1730.748	105.334	-16.431	< 2e-16 ***
6	E2	-349.078	97.568	-3.578	0.000945 ***
7	M	7047.412	102.589	68.695	< 2e-16 ***
8	E1:M	-3066.035	149.330	-20.532	< 2e-16 ***
9	E2:M	1836.488	131.167	14.001	< 2e-16 ***





Observation 33 Deleted

1 Coefficients:

2		Estimate	Std. Error	t value	Pr(> t)	
3	(Intercept)	11199.714	30.533	366.802	< 2e-16	***
4	X	498.418	2.152	231.640	< 2e-16	***
5	E1n	-1741.336	40.683	-42.803	< 2e-16	***
6	E2n	-357.042	37.681	-9.475	1.49e-11	***
7	M	7040.580	39.619	177.707	< 2e-16	***
8	E1n:M	-3051.763	57.674	-52.914	< 2e-16	***
9	E2n:M	1997.531	51.785	38.574	< 2e-16	***

- increments of approximately 500 are added to a starting salary that is specified for each of the six education-management groups.
- Since the final regression model is not additive, it is rather difficult to directly interpret the coefficients of the indicator variables.
- To see how the qualitative variables affect salary differentials, we use the coefficients to form estimates of the base salary for each of the six categories. These

Table: Regression Equations for the Six Categories of Education and Management

Category	E	M	Coefficients	Estimation
1	1	0	$\beta_0 + \gamma_1$	9459
2	1	1	$\beta_0 + \gamma_1 + \delta_1 + \alpha_1$	13448
3	2	0	$\beta_0 + \gamma_2$	10843
4	2	1	$\beta_0 + \gamma_2 + \delta_1 + \alpha_2$	19880
5	3	0	β_0	11200
6	3	1	$\beta_0 + \delta_1$	18240

