

CH. 13 Linear Regression

RUNNING ANALYSES USING JAMOV

INTERPRETTING OUTPUT

USING APA STYLE

PY 204 Statistics for Research
Dr. Valenti

Outline for Ch. 13 – Linear Regression

- 1. How to run linear regression using JAMOV**
2. How to interpret JAMOV output
3. How to write up results using APA style

Regression using JAMOVl: Example Study

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- A record company boss was interested in estimating future record sales from advertising money spent.
- Data:
 - 200 different record releases
 - Outcome variable (the var you are *trying to predict*):
 - **# of records sold** in the week after its release
 - Predictor variable (the var you are *predicting from*):
 - The **amount of money** (in £) **spent on advertising** for the record, before its release

What is the predictor variable?
What is the outcome variable?

Regression Using JAMOV

Analyses → Regression → Linear Regression

jamovi - Album_Sales

Data Analyses

Exploration T-Tests ANOVA Regression Frequencies Factor

Adverts Sales

	Adverts	Sales
1	10.256	330
2	985.685	120
3	1445.562	360
4	1270.806	270
5	220.806	220
6	308.934	170
7	471.814	70
8	537.352	210
9	514.068	200
10	174.093	300
11	1720.806	290
12	611.479	70
13	251.192	150
14	97.972	190
15	406.814	240
16	265.398	100

Correlation Matrix

Linear Regression

Logistic Regression

2 Outcomes
Binomial

N Outcomes
Multinomial

Ordinal Outcomes

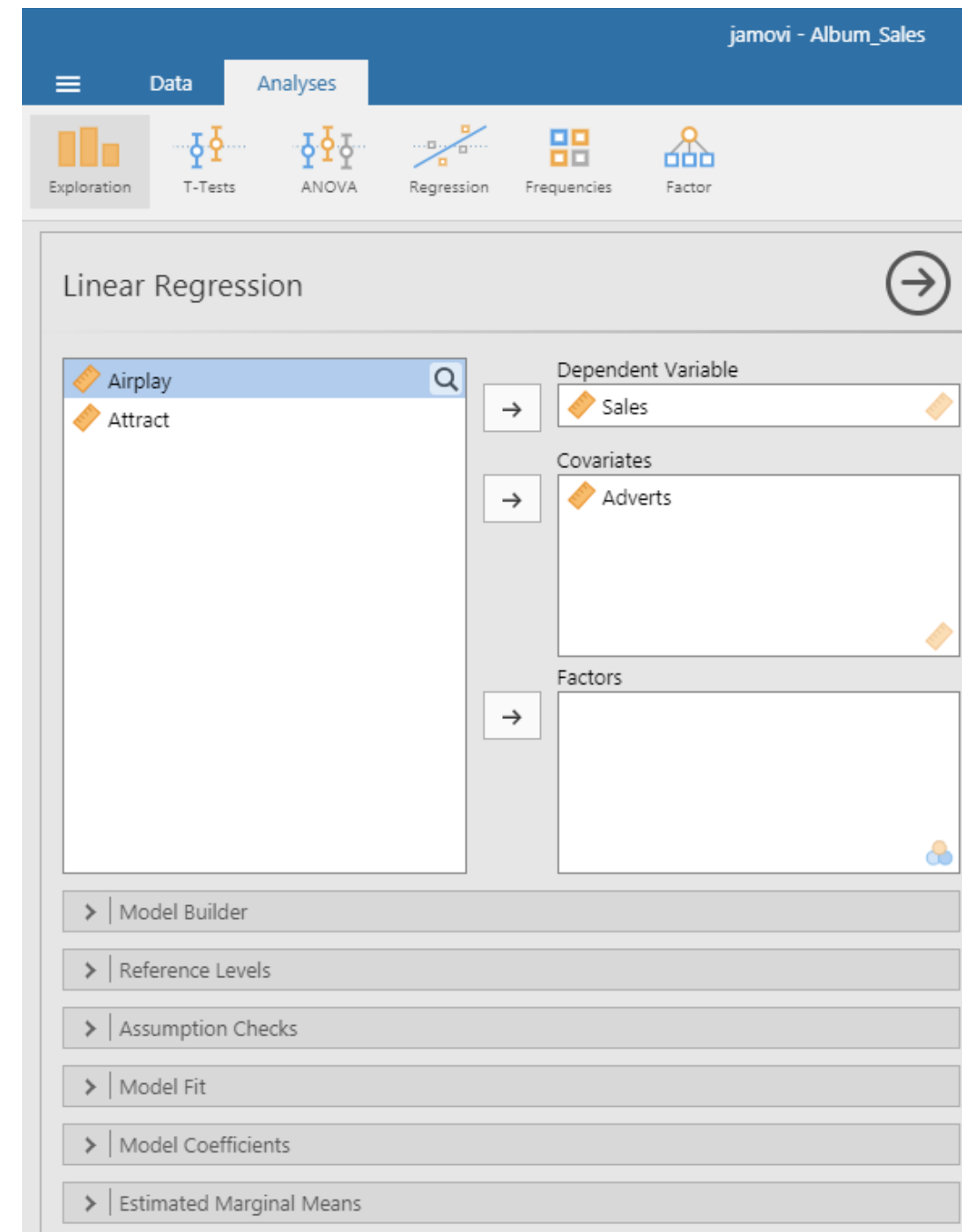
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Regression Using JAMOV

Move **outcome** variable into
“**Dependent Variable**” box.

Move **predictor** variable into
“**Covariates**” box

Leave the “Factors” box and
everything else as is.



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null & alternative hypotheses

- In regression, we test a hypothesis about whether individual predictors (Xs) are related to the outcome variable (Y).
 - **null hypothesis** (words): There is *no relationship* between our predictor and outcome.
 - $H_0: b_1 = 0$
 - **alternative hypothesis** (words): There is *some relationship* between our predictor and outcome.
 - $H_1: b_1 \neq 0$

$$\hat{Y} = b_0 + b_1 X_i$$

JAMOVl Output: Model Fit Measures & Model Coefficients

outcome variable
(Dependent
Variable box)

predictor
variable
(Covariates box)

Linear Regression

Model Fit Measures

Model	R	R ²
1	0.58	0.33

effect size

The % of variance in the **outcome** explained by the **predictor**.

33% of the variance in **record sales** is explained by **ad budget**.

Model Coefficients - Sales

Predictor	Estimate	SE	t	p
Intercept	134.14	7.54	17.80	< .001
Adverts	0.10	0.01	9.98	< .001

JAMOVl Output

Regression Equation:
aka eqn for "best fit" line

$$Y = b_0 + b_1X$$
$$Y = 134 + .10X$$

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intercept (b_0)

Model Coefficients - Sales

Predictor	Estimate	SE	t	p
Intercept	134.14	7.54	17.80	< .001
Adverts	0.10	0.01	9.98	< .001

t statistic for b_1

p -value for
the t statistic
associated
with b_1

ignore
these

regression coefficient,
aka the slope (b_1)

Interpreting the regression coefficient (slope, b_1)

Regression Equation: $Y = 134 + .10X$
aka eqn for "best fit" line

HOW DO WE INTERPRET THE SLOPE?

Model Coefficients - Sales

Outcome

Predictor	Estimate	SE	t	p
Intercept	134.14	7.54	17.80	< .001
Adverts	0.10	0.01	9.98	< .001

Predictor →

For every 1 unit increase in the predictor (X), the outcome (Y) increases by **.10** units.

Here, for every 1 unit increase in ad budget, sales increase by .10 units.

For every £1000 increase in ad budget, sales increase by 100 [.10 x 1000] records.

Interpreting the regression coefficient (slope, b_1)

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regression coefficient,
slope (b_1)

t statistic for b_1

p -value for the
 t statistic
associated w/ b_1

Model Coefficients - Sales Outcome

Predictor	Estimate	SE	t	p
Intercept	134.14	7.54	17.80	< .001
Adverts	0.10	0.01	9.98	< .001

Predictor→

What does it mean for t to be significant ($p < .05$) vs. not significant ($p \geq .05$)?

- Significant t 's mean that the predictor is significantly related to the outcome.
- Non-significant t 's mean that the predictor is unrelated to the outcome.

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Reporting regression results using APA style

Results

Advertising budget was a significant **positive** predictor of record sales, $b = .10$, $t(198) = 9.98$, $p < .001$. For every **£1000** increase in ad budget, sales increase by **100 records**. **Thirty-three percent** of the variance in **record sales** is explained by **ad budget**, a **large** effect.

For simple regression, use $t(N - 2)$, where N is the sample size (shown on earlier slide as $N = 200$ record releases).

Linear Regression

Model Fit Measures

Model	R	R ²
1	0.58	0.33

Model Coefficients - Sales

Predictor	Estimate	SE	t	p
Intercept	134.14	7.54	17.80	< .001
Adverts	0.10	0.01	9.98	< .001

Study Example two – Pubs and Death

A researcher investigated whether the number of pubs in a given region of London was related to how many deaths occurred in that region. $N = 8$ (8 regions of London)

Model Fit Measures

Model	R	R ²
1	0.81	0.65

Model Coefficients - mortality

Predictor	Estimate	SE	t	p
Intercept	3351.96	781.24	4.29	0.005
pubs	14.34	4.30	3.33	0.016

Predictor (X): **number of pubs**

Outcome (Y): **number of deaths**

Is there a significant relationship between the predictor and outcome? How do you know? **Yes, $p = .016$, which is less than .05 (the traditional significance level)**

Regression equation: $\hat{Y} = 3351.96 + 14.34X$

Results

Number of pubs in region of London was a **significant positive** predictor of number of deaths in that region ($b = 14.34$, $t(6) = 3.33$, $p = .016$). For every one additional pub in a region, number of deaths **increase** by **14.34**. **Sixty-five** percent of the variance in **deaths** can be explained by **the number of pubs in the region**, a **very large** effect.