

Topic materials:

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**Module Title:** Introduction to Statistics

**Session Title:** Path Diagrams

### **Topic title: Mediation**

After working through this session you should be able to:

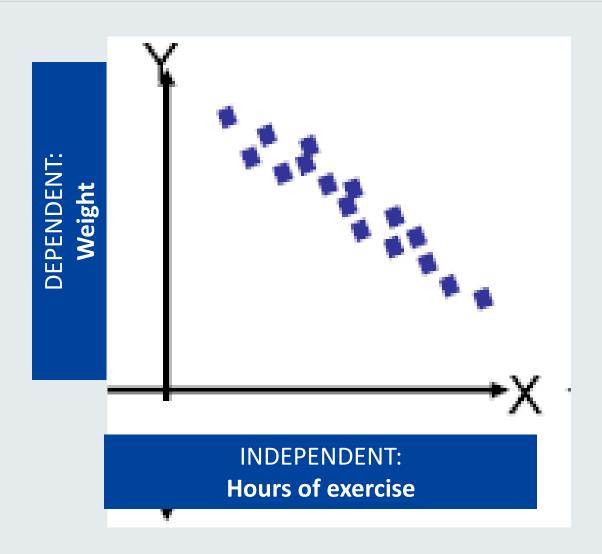
- To understand the concept of mediation
- To understand what a path diagram is and how to build it
- To understand the concept of direct, indirect and total effect and how to compute them

16 people were observed to see if the weight of a person, related to the hours of exercise they conducted. The following hypothesis was investigated:

**Hypothesis** 'The higher the number of hours of exercise the lower the weight'.

Plotting the data is essential to understand and visually assess the relationship between pairs of continuous variables

The plot of data points (x,y) with x =hours of exercise and y =weight of a person where the data is continuous is called a **scatterplot**.



#### **Questions:**

Q1: How strong is the linear relationship? Understand the direction and magnitude of the linear relationship

A1: Correlation Coefficient (Pearson) *r*=-0.85

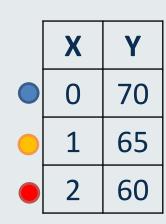
There is strong, negative, linear association between hours of exercise and weight loss (r=-0.85)

Q2: Can the relationship between variables be described by fitting a line to the observed data?

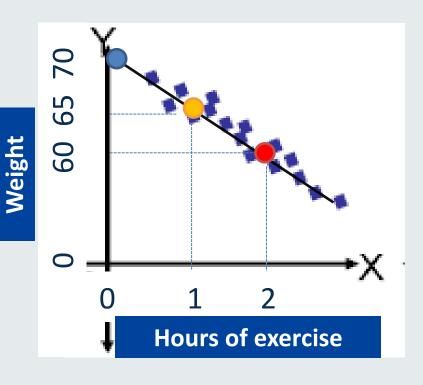
A2: Yes, because there is a **linear relationship**. The relationship is expressed as an equation

$$y = \beta_0 + \beta_1 x$$

where  $\beta_0$  is the y intercept = 70 where  $\beta_1$  is the slope of the line = -5



$$y = 70 - 5x$$



$$\beta_0$$
=70;  $\beta_1$ =-5;



#### **Interpretation:**

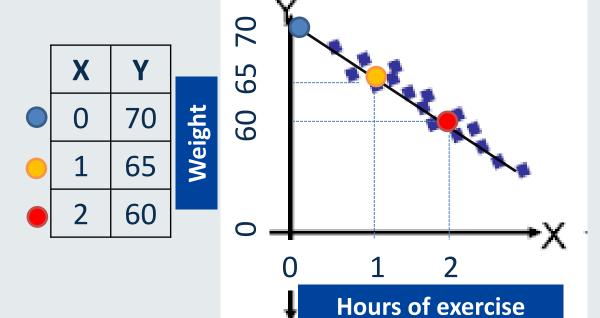
- $\beta_0 = 70$ , When hours of exercise = 0, weight is 70kg.
- $\beta_1$  = -5, Each additional hour of exercise decreases weight by 5kg.

#### **Linear regression model:**

- To measure to what extent there is a linear relationship between two variables
- A rule that predicts weight given the hours of exercise.



$$y = 70 - 5x$$



Simple linear regression

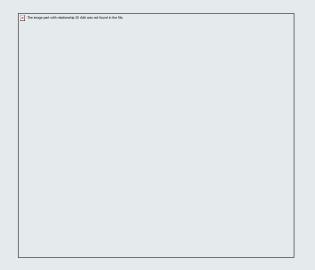
Multiple linear regression

$$y = 70 - 5x + \varepsilon$$

Where: y=weight; x=exercise;



Where: y=weight;  $x_1$ =exercise;  $x_2$ =diet;

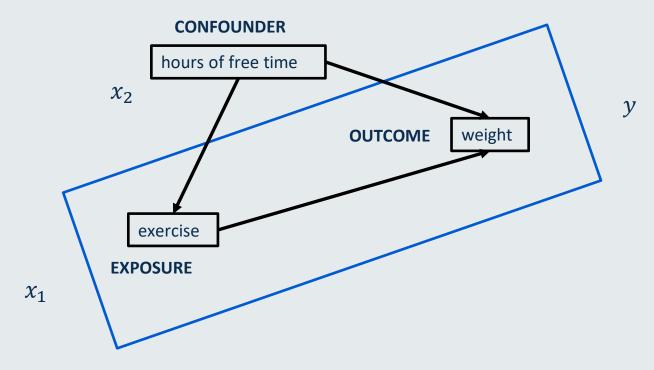




A simple regression model (one independent variable) fits a regression line  $y = \beta_0 + \beta_1 x_1$ 

A multiple regression model with two explanatory variables fits a regression plane  $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2$ 

 $x_2$  is a **confounding variable** when it has an effect both on the dependent Y and independent  $x_1$  variable.



Using multiple linear regression allows us to hold all <u>other independent variables constant</u>, allowing us to get an estimate of the effect of the independent variable of interest <u>while adjusting for other variables in the model which are hypothesized to be confounders</u>.

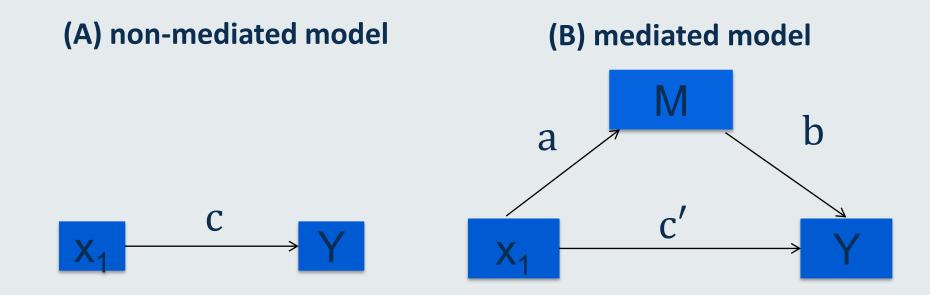
#### **Mediation**

The third variable  $x_2$  can take a role of **mediator**. A mediator explains **a portion of the association** between Y and x1. When  $x_2$  is a mediator will denote it "**M**".

Mediation is a hypothesised causal mechanism by which one variable affects another variable.

A **mediator** (M) of the causal effect of independent variable  $(x_1)$  on dependent variable (Y) is a variable  $x_2$  on the causal pathway from  $x_1$  to Y.

#### **Mediation**



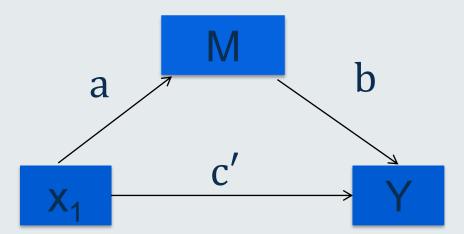
- In a **non-mediated model** (A), the **total effect** of the independent variable  $x_1$  on the dependent Y is denoted by the path c
- Under a **mediated model** (B), the total causal effect c can be split into an indirect (or mediated) part with paths a and b and a direct (non-mediated) path c'

#### **Mediation**

(a) non-mediated model

(b) mediated model





- **Direct** effect = c'
- **Indirect** effect (or "mediated" effect) = a\*b
- c = Total effect = direct + indirect effect = c' + a \* b

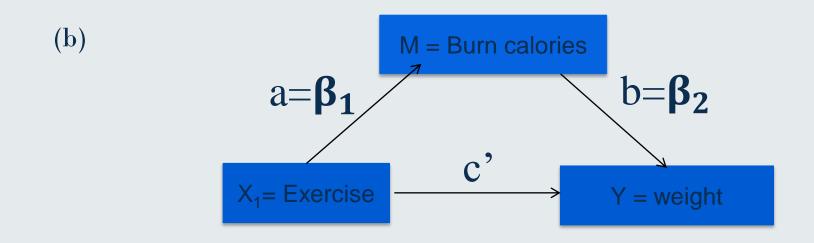
### Investigating a Mediation Effect: Computing a, b and c

We want to look at the relationship between exercise and weight and consider the calories burned as a mediator of the exercise – weight relationship.



#### 1. Estimate of path c:

$$Y = \beta_0 + \beta X_1 + \epsilon$$



#### 2. Estimate of path a:

$$M = \beta_0 + \beta_1 X_1 + \epsilon$$

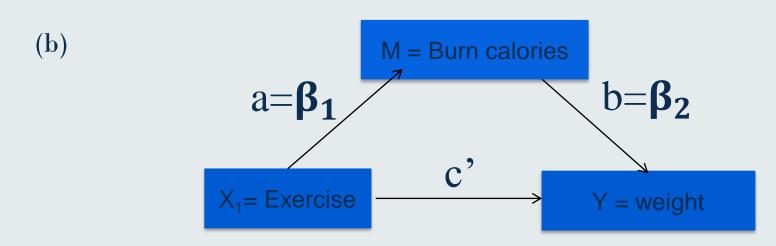
#### 3. Estimate of path b:

$$Y = \beta_0 + \beta_2 M + \beta_3 X_1 + \varepsilon$$

### Investigating a Mediation Effect: Computing a, b and c



## 4: Estimate of path c': 2 different ways:



i. 
$$c = c' + a * b$$
  
 $\beta = c' + \beta_1 * \beta_2$ 

ii. From step 3 model:

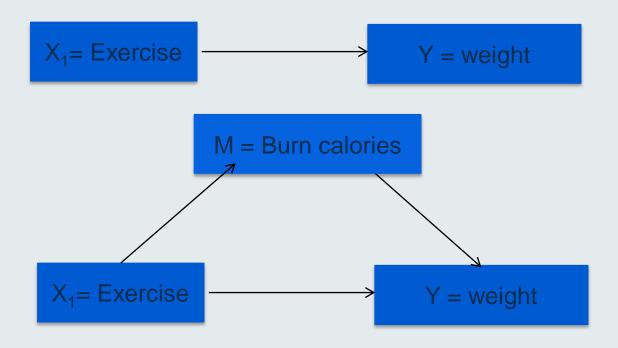
$$Y = \beta_0 + \beta_2 M + \beta_3 X_1 + \varepsilon$$
$$c' = \beta_3$$

### **Knowledge Check**

Q1: In a mediated model, which of the next sentences MUST be TRUE?

- a) The independent variable (X) causes the outcome variable (Y)
- b) The independent variable (X) causes the mediator variable (M)
- c) The mediator (M) causes the outcome variable (Y) when controlling for the independent variable (X).

Q2: Given the two path diagrams below and the set of models, compute a, b, c and c'



Y= 70 - 5 
$$X_1 + \varepsilon$$

M= 0.5+  $2X_1 + \varepsilon$ 

Y = 69 -1.5  $M$  -2  $X_1 + \varepsilon$ 

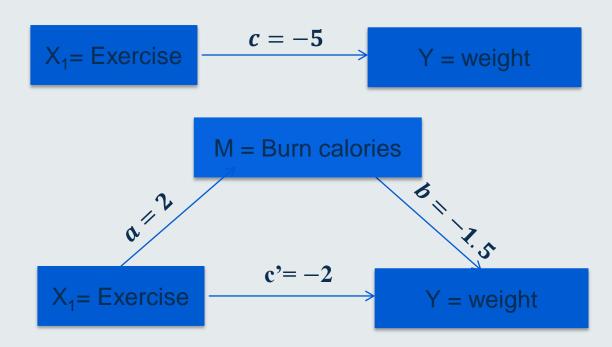
### **Knowledge Check Solutions**

Q1: In a mediated model, which of the next sentences MUST be TRUE?

- a) The independent variable (X) causes the outcome variable (Y)
- b) The independent variable (X) causes the mediator variable (M)
- c) The mediator (M) causes the outcome variable (Y) when controlling for the independent variable (X).

  Answer: c) must be TRUE.

Q2: Given the two path diagrams below and the set of models, compute a, b, c and c'



Y= 70 - 5 
$$X_1 + \varepsilon$$
M= 0.5+  $2X_1 + \varepsilon$ 
Y = 69 - 1.5 M - 2  $X_1 + \varepsilon$ 

$$\beta = c' + \beta_1 * \beta_2$$

$$-5 = c' + 2 * -1.5;$$

$$-5 = c' - 3;$$

$$c' = -2 = \beta_3$$

Answer: a=2; b=-1.5; c=-5; c'=-2

#### References

MacKinnon, D. P., Fairchild, A. J. and Fritz, M.S (2007). Mediation analysis, Annual Review of Psychology, 58, 593–614

David Kenny's Website on mediation: http://davidakenny.net/cm/mediate.htm

Hayes, A.F. (2013). Introduction to Mediation, Moderation, and Conditional Process Analysis, Guildford Press.

MacKinnon, D. P., Fairchild, A. J. and Fritz, M.S (2007). Mediation analysis, Annual Review of Psychology, 58, 593–614



### Thank you



Please contact your module leader or the course lecturer of your programme, or visit the module's forum for any questions you may have.

If you have comments on the materials (spotted typos or missing points) please contact Dr Iniesta:

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For any other comments or remarks on the module structure, please contact one of the three module leaders of the Biostatistics and Health Informatics

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