

MSBA7003 Quantitative Analysis Methods

Tutorial 04

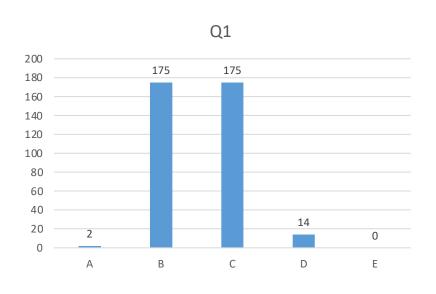
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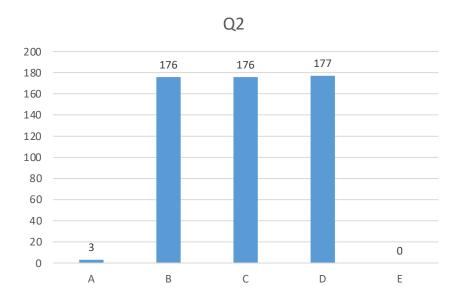
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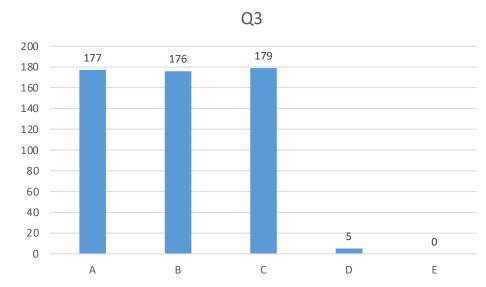
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Assignment 3









Agenda

- Solutions to Assignment 03
 - Q1 & Q2 Linear Programming Model
 - Q3 -- Efficiency Evaluation (DEA)
- Graphical Solution Method for LP
- Mixed Integer Programming
 - Violating Proportionality
- The Naïve estimator
- Causal Inference
 - Maternal Age and Perinatal Outcomes



Jimmy, an aircraft engineer, is designing the passenger cabin for a new aircraft model. This model will be used by an airline for a specific route. Jimmy needs to decide on the number of seats for both business class and economy class. There must be 4 seats per row in business class; the seat length is 1.2 length units; the seat weight is 2.2 weight units. In economy class, there must be 6 seats per row; the seat length is 1 unit; the seat weight is 1 weight unit. The total cabin length is 32 length units, and the seat weight limit is 200 weight units. Additionally, due to limited demand, business class cannot have more than 6 rows. Jimmy's optimization goal is to maximize the expected profit of the aircraft model for its entire lifecycle on this route. It is known that the profit margin of business class seats is twice that of economy class seats.

Which of the following statement(s) is(are) true?

- A) In the optimal design plan, there should be 4 rows for business class and 27 rows for economy class.
- B) In the optimal design plan, there should be 5 rows for business class and 26 rows for economy class.
- C) In the following problem formulation given by GPT4, constraints 1-3 are all wrong.
- D) The total weight constraint is not binding.



• let x, y denote the # of rows of business class, and economy class, respectively. We have the following formulation:

$$\max (2 * 4)x + 6y$$

$$1.2x + y \le 32$$

$$(2.2 * 4)x + 6y \le 200$$

$$x \le 6$$

$$x, y \ge 0$$

We use Excel to build and solve the model as follows.

	# of business	# of economy			
Decision Variables	5	26			
coeff.	8	6			
Constratints	1.2	1	32		
	8.8	6	200	<=	
	1		5	<=	

Note: sensitivity report is not available if having an integer constraint.



- The Heinlein and Krampf Brokerage firm is instructed by a client to invest \$250,000 among five possible asset classes, as shown in the table below. The return of each security is uncertain and depends on the market scenario. The client listed the following guidelines:
- (1) Investment Grade Bonds should constitute at least 20% of the invested value;

$$x_1 \ge 20\%(x_1 + x_2 + x_3 + x_4 + x_5)$$

- (2) At least 40% of the invested value should be placed in a combination of international equity and S&P 500; $x_4 + x_5 \ge 40\%(x_1 + x_2 + x_3 + x_4 + x_5)$
- (3) No more than 80% of the amount invested in bonds should be placed in hedge funds; $x_3 \le 80\%x_1$
- (4) No short selling, and it is okay to hold some cash.

$$x_1 + x_2 + x_3 + x_4 + x_5 \le 250,000$$

Projected Return (%)	in Scenario 1	In Scenario 2
Investment Grade Bonds	1.1	-2.3
Commodities	-5.0	6.8
Hedge Funds	-1.7	4.9
International Equity	-0.3	8.4
S&P 500	-7.5	11.8



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- The objective is to maximize the worst-case return.
- Define x_i as the amount of money allocated to class i. Denote as r_{ij} as the return rate of class i in scenario j.
- The worst-case payoff is $\min\{\sum x_i r_{i1}, \sum x_i r_{i2}\}$.
- Hence, the LP can be formulated as follows

max t

s.t.
$$t \le 1.1x_1 - 5.0x_2 - 1.7x_3 - 0.3x_4 - 7.5x_5$$

 $t \le -2.3x_1 + 6.8x_2 + 4.9x_3 + 8.4x_4 + 11.8x_5$
 $x_1 \ge 20\%(x_1 + x_2 + x_3 + x_4 + x_5)$
 $x_4 + x_5 \ge 40\%(x_1 + x_2 + x_3 + x_4 + x_5)$
 $x_3 \le 80\%x_1$
 $x_1 + x_2 + x_3 + x_4 + x_5 \le 250,000$
 $x_1, x_2, x_3, x_4, x_5 \ge 0$



Investment	Investment Grade Bonds	Commodities	Hedge Funds	International Equity	S&P 500			
Scer.1-Projected Return (%)	0.011	-0.05	-0.017	-0.003	-0.075	1350		
Scer.2-Projected								
Return (%)	-0.023	0.068	0.049	0.084	0.118	4950		
Amount	150000	0	0	100000	0			
Total Return	1350							
Constraints								
	1	1	1	1	1	250000	<=	250000
	1					150000	>=	50000
				1	1	100000	>=	100000
			1			0	<=	120000

- A) It is optimal to hold some cash.
- B) In the optimal solution, the investment in S&P 500 is zero.
- C) In the optimal solution, the worst-case expected return is \$1,350.



Variable Cells

		Final	Reduced	Objective	Allowable	Allowable
Cell	Name	Value	Cost	Coefficient	Increase	Decrease
\$B\$6	Total Return Investment Grade Bonds	1350	0	1	1E+30	1
\$B\$4	Amount Investment Grade Bonds	150000	0	0	1E+30	0.009
\$C\$4	Amount Commodities	0	-0.061	0	0.061	1E+30
\$D\$4	Amount Hedge Funds	0	-0.028	0	0.028	1E+30
\$E\$4	Amount International Equity	100000	0	0	0.014	0.0135
\$F\$4	Amount S&P 500	0	-0.072	0	0.072	1E+30

Constraints

		Final	Shadow	Constraint	Allowable	Allowable
Cell	Name	Value	Price	R.H. Side	Increase	Decrease
\$B\$6	Total Return Investment Grade Bonds	1350	1	0	3600	1350
\$B\$6	Total Return Investment Grade Bonds	1350	0	0	1E+30	3600
\$G\$10		150000	0	0	100000	1E+30
\$G\$11		100000	-0.014	0	96428.57143	29752.06612
\$G\$12		0	0	0	1E+30	120000
\$G\$9		250000	0.0054	250000	1E+30	250000

D) In the optimal solution, the return for an additional invested dollar is \$0.0054.



• Q3. The Salem Board of Education wants to evaluate the efficiency of the town's four elementary schools. The three outputs of these schools are: (1) average reading score, (2) average mathematics score, and (3) average self-esteem score. The three inputs to these schools are: (1) average educational level of mothers (defined by highest grade completed: 12 = high school graduate, 16 = college graduate, and so on), (2) number of parent visits to school (per child), and (3) teacher-to-student ratio. The relevant information for the four schools is given in the table below.

School	Input 1	Input 2	Input 3	Output 1	Output 1	Output 3
Α	14.5	3	0.25	3.5	2.7	3
В	13	2	0.13	3.3	2.5	2.4
С	15.5	4	0.28	3.8	3	3.3
D	16.2	3	0.33	4	3.8	4



- We take school A as an example. For schools B, C, and D, the solving process is similar.
- Assume constant returns to scale

ng constant r	eturns to scal	e			
ng A					
Input 1	Input 2	Input 3	Output 1	Output 1	Output 3
0.0689655	0	0	0.251341	0	0.0279693
TOV		TIV			Efficiency
0.9636015	<=	1	fixed as 1		96%
0.8965517	<=	0.8965517			100%
1.0473946	<=	1.0689655			98%
1.1172414	<=	1.1172414			100%
	ng A Input 1 0.0689655 TOV 0.9636015 0.8965517 1.0473946	ng A Input 1 Input 2 0.0689655 0	Input 1 Input 2 Input 3 0.0689655 0 0 TOV TIV 0.9636015 = 1 0.8965517 = 0.8965517 1.0473946 = 1.0689655	Input 1 Input 2 Input 3 Output 1 0.0689655 0 0 0.251341 TOV TIV 0.9636015 <= 1 fixed as 1 0.8965517 <= 0.8965517 1.0473946 <= 1.0689655	Input 1



• If we assume non-constant returns to scale, we build the following model to check whether school A is efficient or not.

Assuming	non-consta	nt returns	to scale					
Evaluating	Α						Weight	
Α	14.5	3	0.25	3.5	2.7	3	0	
В	13	2	0.13	3.3	2.5	2.4	0.53125	
С	15.5	4	0.28	3.8	3	3.3	0	
D	16.2	3	0.33	4	3.8	4	0.46875	
Virtual	14.5	2.46875	0.22375	3.628125	3.109375	3.15	Factor k	1
		k-multi	iple output	3.628125	2.798839	3.109821	1.036607	

• Under the optimal solution, if K=1, school A is efficient; otherwise, it is not efficient.



Consider the following linear programming problem.

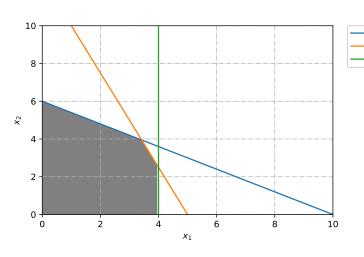
$$\max \quad c'x$$
s.t. $Ax \le b$,
$$x \ge 0$$
,
where $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, $c = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$, $A = \begin{bmatrix} 3 & 5 \\ 5 & 2 \\ 2 & 0 \end{bmatrix}$, and $b = \begin{bmatrix} 30 \\ 25 \\ 8 \end{bmatrix}$.

Which of the following statement(s) is(are) true?

- A) The corner points of the feasible region are (x1,x2) = (0,0), (0,3), (4,0), and (4,2.5).
- B) The optimal solution is (x1,x2) = (4,2.5).
- C) There is a redundant constraint.
- D) There are infinitely many optimal solutions.
- E) None of the above.



 $5x_1 + 2x_2 = 25$



a unique optimal solution

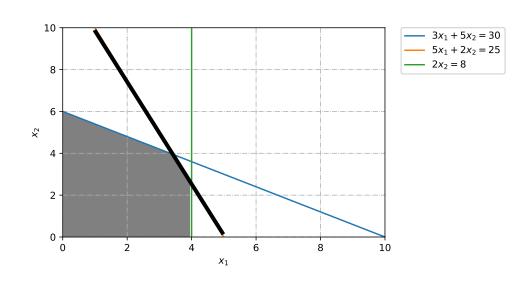
- Objective: max $3x_1 + x_2$
- Constraint 1: $3x_1 + 5x_2 \le 30$; Constraint 2: $5x_1 + 2x_2 \le 25$; Constraint 3: $2x_1 \le 8$; Constraint 4: $x_1, x_2 \ge 0$.
- Isoprofit line solution method/ Corner point solution method

x_1	x_2	$3x_1 + x_2$ (max)
0	0	0
4	0	12
4	2.5	14.5
3.42	3.95	14.2
0	6	6

The feasible region is shaped by all the constraints



$$\max \quad 6x_1 + 2.4x_2$$
s.t. $3x_1 + 5x_2 \le 30$
 $5x_1 + 2x_2 \le 25$
 $2x_1 \le 8$
 $x_1, x_2 \ge 0$

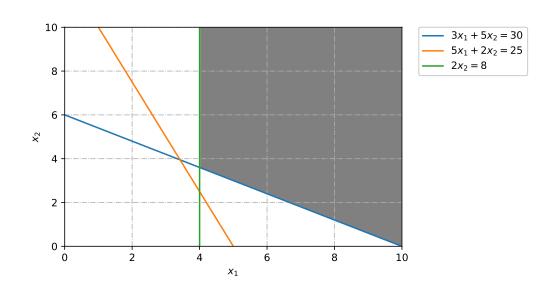


• Extension 1: Objective: max $6x_1 + 2.4x_2$

multiple optimal solutions



$$\begin{array}{ll} \max & 3x_1 + x_2 \\ \text{s.t.} & 3x_1 + 5x_2 \geq 30 \\ & 5x_1 + 2x_2 \geq 25 \\ & 2x_1 \geq 8 \\ & x_1, x_2 \geq 0 \end{array}$$



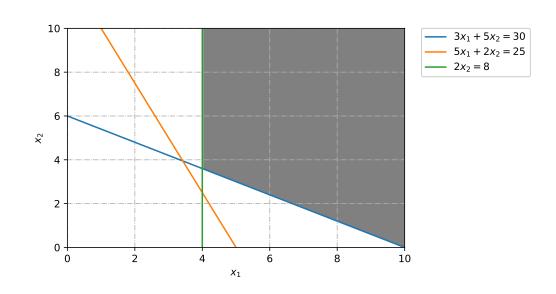
• Extension 2: Constrains: $Ax \ge b$

An unbounded feasible region. Unbounded solution



min
$$3x_1 + x_2$$

s.t. $3x_1 + 5x_2 \ge 30$
 $5x_1 + 2x_2 \ge 25$
 $2x_1 \ge 8$
 $x_1, x_2 \ge 0$



• Extension 3: Objective: $min \quad 3x_1 + x_2$ Constrains: $Ax \ge b$

An unbounded region. A bounded solution



• The SUPERSUDS CORPORATION is developing its marketing plans for next year's new products. For three of these products, the decision has been made to purchase a total of five TV spots for commercials on national television networks. The problem we will focus on is how to allocate the five spots to these three products, with a maximum of three spots (and a minimum of zero) for each product.

	Profit						
Number of TV Spots		Product					
	1	2	3				
0	0	0	0				
1	1	0	-1				
2	3	2	2				
3	3	3	4				



One Formulation with Auxiliary Binary Variables.

- A natural formulation would be to let x_1, x_2, x_3 be the number of TV spots allocated to the respective products.
- However, we cannot write a linear objective function in terms of these integer decision variables. For example, the profit function of product 1, $F_1(x_1)$, is not linear.
- Now we introduce an auxiliary binary variable y_{ij} to replace the original decision variables, where $y_{ij} = \begin{cases} 1 & \text{if } x_i = j \\ 0 & \text{otherwise.} \end{cases}$
- The values of y_{ij} can imply the value of x_i .
- For example, $y_{21} = 0$, $y_{22} = 0$, and $y_{23} = 1$ mean that $x_2 = 3$.
- These definitions are enforced by adding the constraints.

$$y_{i1} + y_{i2} + y_{i3} \le 1$$
 for $i = 1, 2, 3$.



• The resulting linear integer programming model is

Maximize
$$Z = y_{11} + 3y_{12} + 3y_{13} + 2y_{22} + 3y_{23} - y_{31} + 2y_{32} + 4y_{33}$$
, subject to

$$y_{11} + y_{12} + y_{13} \le 1$$

$$y_{21} + y_{22} + y_{23} \le 1$$

$$y_{31} + y_{32} + y_{33} \le 1$$

$$y_{11} + 2y_{12} + 3y_{13} + y_{21} + 2y_{22} + 3y_{23} + y_{31} + 2y_{32} + 3y_{33} = 5$$

each y_{ij} is a binary variable.



• The optimal solution is

	y11	y12	y13	y21	y22	y23	y31	y32	y33			
pofits		1	3	3	0	2	3	-1	2	4	total profits	
Decision												
Variables		0	1	0	0	0	0	0	0	1	7	
Constraints												
		1	1	1							1 <=	
					1	1	1				0 <=	
								1	1	1	1 <=	
		1	2	3	1	2	3	1	2	3	5 =	

$$y_{11} = 0,$$
 $y_{12} = 1,$ $y_{13} = 0,$ so $x_1 = 2$
 $y_{21} = 0,$ $y_{22} = 0,$ $y_{23} = 0,$ so $x_2 = 0$
 $y_{31} = 0,$ $y_{32} = 0,$ $y_{33} = 1,$ so $x_3 = 3.$



Another Formulation with Auxiliary Binary Variables.

ullet We now redefine the above auxiliary binary variables y_{ij} as follows:

$$y_{ij} = \begin{cases} 1 & \text{if } x_i \ge j \\ 0 & \text{otherwise.} \end{cases}$$

• Therefore,

$$x_i = 0$$
 \Rightarrow $y_{i1} = 0$, $y_{i2} = 0$, $y_{i3} = 0$, $x_i = 1$ \Rightarrow $y_{i1} = 1$, $y_{i2} = 0$, $y_{i3} = 0$, $x_i = 2$ \Rightarrow $y_{i1} = 1$, $y_{i2} = 1$, $y_{i3} = 0$, $x_i = 3$ \Rightarrow $y_{i1} = 1$, $y_{i2} = 1$, $y_{i3} = 1$, so $x_i = y_{i1} + y_{i2} + y_{i3}$

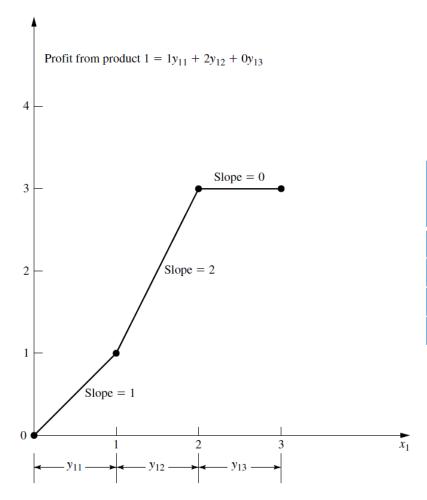
for i = 1, 2, 3.

These definitions are enforced by adding the constraints

$$y_{i2} \le y_{i1}$$
 and $y_{i3} \le y_{i2}$, for $i = 1, 2, 3$.



• The new definition of the y_{ij} also changes the objective function.



Number of TV Spots of Product 1	Profit of Product 1
0	0
1	1
2	3
3	3



The new linear integer programming problem is

Maximize
$$Z = y_{11} + 2y_{12} + 2y_{22} + y_{23} - y_{31} + 3y_{32} + 2y_{33}$$
, subject to

$$y_{12} - y_{11} \le 0$$

$$y_{13} - y_{12} \le 0$$

$$y_{22} - y_{21} \le 0$$

$$y_{23} - y_{22} \le 0$$

$$y_{32} - y_{31} \le 0$$

$$y_{33} - y_{32} \le 0$$

$$y_{11} + y_{12} + y_{13} + y_{21} + y_{22} + y_{23} + y_{31} + y_{32} + y_{33} = 5$$

each y_{ij} is a binary variable.



Solving this problem gives an optimal solution of

$$y_{11} = 1,$$
 $y_{12} = 1,$ $y_{13} = 0,$ so $x_1 = 2$
 $y_{21} = 0,$ $y_{22} = 0,$ $y_{23} = 0,$ so $x_2 = 0$
 $y_{31} = 1,$ $y_{32} = 1,$ $y_{33} = 1,$ so $x_3 = 3.$

	y11	y12	y13	y21	y22	y23	y31	y32	y33			
coeff.	1	. 2	-1	0	2	1	-1	3	2		total profits	
Decision												
Variables	1	. 1	. 0	0	0	0	1	1	1		7	
Constraints												
	1	. 1	. 1	1	1	1	1	1	1	5	=	5
		1								1	<=	1
					1					0	<=	0
								1		1	<=	1
			1							0	<=	1
						1				0	<=	0
									1	1	<=	1



- Comparison of two formulations:
 - Decision variables: They have the same number of binary variables (the prime consideration in determining computational effort for IP problems).
 - Constraints: They also have some *special structure:* constraints for *mutually exclusive alternatives* in the first model and constraints for *contingent decisions* in the second.
- The second model does have more functional constraints than the first.
- The first formulation may be better when the problem is on a large scale.



The Naïve estimator

- We would like to estimate the impact of smoking (D) for the general population with a random sample. In the sample, people can be classified along two dimensions. One is smoker versus non-smoker, and the other is high family income versus low family income. The number of people in each class and their average health condition are shown in the following tables.
- What is the Naïve estimator of the impact of smoking on health?

•	Why is the Naïve
	estimator biased?

Number	Smoker	Non-Smoker	Total
High Family Income	3	8	11
Low Family Income	7	3	10
Total	10	11	21

Health Condition	Smoker	Non-Smoker	Average
High Family Income	6	10	8.91
Low Family Income	5	8	5.90
Average	5.30	9.45	



The Naïve estimator

• If we use regression to estimate the impact of smoking based on the given data, what could be the most appropriate model specification (or functional form)?



The Naïve estimator

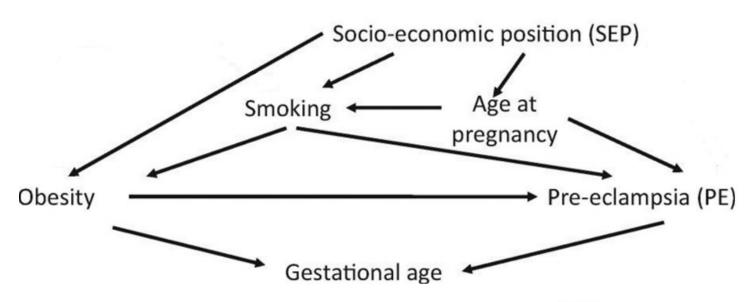
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Total	10	11	21

Health Condition	Smoker	Non-Smoker	Average
High Family Income	6	10	8.91
Low Family Income	5	8	5.90
Average	5.30	9.45	

If health conditions (Y^0, Y^1) are independent of smoking, conditioning on family income, what is the correct estimate of the impact of smoking on health for the general population?



- We are studying the mechanisms underlying the association between obesity and adverse perinatal (围生期) outcomes (Pre-eclampsia, 妊娠毒血症). We assume that the causal graph is correct [i.e. there are no other variables (nodes) or arrows that should be included].
- How to obtain a valid estimate of the causal effect of obesity on PE.



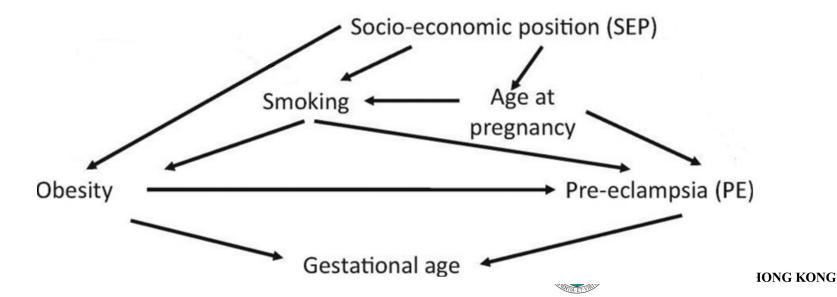


Notes for Back-door-blocking Strategy

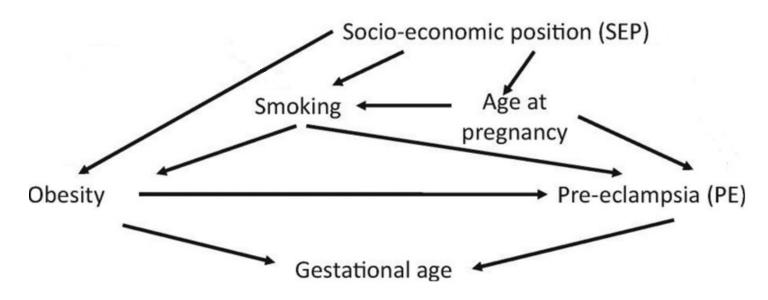
- Backdoor paths can transport non-causal associations and must be blocked to avoid spurious associations.
- The backdoor-blocking strategy allows us to identify an appropriate set of covariates for which we should adjust to estimate the causal effect of X on Y
 - Control for a variable = Adjust for a variable = Condition on a variable = using the variable in the regression model
- To estimate the causal effect of X on Y
- Remove all arrows starting from X
 - An unblocked path (without a collider on the way) starting from X transports causes from X to Y
- Identify all unblocked back-door paths
 - unblocked backdoor paths may introduce an association that is not due to the causal effect of X on Y
- Determine whether a set S of covariates is sufficient to block all backdoor paths



- Unblocked backdoor paths
 - Obesity ← Smoking ←SEP → Age at pregnancy → PE
 - Obesity \leftarrow SEP \rightarrow Age at pregnancy \rightarrow PE;
 - Obesity \leftarrow Smoking \leftarrow Age at pregnancy \rightarrow PE;
 - Obesity \leftarrow Smoking \rightarrow PE;
- There is also one blocked path
 - Obesity SEP- Smoking Age at pregnancy PE; this is blocked because age and SEP collide with smoking.

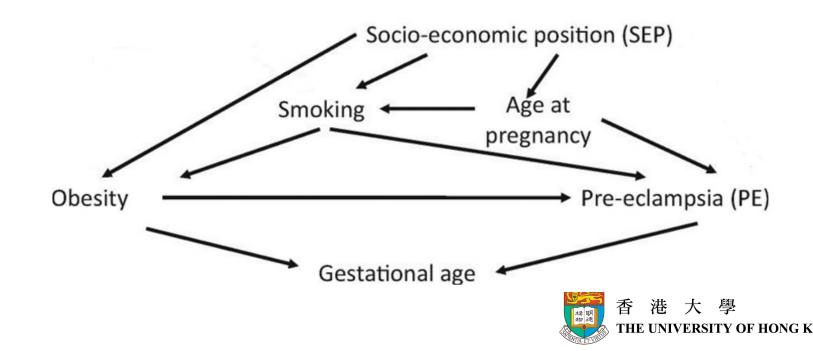


- To block the path Obesity ← Smoking → PE, we must condition "smoking"
- When we do that, a "spurious" association between Pregnancy age and SEP is generated. This new path is different from the above path, i.e., PE-Age at pregnancy-SEP-Obesity.





- However, blocking the original path will block the generated path. To do this, we can condition on either **Age at pregnancy** or **SEP**.
- To conclude, if we assume the causal graph is correct, then conditioning on Age at pregnancy (or SEP) and Smoking will provide a valid causal estimate.



- Should we condition on gestational?
- To do so potentially introduces a "spurious" association between Obesity and PE.
- Spurious association would be inverse (or consistent) and so this 'collider' bias could produce an effect estimate that is weaker (or stronger) than true positive effect.

