p8122_HW1

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Question 1a

```
individuals = data.frame(
    Individual = 1:8,
    Y0 = c(0, 1, 0, 1, 1, 0, 1, 0),
    Y1 = c(0, 0, 1, 0, 0, 1, 0, 0)
)

# calculate treatment effects
individuals = individuals |>
    mutate(TE = Y1 - Y0) # add a col TE

individuals |>
    kable()
```

Individual	Y0	Y1	TE
1	0	0	0
2	1	0	-1
3	0	1	1
4	1	0	-1
5	1	0	-1
6	0	1	1
7	1	0	-1
8	0	0	0

Question 1b

```
ACE = individuals |>
   summarise(ACE = mean(TE)) |>
   pull(ACE)

cat("The average causal effect (ACE) is:", ACE, "\n")
```

The average causal effect (ACE) is: -0.25

Question 1c

Table: Observed Outcomes under Assigned Treatment

```
individuals |>
  select(Individual, Y0, Y1, Assigned_Treatment, Observed) |>
  kable()
```

Individual	Y0	Y1	Assigned_Treatment	Observed
1	0	0	1	0
2	1	0	0	1
3	0	1	1	1
4	1	0	1	0
5	1	0	0	1
6	0	1	0	0
7	1	0	0	1
8	0	0	1	0

```
mean_treatment = individuals |>
  filter(Assigned_Treatment == 1) |>
  summarise(mean_treatment = mean(Observed)) |>
  pull(mean_treatment)

mean_control = individuals |>
  filter(Assigned_Treatment == 0) |>
  summarise(mean_control = mean(Observed)) |>
  pull(mean_control)

association = mean_treatment - mean_control
  cat("The association between treatment and outcome under specific treatment assignment is:", association
```

The association between treatment and outcome under specific treatment assignment is: -0.5

Interpretation:

The association between treatment and outcome under the specific treatment assignment is -0.5, which indicates individuals in the treatment group had worse health status (mean = 0.25) compared to the control group (mean = 0.75).

Question 1d

Table: Observed Outcomes under Random Assignment

```
individuals |>
   select(Individual, Y0, Y1, Random_Assigned_Treatment, Random_Assigned_Observed) |>
   kable()
```

Individual	Y0	Y1	Random_Assigned_Treatment	Random_Assigned_Observed
1	0	0	0	0
2	1	0	1	0
3	0	1	0	0
4	1	0	0	1
5	1	0	1	0
6	0	1	0	0
7	1	0	1	0
8	0	0	1	0

```
# Calculate the association for random assignment
mean_treatment_random = individuals |>
    filter(Random_Assigned_Treatment == 1) |>
    summarise(mean_treatment_random = mean(Random_Assigned_Observed)) |>
    pull(mean_treatment_random)

mean_control_random = individuals |>
    filter(Random_Assigned_Treatment == 0) |>
    summarise(mean_control_random = mean(Random_Assigned_Observed)) |>
    pull(mean_control_random)

association_random = mean_treatment_random - mean_control_random
cat("The association between treatment and outcome under random assignment is:", association_random, "\footnote{Control_random}")
```

The association between treatment and outcome under random assignment is: -0.25

```
# Compare with part 1b result (ACE)
cat("The ACE calculated in part 1b is:", ACE, "\n")
```

```
## The ACE calculated in part 1b is: -0.25
```

```
cat("Comparison: Association under random assignment:", association_random, "vs ACE:", ACE, "\n")
```

Comparison: Association under random assignment: -0.25 vs ACE: -0.25

Question 2a

The unit is the patient who is being treated for blood pressure management.

Question 2b

The treatment is the dosage of medication: 'High Dose' or 'Low Dose'

Question 2c

In this case, we could define two potential outcomes:

- Y_{high} : Blood pressure could be too high if the patient continues with the high dose of medication.
- Y_{low}: Blood pressure could be perfect (or controlled) under the low dose of medication.

Question 2d

```
outcomes_numeric = data.frame(
   Treatment = c("High Dose", "Low Dose"),
   Y_high = c(0, NA),
   Y_low = c(NA, 1)
)

# Calculate causal effect: Y_low - Y_high
causal_effect = outcomes_numeric$Y_low[2] - outcomes_numeric$Y_high[1]
causal_effect
```

[1] 1

Interpretation: Switching to the low dose results in better control of blood pressure.

Question 2e

SUTVA is plausible in this scenario: * No interference: The patient's outcome is not influenced by others' treatment. * Consistency: The low dose consistently results in better blood pressure control.

Question 2f

- SUTVA must be plausible to ensure that the potential outcome framework is valid.
- This ensures that the treatment's effect is not influenced by other external factors and the outcome is consistent across treatments.

Question 2g

- Probabilistic: No, the assignment is deterministic based on blood pressure readings.
- Individualistic: Yes, the treatment is specifically tailored for this patient.
- Unconfounded: Yes, treatment is based directly on the observed blood pressure levels, minimizing confounding.
- Controlled: Yes, the physician adjusts the dosage in a controlled manner based on the outcome.

Question 2h

```
set.seed(123)
random_assignment = sample(c("High Dose", "Low Dose"), size = 1, replace = TRUE)
cat("Random Assignment: The patient is assigned to:", random_assignment, "\n")
## Random Assignment: The patient is assigned to: High Dose
random_outcome = ifelse(random_assignment == "High Dose", "Too High", "Perfect")
cat("Observed outcome under random assignment:", random_outcome, "\n")
## Observed outcome under random assignment: Too High
n = 10  # Number of random assignments
results = data.frame(Trial = 1:n, Assigned_Treatment = NA, Outcome = NA)
for (i in 1:n) {
  assigned_treatment = sample(c("High Dose", "Low Dose"), size = 1, replace = TRUE)
  observed_outcome = ifelse(assigned_treatment == "High Dose", "Too High", "Perfect")
 results$Assigned_Treatment[i] = assigned_treatment
  results$Outcome[i] = observed_outcome
cat("\nTable: Outcomes under repeated random assignment\n")
##
## Table: Outcomes under repeated random assignment
print(results)
      Trial Assigned_Treatment Outcome
##
## 1
                    High Dose Too High
## 2
         2
                    High Dose Too High
## 3
         3
                    Low Dose Perfect
## 4
                    High Dose Too High
         4
## 5
                    Low Dose Perfect
        5
                     Low Dose Perfect
## 6
         6
## 7
         7
                    Low Dose Perfect
## 8
        8
                    High Dose Too High
## 9
                    High Dose Too High
         9
## 10
        10
                     Low Dose Perfect
```