

```
# Loading necessary libraries
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.4      v readr      2.1.5
```

```
## v forcats    1.0.0      v stringr    1.5.1
```

```
## v ggplot2    3.5.1      v tibble     3.2.1
```

```
## v lubridate  1.9.4      v tidyr      1.3.1
```

```
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()      masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(ggplot2)
```

```
vaccine_data <- read.csv("~/Desktop/appld stats/vaccine-1.csv") head(vaccine_data)
```

```
##      X Vaccine_Type Age_Group Antibody_Titer
## 1 1      Vaccine_A      Young      71.59287
## 2 2      Vaccine_A      Elderly      76.54734
## 3 3      Vaccine_A      Young     103.38062
## 4 4      Vaccine_A      Elderly      81.05763
## 5 5      Vaccine_A      Young      81.93932
## 6 6      Vaccine_A      Elderly     105.72597
```

```
str(vaccine_data)
```

```
## 'data.frame':      40 obs. of  4 variables:
```

```
## $ X              : int 1 2 3 4 5 6 7 8 9 10 ...
```

```
## $ Vaccine_Type   : chr "Vaccine_A" "Vaccine_A" "Vaccine_A" "Vaccine_A" ...
```

```
## $ Age_Group      : chr "Young" "Elderly" "Young" "Elderly" ...
```

```
## $ Antibody_Titer : num 71.6 76.5 103.4 81.1 81.9 ...
```

```
#Stating the Hypotheses # We will
```

```
test the following:
```

1

```
# 1.Main effect of Vaccine Type:
```

```
# H0: There is no difference in antibody titers between Vaccine_A and Vaccine_B. # H1: There is a difference in antibody titers between Vaccine_A and Vaccine_B.
```

```
# 2.Main effect of Age Group:
```

```
# H0: There is no difference in antibody titers between Young and Elderly groups.
```

```
# H1: There is a difference in antibody titers between Young and Elderly groups.
```

```
# 3.Interaction effect between Vaccine Type and Age Group:
```

```
# H0: There is no interaction effect on antibody titers.
```

H1: There is an interaction effect between Vaccine Type and Age Group on antibody titers.

Convert columns to factors

```
vaccine_data$Vaccine_Type <- as.factor(vaccine_data$Vaccine_Type)
```

```
vaccine_data$Age_Group <- as.factor(vaccine_data$Age_Group) # Perform two-way ANOVA
```

```
anova_model <- aov(Antibody_Titer ~ Vaccine_Type * Age_Group, data = vaccine_data) summary(anova_model)
```

##		Df	Sum Sq	Mean Sq	F value	Pr(>F)
## Vaccine_Type	1	505	505.1	2.215	0.145	
## Age_Group	1	60	60.4	0.265	0.610	
## Vaccine_Type:Age_Group	1	79	78.6	0.345	0.561	
## Residuals	36	8208	228.0			

Results and Interpretation

Based on the two-way ANOVA results:

Main Effect of Vaccine Type:

- $F(1, 36) = 2.215$, $p = 0.145$

- Since $p > 0.05$, the effect of vaccine type on antibody titer is not statistically significant.

Main Effect of Age Group:

- $F(1, 36) = 0.265$, $p = 0.610$

- Since $p > 0.05$, the effect of age group on antibody titer is not statistically significant.

#

Interaction Effect (Vaccine_Type × Age_Group):

- $F(1, 36) = 0.345$, $p = 0.561$

- Since $p > 0.05$, there is no significant interaction effect between vaccine type and age group.

Final Conclusion:

There is no statistically significant effect of vaccine type, age group, or their interaction on antibody titer.