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# ASSIGNMENT-02 STATISTICAL INFERENCE IN BIOENGINEERING

# BM23MTECH11006 P TEJA VENKATA RAMANA KUMAR

# Q1):-

```
import numpy as np
import pandas as pd
from scipy import stats
```

```
import numpy as np
import statsmodels.stats.api as sms
# Effect size
effect size = 2
# Significance level (alpha)
alpha = 0.05
# Desired power
power = 0.95
a)
# Calculate the minimum sample size
n=sms.tt ind solve power(effect size=effect size, nobs1=None, alpha=alpha, power=power,
ratio=1.0, alternative='two-sided')
print("Minimum sample size:", n)
Minimum sample size: 7.60895791594358
b)
alpha new = 0.01
# Calculate the sample size for the new alpha
n_new = sms.tt_ind_solve_power(effect_size=effect_size, nobs1=None, alpha=alpha_new,
power=power, ratio=1.0, alternative='two-sided')
print("Sample size for alpha = 0.01:", n_new)
Sample size for alpha = 0.01: 10.710971202419103
```

#### C)

power\_new = 0.99
# Calculate the sample size for the new power
n\_power = sms.tt\_ind\_solve\_power(effect\_size=effect\_size, nobs1=None, alpha=alpha,
power=power\_new, ratio=1.0, alternative='two-sided')
print("Sample size for power = 0.99:", n\_power)
Sample size for power = 0.99: 10.267583224566895

```
d)
n = 5
# Calculate the smallest detectable difference
detectable difference = sms.tt ind solve power(effect size=None, nobs1=n, alpha=alpha,
power=power_new, ratio=1.0, alternative='two-sided')
print("Smallest difference, |\mu c - \mu d| =", detectable_difference*10)
Smallest difference, |\mu c - \mu d| = 31.22095933284507
e)
desired difference = 2
# Calculate the probability of detecting the desired difference
power_for_desired_difference = sms.tt_ind_solve_power(effect_size=desired_difference, nobs1=n,
alpha=alpha, power=None, ratio=1.0, alternative='two-sided')
print("Probabilityofdetectingadifference=", power_for_desired_difference)
Probability of detecting a difference = 0.790542363607261
Q2):-
a)
desired difference = 2
n new=3
# Calculate the probability of detecting the desired difference
power_for_desired_difference = sms.tt_ind_solve_power(effect_size=desired_difference,
nobs1=n_new, alpha=alpha, power=None, ratio=1.0, alternative='two-sided')
print("n=3 then Power value =", power_for_desired_difference)
n=3 then Power value = 0.46264080923120443
b)
without delivery=np.array([138,155,121])
with_delivery=np.array([121,142,131])
# Set the significance level
alpha = 0.05
print(stats.ttest ind(without delivery, with delivery, equal var=True))
# Perform the two-sample t-test
t stat, p value = stats.ttest ind(without delivery, with delivery)
# Check if the p-value is less than the significance level
if p_value < alpha:</pre>
  print(f"The P value = \{p\_value\} < \alpha = 0.05. So Null hypotheses rejected.\nThe delivery mechanism
works in this experiment.")
  print(f"The P value = \{p\_value\} > \alpha = 0.05. So Null hypotheses can not be rejected.\nThe delivery
mechanism not works in this experiment.")
Ttest_indResult(statistic=0.5778319966569868, pvalue=0.5943464395237137)
The P value = 0.5943464395237137 > \alpha = 0.05. So Null hypotheses can not be rejected.
The delivery mechanism not works in this experiment.
```

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Q3):-
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```
a)
#Calcium concentration in seawater µo
#The average value of the sample of calcium concentration in animals µa
#Null hypotheses H0: μο = μα
#Alternative hypotheses Ha: µo ≠ µa
b)
data=np.array([28, 27, 29, 29, 30, 30, 31, 30, 33, 27, 30, 32,31])
population_mean = 32
print(stats.ttest_1samp(data,population_mean))
t_stat, p_value = stats.ttest_1samp(data, population_mean)
if p_value < 0.05:
  print(f"The P value is ={p_value} < 0.05. So H0 can be rejected.")
else:
  print(f"The P value is ={p value} > 0.05. So H0 can not be rejected.")
TtestResult(statistic=-4.501673049096729, pvalue=0.0007245394934862643, df=12)
The P value is =0.0007245394934862643 < 0.05. So H0 can be rejected.
c)
#Calcium concentration in seawater µo
# The average value of the sample of calcium concentration in animals µa
# Null hypotheses H0: μο > μa
# Alternative hypotheses Ha: μο < μa
print(stats.ttest_1samp(data, population_mean, alternative='greater'))
t_stat, p_value = stats.ttest_1samp(data, population_mean, alternative='greater')
if p value < 0.05:
  print(f"The P value is =\{p_value\} < 0.05. So H0 can be rejected. The species maintains a coelomic
calcium level greater than that of their environment.")
  print(f"The P value is ={p_value} > 0.05. So H0 cannot be rejected. The species maintains a
coelomic calcium levelless than that of their environment")
TtestResult(statistic=-4.501673049096729, pvalue=0.9996377302532569, df=12)
The P value is =0.9996377302532569 > 0.05. So H0 cannot be rejected. The species maintains a
coelomic calcium levelless than that of their environment
```

# Q4):-

male=np.array([220.1,218.6,229.6,228.8,222.0,224.1,226.5]) female=np.array([223.4,221.5,230.2,224.3,223.8,230.8]) a) The average value of the sample of Serum cholesterol concentrations in male µm The average value of the sample of Serum cholesterol concentrations in female uf Null hypotheses H0:  $\mu m = \mu f$ Alternative hypotheses Ha: µm ≠ µf b) **#Shapiro Test for checking Normality** print("shapiro male results:",stats.shapiro(male)) print("shapiro female results:",stats.shapiro(female)) \_,p\_value\_male=stats.shapiro(male) \_,p\_value\_female=stats.shapiro(female) if p\_value\_male > 0.05 and p\_value\_female > 0.05: print("Both male and female data appear to be normally distributed.") else: print("At least one of the datasets does not appear to be normally distributed.") print("By the Shpiro-Wilk normality test the P-value > Alpha. So Null hypotheses can not be rejected and the data was normally distributed.") shapiro male results: ShapiroResult(statistic=0.9439237117767334, pvalue=0.6742751598358154) shapiro female results: ShapiroResult(statistic=0.841784656047821, pvalue=0.13486963510513306) Both male and female data appear to be normally distributed. By the Shpiro-Wilk normality test the P-value > Alpha. So Null hypotheses can not be rejected and the data was normally distributed. c) This experiment has two independent data set groups: male & female The data are normally distributed. The aim was to compare the two distinct sample means. Two sample t-test was suitable for the above conditions. d) print(stats.ttest\_ind(male,female)) t\_stat, p\_value = stats.ttest\_ind(male, female) if p\_value < 0.05: print("Reject the null hypothesis. There is evidence that male and female turtles have different mean serum cholesterol concentrations.") else: print(f"the P-value={p\_value} > Alpha (0.05). There is no strong evidence to reject the null hypotheses. So, the mean of both mele and female serum cholesterol concentrations are same.") Ttest\_indResult(statistic=-0.6268108404512706, pvalue=0.543572996867541) the P-value=0.543572996867541 > Alpha (0.05). There is no strong evidence to reject the null hypotheses. So, the mean of both mele and female serum cholesterol concentrations are same.

### Q5):-

```
exam after coffee=np.array([72,64,81,90,73,54,87])
exam_after_tea=np.array([75,64,79,85,82,90,90])
a)
""" The same group of students gave the two sets of exams in the two different situations. The two
situations involved consuming tea or coffee before the exam. The given data looks suitable for a
sample t-test."""
# Shapiro Test for checking Normality
print(stats.shapiro(exam_after_coffee))
print(stats.shapiro(exam_after_tea))
_,p_value_coffee=stats.shapiro(exam_after_coffee)
_,p_value_tea=stats.shapiro(exam_after_tea)
if p value coffee > 0.05 and p value tea > 0.05:
  print("Both exam_after_coffee and exam after tea data appear to be normally distributed.")
else:
  print("At least one of the datasets does not appear to be normally distributed.")
print("By the Shpiro-Wilk normality test the P-value > Alpha. So Null hypotheses can not be rejected
and the data was normally distributed.")
ShapiroResult(statistic=0.9621812105178833, pvalue=0.8372160792350769)
ShapiroResult(statistic=0.918850302696228, pvalue=0.46051329374313354)
Both exam_after_coffee and exam after tea data appear to be normally distributed.
By the Shpiro-Wilk normality test the P-value > Alpha. So Null hypotheses can not be rejected and
the data was normally distributed.
# Levene Test for checking Variance
stats.levene(exam_after_coffee,exam_after_tea)
LeveneResult(statistic=0.7067307692307687, pvalue=0.4169679986979009)
b)
"""The paired observations should be independent of each other.

    The variance of the pairs was approximately equal, and the data were considered normally

distributed."""
c)
"""The average value of the sample of exam after coffee μc
The average value of the sample of exam after tea µt
Null hypotheses H0: (\mu c - \mu t) = 0
Alternative hypotheses Ha: (μm - μf) ≠ 0"""
print(stats.ttest_rel(exam_after_coffee, exam_after_tea))
t_stat, p_value = stats.ttest_rel(exam_after_coffee, exam_after_tea)
if p_value < 0.05:
  print("Reject the null hypothesis. There is evidence that coffee and tea have different effects on
exam performance.")
else:
  print(f"The P-value={p_value} > Alpha (0.05). Null hypotheses cannot be rejected. There is no
significant difference between the means of the students' exams after the coffee and after the tea.")
TtestResult(statistic=-1.2025868531803559, pvalue=0.27443586794212094, df=6)
```

The P-value=0.27443586794212094 > Alpha (0.05). Null hypotheses cannot be rejected. There is no significant difference between the means of the students' exams after the coffee and after the tea.

# Q6):-

import pandas as pd from scipy import stats BME=pd.DataFrame([78.49, 58.25, 76.33, 75.39, 73.00, 83.62, 67.81, 72.84, 82.02, 77.64, 71.44, 61.54, 66.20, 47.35, 82.45, 81.11, 72.84, 77.09, 69.41, 76.14, 72.05, 78.43, 59.73, 71.60, 82.22]) NonBME=pd.DataFrame([82.81, 78.36, 71.43, 55.71, 85.42, 26.17, 59.20, 62.38, 57.35, 70.45, 66.53, 74.57, 77.74, 73.85, 83.07, 83.66, 84.64, 85.65, 77.77, 69.47, 80.22, 74.72, 77.79]) BME.describe() count 25.000000 mean 72.599600 8.729939 std min 47.350000 **25%** 69.410000 appropriate? **50%** 73.000000 **75%** 78.430000 max 83.620000 BME.var() 76.211837 NonBME.describe() count 23.000000 mean 72.128696 13.487674 std min 26.170000 **25%** | 68.000000 **50%** 74.720000 **75%** 81.515000 max | 85.650000 NonBME.var() 181.917348

