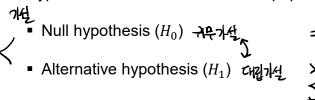
Hypothesis หมู

Hypothesis is a statement about a population parameter



● Testing युद्धभ्यः

Is a rule for which sample values the decision is made to accept \mathcal{H}_0 as true

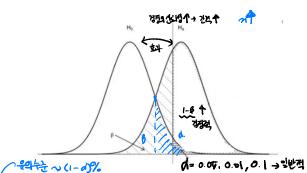


Inference Testing I

- Accept or reject
 - lacktriangle Rejection region : The area of the distribution where H_0 can be rejected
 - lacktriangle Acceptance region : The area of the distribution where H_0 can be accepted



- Statistical errors
 - Type I error (α) : also called the significance level, false positive
 - Type II error (β): false negative

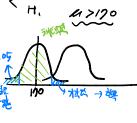


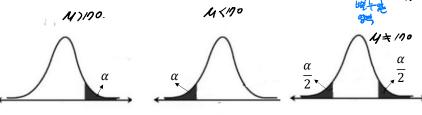


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物	Ho tive	Ho False
Ho reject	X→ de	
Ho acc	0	X>514 (M)

Inference Testing I

- Testing types
 - One-sided testing : left or right tailed rejection region H•
 - Two-sided testing : left and right tailed rejection regions

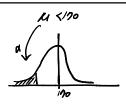




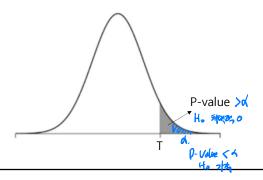
One sided testing

Two sided testing





- Test statistic
 - Test statistic : A statistic from a sample to make a decision about hypothesis on inference testing
- P-value





Inference Testing I

- Testing about μ
 - With large sample

$$H_0$$
: $\mu = \mu_0$

$$T = \frac{\bar{X} - \mu_0}{S / \sqrt{n}} \sim N(0,1)$$

• Rejection region:

$$H_1$$
: $\mu > \mu_0$, $T \ge Z_{\alpha}$

$$H_1$$
: $\mu < \mu_0$, $T \leq -Z_{\alpha}$

$$H_0$$
: $\mu \neq \mu_0$, $|T| \geq Z_{\frac{\alpha}{2}}$



- Testing about μ
 - With small sample

$$H_0: \mu = \mu_0$$

$$T = \frac{\bar{X} - \mu_0}{\sqrt[S]{\sqrt{n}}} \sim t_{(n-1)}$$

Rejection region:

$$H_1$$
: $\mu > \mu_0$, $T \ge \underline{t_{(n-1,\alpha)}}$

$$T \geq t_{(n-1,\alpha)}$$

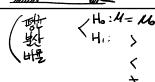
1性時.

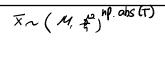
$$H_1$$
: $\mu < \mu_0$, $T \leq -t_{(n-1,\alpha)}$

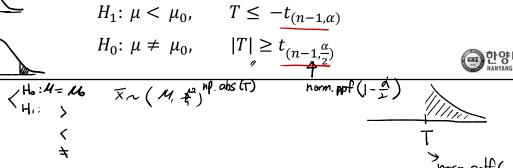
$$H_0$$
: $\mu \neq \mu_0$

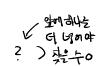
$$|T| \ge t_{(n-1,\frac{\alpha}{2})}$$









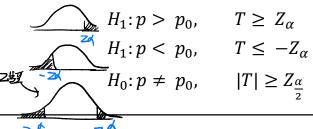


Inference Testing I

Testing about p

$$T = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0.(1 - p_0)}{n}}} \sim N(0,1)$$

Rejection region:





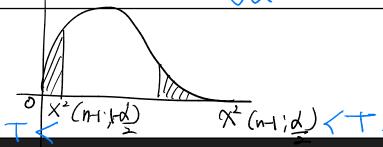
• Testing about σ^2 &

$$H_0$$
: $\sigma^2=\sigma_0^2$
$$T=rac{(n-1)s^2}{\sigma_0^2}\sim\chi^2_{(n-1)}$$
 কাইন্

Rejection region:

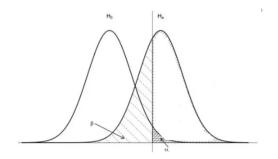
$$\begin{split} H_1: \sigma^2 &> \sigma_0^2, & T &\geq \chi^2_{(\alpha, n-1)} \\ H_1: \sigma^2 &< \sigma_0^2, & T &\leq \chi^2_{(1-\alpha, n-1)} \\ H_1: \sigma^2 &\neq \sigma_0^2, & T &\geq \chi^2_{\left(\frac{\alpha}{2}, n-1\right)} & T &\leq \chi^2_{\left(1-\frac{\alpha}{2}, n-1\right)} \end{split}$$





Inference Testing I

- Practice
 - Calculate the pooled estimator of variance with A, B groups.





In [7]: import numpy as np from scipy.stats import ttest_1samp, chi2, norm from statsmodels.stats.proportion import proportions_ztest

- .ttest_1samp
- -. T-test with one sample
- .proportions ztest
- -. Z-test for proportions



Inference Testing I

modata

.ttest_1samp(x, popmean, alternative)

- -. x : array like samples
- -. popmean: the mean of H0= Mo -. alternative: 'two-sided' or 'less' or 'greater'





with Jample +

.proportions_ztest(count, nobs, value, alternative)

- -. count : number of successes
- -. nobs: number of trials
- -. value: the value of H0 equal to the proportion
- -. alternative: 'two-sided' or 'smaller' or 'larger'



Inference Testing I

Practice

• With 'can' data, perform two-sided test whether H_0 : $\mu=100$ or not at $\alpha=0.05$



Solution

```
In [22]:
    def z_stat(x, popmean, alpha, alternative):
        # a/ternative = 'two-sided' or '/ess' or 'greater'
        z_val = (np.mean(x)-popmean)/(np.std(x, ddof=1)/np.sqrt(len(x)))
        print('Test statistic is {}'.format(np.round(z_val, 4)))

        if alternative =='two-sided':
            prnt('reject H0') if np.abs(z_val)> norm.ppf(1-alpha/2) else print('accept H0')
        elif alternative == 'less':
            print('reject H0') if z_val < norm.ppf(alpha) else print('accept H0')
        else:
            print('reject H0') if z_val > norm.ppf(1-alpha) else print('accept H0')

        z_stat(can, 100, 0.05, 'two-sided')

Test statistic is 0.2352
```

accept HO

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Inference Testing I

Practice

• Use only the first 20 observations in 'can' data, and test H_0 : $\mu=100$ vs. H_1 : $\mu>100$ at $\alpha=0.05$

```
In [26]: can_small = can[0:20]
print(can_small)

[101.8 101.5 102.6 101. 101.8 96.8 102.4 100. 98.8 98.1 98.8 98.
99.4 95.5 100.1 100.5 97.4 100.2 101.4 98.7]
```



Solution



```
In [36]: stat, pval = ttest_lsamp(can_small, popmean = 100, alternative = 'greater')
    print('Test statistic is {}'.format(np.round(stat,4)))
    print('p-value is {}'.format(np.round(pval,4)))
    print('HO is rejected') if pval < 0.05 else print('HO is accepted')</pre>
```

Test statistic is -0.5934 p-value is 0.72 HO is accepted



Inference Testing I

Practice

• A company found that failure rate of A product is 15%, and it could be improved if they changed the material of the product. They produced the A product with new material, and found 54 failures out of 400 pieces of the product. Then do you think the failure rate decreased at $\alpha=0.05$.



Solution

```
In [37]: stat, pval = proportions_ztest(54, 400, value = 0.15, alternative = 'smaller')
    print('Test statistic is {}'.format(np.round(stat,4)))
    print('p-value is {}'.format(np.round(pval,4)))
    print('HO is rejected') if pval < 0.05 else print('HO is accepted')

Test statistic is -0.8779
    p-value is 0.19
    HO is accepted</pre>
```



Inference Testing I

Practice

• Using 'can' data, and test H_0 : $\sigma^2 = 2.1$ vs. H_1 : $\sigma^2 > 2.1$ at $\alpha = 0.1$



Solution

```
In [39]: s20 =2.1
    n = len(can)
    s2 = np.var(can, ddof=1)
    stat = (n-1)*s2/s20
    pval = 1-chi2.cdf(stat,n-1)

print('Test statistic is {}'.format(np.round(stat,4)))
    print('p-value is {}'.format(np.round(pval,4)))
    print('HO is rejected') if pval < 0.05 else print('HO is accepted')

Test statistic is 87.0152
    p-value is 0.2515
    HO is accepted</pre>
```

