

# One-Sample Test for Proportion

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# Approximated One-Sample $Z$ Test for Proportion

# DM-TKA Example

In DM-TKR Data, there are 5 infective patients of total 78 patients, the sample proportion is  $6.41\% = 5/78$ . The infectiive probability in U.S. is about 1%. Does our sample differ from U.S. population?

# DM-TKA Example

1.  $\pi$  be infective probability in population
2. Random sample of  $n$  observations
3.  $X_i$  be random variables for each individual,  $i = 1, \dots, n$ ,

$$X_i = \begin{cases} 1, & \text{infection with probability } \pi, \\ 0, & \text{no infection with probability } 1 - \pi. \end{cases} \quad (1)$$

4.  $Y = \sum_{i=1}^n X_i$  has Binomial distribution  $(n, \pi)$

There is quite a variety of hypotheses about the DM population infective probability  $\pi$ .

# Hypothesis

- $H_0 : \pi = \pi_0 = 0.01,$
- $H_A : \pi \neq \pi_0.$

# Test Statistics

The observable sample proportion

$$\hat{\pi} = \frac{Y}{n} = \frac{\sum_{i=1}^n X_i}{n}, \quad (2)$$

The sample distribution of the sample proportion  $\hat{\pi}$

$$\hat{\pi} \sim N\left(\pi, \frac{\sqrt{\pi(1-\pi)}}{\sqrt{n}}\right) \quad (3)$$

The observed sample test statistic under  $H_0$

$$Z = \frac{(\hat{\pi} - \pi_0)}{\sqrt{(\pi_0(1 - \pi_0)/n)}} \sim N(0, 1) \quad (4)$$

(approximate distribution)

(It is called approximated Z test since it use the Z statistic.)

# Testing Hypothesis: $Z$ Value Method

The test statistic  $Z$  depends upon

1. The sample proportion  $\hat{\pi}$
2. The hypothesized target general population proportion  $\pi$
3. The population standard deviation,  $\sqrt{\pi(1 - \pi)}$ .

If the null hypothesis  $H_0$  is true, then the hypothesized population proportion  $\pi_0 = 0.01$  is equal to the population proportion,  $\pi$ .

# Testing Hypothesis: $Z$ Value Method

1. Prescribe Type I Error  $\alpha$
2.  $Z_{1-\alpha/2}$  be the corresponding percentile from  $N(0,1)$  such that  
 $P(Z < Z_\alpha) = \alpha$
3. Under  $H_0 : \pi = \pi_0$ , the observed test statistic

$$z = \frac{(\hat{\pi} - \pi_0)}{\sqrt{\pi_0(1 - \pi_0)} / \sqrt{n}}. \quad (5)$$



# Testing Hypothesis: $Z$ Value Method

1. For two-sided alternative test,  $H_A : \pi \neq \pi_0$

2. Reject the  $H_0$  when

$$|z| > Z_{1-\alpha/2}.$$

# Critical Value and Critical Region Methods

Given the significant level  $\alpha$

$$P(|Z| > Z_{1-\alpha/2}) = \alpha$$

$$P\left(\left|\frac{(\hat{\pi} - \pi)}{\sqrt{\pi(1-\pi)}/\sqrt{n}}\right| > Z_{1-\alpha/2}\right) = \alpha$$

$$P\left(\hat{\pi} < \pi - Z_{1-\alpha/2} \frac{\sqrt{\pi(1-\pi)}}{\sqrt{n}} \quad \text{or} \quad \hat{\pi} > \pi + Z_{1-\alpha/2} \frac{\sqrt{\pi(1-\pi)}}{\sqrt{n}}\right)$$

# Critical Value and Critical Region Methods

Under  $H_0 : \pi = \pi_0$ , we choose the two **critical values** for the two-sided Z test are

$$c_{\alpha,1} = \pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \quad (6)$$

and 
$$c_{\alpha,2} = \pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}. \quad (7)$$

We will reject the  $H_0$  based on the **critical region**

when  $\hat{\pi} = \frac{y}{n} = \frac{\sum_{i=1}^n x_i}{n}$

$$\hat{\pi} < c_{\alpha,1} = \pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}. \quad (8)$$

or 
$$\hat{\pi} > c_{\alpha,2} = \pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}. \quad (9)$$

# Confidence Interval Method

The two-sided  $(1 - \alpha) \times 100\%$  confidence interval of the population proportion  $\pi$  based on the sample statistic  $\hat{\pi}$ , and the two-sided alternative hypothesis  $H_A : \pi \neq \pi_0$ , is

$$P[|Z| < Z_{1-\alpha/2}] = 1 - \alpha$$

$$P\left[\left|\frac{\hat{\pi} - \pi}{\sqrt{\pi(1 - \pi)}/\sqrt{n}}\right| < Z_{1-\alpha/2}\right] = 1 - \alpha$$

$$P\left[\left|(\hat{\pi} - \pi)\right| < Z_{1-\alpha/2} \times \frac{\sqrt{\pi(1 - \pi)}}{\sqrt{n}}\right] = 1 - \alpha$$

$$P\left[\pi > \hat{\pi} - Z_{1-\alpha/2} \times \frac{\sqrt{\pi(1 - \pi)}}{\sqrt{n}}\right. \\ \left.\text{and } \pi < \hat{\pi} + Z_{1-\alpha/2} \times \frac{\sqrt{\pi(1 - \pi)}}{\sqrt{n}}\right] = 1 - \alpha.$$

# Asymptotic Confidence Interval Method

The two-sided  $(1 - \alpha) \times 100\%$  confidence interval of the population proportion  $\pi$  based on the sample statistic  $\hat{\pi}$ , is

$$\left( \hat{\pi} - Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}}, \hat{\pi} + Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}} \right). \quad (10)$$

We will reject the two-sided test when the two-sided  $(1 - \alpha) \times 100\%$  confidence interval of the population does not contain the hypothesized population proportion  $\pi_0$  under  $H_0$ .

# Asymptotic Confidence Interval Method

For  $H_0 : \pi = \pi_0$  versus  $H_A : \pi \neq \pi_0$ , we will reject the  $H_0$  when

$$\pi_0 < \hat{\pi} - Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}}, \quad (11)$$

or

$$\pi_0 > \hat{\pi} + Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}}. \quad (12)$$

That is when the hypothesized proportion  $\pi_0$  is below the lower or above the upper confident limit, we will reject  $H_0$ .

# Score (Wilson) Confidence Interval Method

For  $H_0 : \pi = \pi_0$  versus  $H_A : \pi \neq \pi_0$ , we can calculate the C.I. based on the **score confidence interval (Wilson's confidence interval)** as

$$\left[ \hat{\pi} - Z_{1-\alpha/2} \times \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \quad \hat{\pi} + Z_{1-\alpha/2} \times \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \right]. \quad (13)$$

# Likelihood Confidence Interval Method

1. For a sample proportion  $\hat{\pi}$  based on  $n$ , we can find the value  $\pi_{0L}$  and  $\pi_{0R}$  for the null hypothesis parameters that lead to the test statistic value  $z$

$$\frac{|\hat{\pi} - \pi_0|}{\sqrt{\pi_0(1 - \pi_0)/n}} = 1.96 \quad (14)$$

and a two-sided  $p$  – value = 0.05.

2. By solving the equation for  $\pi_0$ , this equation is quadratic in  $\pi_0$ .
3. Alternative, one can determine the limits by trial and error.



## $p$ -Value Method

1. We have collected the data and the observed sample statistic is  $\hat{\pi}$ .

2. Consider the two-sided hypothesis

$$H_0 : \pi = \pi_0 \text{ versus } H_A : \pi \neq \pi_0.$$

3. The observed two-sided  $Z$  test sample statistic is

$$z = \frac{(\hat{\pi} - \pi_0)}{(\sqrt{\pi_0(1 - \pi_0)} / \sqrt{n})}. \quad (15)$$

4. The  $p$ -value is defined as

- The  $p$ -value is the probability of obtaining a result as/or more extreme than you did by chance alone assuming the null hypothesis  $H_0$  is true.

## $p$ -Value Method

The  $p$ -value for two-sided test is calculated as

$$\begin{aligned} & P(|\bar{Y}| > |\bar{y}| \mid \pi = \pi_0) \\ = & P(|\pi - \pi_0| > |\hat{\pi} - \pi_0| \mid \pi = \pi_0) \\ = & P\left(\left|\frac{(\pi - \pi_0)}{\sqrt{\pi_0(1 - \pi_0)}/\sqrt{n}}\right| > \left|\frac{(\hat{\pi} - \pi_0)}{\sqrt{\pi_0(1 - \pi_0)}/\sqrt{n}}\right| \mid \pi = \pi_0\right) \\ = & P(|Z| > |z| \mid \pi = \pi_0) \\ = & 2[1 - P(Z \leq |z| \mid \pi = \pi_0)] \\ = & 2[1 - \Phi(|z|)], \end{aligned}$$

We will reject the two-sided null hypothesis  $H_0$  when  $p$ -value,  $2[1 - \Phi(|z|)]$ , is less than the significant level  $\alpha$ .

# DM-TKA Example

In DM-TKR Data, there are 5 infective patients of total 78 patients, the sample proportion is  $6.41\% = 5/78$ . The infectiive probability in U.S. is about 1%. Do our sample differ from U.S. population?

# DM-TKA Example

1. We wish to test the null hypothesis and alternative hypothesis are

$$H_0 : \pi = \pi_0 (= 0.01) \quad \text{versus} \quad H_A : \pi \neq \pi_0.$$

2. We have collected the data.
3. The observed sample proportion ( $\hat{\pi}$ , test statistic) is 6.4%.
4. Let the significant level  $\alpha = 0.05$ , and  $Z_{1-\alpha/2} = 1.960$ .

## DM-TKA Example

For two-sided test, the critical value (and critical region) for  $\hat{\pi}$  is

$$\begin{aligned}\pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} &= 0.01 - 1.960 \times \frac{0.01 * 0.99}{\sqrt{78}} \\ &= 0.01 + 1.960 \times 0.01127 \\ &= -0.0121\end{aligned}$$

and

$$\begin{aligned}\pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} &= 0.01 + 1.960 \times \frac{0.01 * 0.99}{\sqrt{78}} \\ &= 0.01 + 1.960 \times 0.01127 \\ &= 0.0322.\end{aligned}$$

Critical values,  $(c_{\alpha,1}, c_{\alpha,2})$ , are  $(-0.0121, 0.0322)$ .

## DM-TKA Example

We decide to reject the null hypothesis  $H_0$  if

$$\hat{\pi} < -0.0121 = \pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi(1-\pi)}}{\sqrt{n}}$$

or

$$\hat{\pi} > 0.0322 = \pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi(1-\pi)}}{\sqrt{n}}.$$

Now the observed sample proportion  $\hat{\pi} = 6.41\% > 0.0322$ , so we reject the null hypothesis.

# DM-TKA Example

1. The observed sample test statistic,  $z$ , is calculated as

$$z = \frac{(\hat{\pi} - \pi_0)}{\sqrt{\pi_0(1 - \pi_0)} / \sqrt{n}} = \frac{(0.0641 - 0.01)}{\sqrt{0.01 \cdot 0.09} / \sqrt{78}} = 4.80.$$

2. The observed sample test statistic,  $z$ , is 4.80 which is greater than the  $Z$  critical value,  $Z_{1-\alpha/2} = 1.960$ .
3. So we reject the null hypothesis  $H_0$ .

## DM-TKA Example

1. The two-sided  $(1 - \alpha) \times 100\%$  confidence interval for DM population proportion  $\pi$  based on the sample statistic,  $\hat{\pi}$ , can be calculated as

$$\left( \hat{\pi} - Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}}, \hat{\pi} + Z_{1-\alpha/2} \times \frac{\sqrt{\hat{\pi}(1 - \hat{\pi})}}{\sqrt{n}} \right) \\ = (0.00975, 0.1185).$$

2. The  $(1 - \alpha) \times 100\%$  confidence interval for DM population proportion  $\pi$  is  $(0.00975, 0.1185)$ .
3. This interval does contain  $\pi_0 = 0.01$ .
4. So we do not reject the null hypothesis  $H_0$ .



# DM-TKA Example

1. The  $p$ -value based on the observed sample test statistic,  $z = 4.80$ , can be calculated as

$$2[1 - P(\hat{\pi} > |\bar{x}|)] = 2[1 - \Phi(|z|)] = \Phi(4.80) < 0.0001.$$

2. The  $p$ -value,  $< 0.0001$ , is less than the significant level  $\alpha = 0.05$ .
3. So we reject the null hypothesis.

## DM-TKA Example: R

```
> y<-5; n<-78 # assign y and n in binomial
> alpha<-0.05 # assign significant level alpha
> pihat<-y/n # sample proportion
> pihat
[1] 0.06410256
> qihat<-1-pihat
> se0<-sqrt(pi0*(1-pi0)/n) # s.e. under H0
> se1<-sqrt(pihat*qihat/n) # s.e. Under HA
> Z1alpha<-qnorm(1-alpha/2) # Z_{1-alpha/2} quantile
> ztest<-(pihat-pi0)/se0 # sample Z test statistic
> ztest
[1] 4.802281
```

## DM-TKA Example: R

```
> crit1<-pi0-Z1alpha*se0 # critical value c_{alpha,1}
> crit2<-pi0+Z1alpha*se0 # critical value c_{alpha,2}
> crit1
[1] -0.01208098
> crit2
[1] 0.03208098
> Z.CI.L<-pihat-Z1alpha*se1 # C.I. Lower
> Z.CI.U<-pihat+Z1alpha*se1 # C.I. Upper
> Z.CI.L
[1] 0.009745922
> Z.CI.U
[1] 0.1184592
```

# Power of One-sample $Z$ Test for Proportion

1. The first thing is to decided the possible  $\pi$  value under  $H_A$ , since different  $\pi$  value under  $H_A$  will have different power.
2. Suppose in the DM-TKR example, we have  $\pi_0 = 0.01$  and  $\pi = \pi_A = 0.005$  or  $\pi_A = 0.05$  under  $H_A$ .
3. What is power of the two-sided test?

$$\text{power} = P(\text{reject } H_0 \mid H_A \text{ is true})$$

4. Power is the probability of making the correct decision when the null hypothesis is not true. Specially,

$$\text{power} = 1 - \beta = P(\text{reject } H_0 : \pi = \pi_0 \mid H_A \text{ is true}).$$

# Power

If  $H_A : \pi = \pi_A < \pi_0$  is true, then we will reject  $H_0$  when

$$\hat{\pi} < c_\alpha = \pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \quad (16)$$

or

$$\frac{(\hat{\pi} - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} < \frac{(\pi_0 - Z_{1-\alpha/2}(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}) - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}}. \quad (17)$$

# Power

If  $H_A : \pi = \pi_A < \pi_0$  is true,

$$\text{power} = 1 - \beta = P(\hat{\pi} < c_\alpha \mid \pi = \pi_A) \quad (18)$$

$$= P\left(\hat{\pi} < \pi_0 - Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \mid \pi = \pi_A\right) \quad (19)$$

$$= P\left(Z = \frac{(\hat{\pi} - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} < \frac{(\pi_0 - Z_{1-\alpha/2}(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}) - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} \mid \pi = \pi_A\right)$$

$$= P\left(Z < \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}} \left(-Z_{1-\alpha/2} + \frac{(\pi_0 - \pi_A)}{\sqrt{\pi_0(1-\pi_0)}/n}\right) \mid \pi = \pi_A\right)$$

$$= \Phi\left(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}} \left(-Z_{1-\alpha/2} + \frac{(|\pi_0 - \pi_A|)}{\sqrt{\pi_0(1-\pi_0)}/n}\right)\right). \quad (20)$$

# Power

If  $H_A : \pi = \pi_A > \pi_0$  is true, then we will reject  $H_0$  when

$$\hat{\pi} > c_\alpha = \pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}, \quad (21)$$

or

$$\frac{(\hat{\pi} - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} > \frac{(\pi_0 + Z_{1-\alpha/2}(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}) - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}}. \quad (22)$$

# Power

If  $H_A : \pi = \pi_A > \pi_0$  is true,

$$\text{power} = 1 - \beta = P(\hat{\pi} > c_\alpha \mid \pi = \pi_A) \quad (23)$$

$$= P\left(\hat{\pi} > \pi_0 + Z_{1-\alpha/2} \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}} \mid \pi = \pi_A\right) \quad (24)$$

$$= P\left(Z = \frac{(\hat{\pi} - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} > \frac{(\pi_0 + Z_{1-\alpha/2}(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{n}}) - \pi_A)}{\sqrt{\pi_A(1-\pi_A)}/\sqrt{n}} \mid \pi = \pi_A\right)$$

$$= P\left(Z > \frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}} \left( + Z_{1-\alpha/2} + \frac{(\pi_0 - \pi_A)}{\sqrt{\pi_0(1-\pi_0)}/n} \right) \mid \pi = \pi_A\right)$$

$$= 1 - \Phi\left(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}} \left( + Z_{1-\alpha/2} + \frac{(\pi_0 - \pi_A)}{\sqrt{\pi_0(1-\pi_0)}/n} \right)\right)$$

$$= \Phi\left(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}} \left( - Z_{1-\alpha/2} + \frac{(|\pi_0 - \pi_A|)}{\sqrt{\pi_0(1-\pi_0)}/n} \right)\right). \quad (25)$$



# Power

For one-sided test, we use  $Z_{1-\alpha}$  instead of  $Z_{1-\alpha/2}$ .

# Power

So the power of the two-sided test  $H_0 : \pi = \pi_0$  versus  $H_A : \pi \neq \pi_0$  for the specific alternative  $\pi = \pi_A$ , where the underlying distribution is approximately normal and the population variance  $\sigma^2$  is estimated as  $\pi_A(1 - \pi_A)$ , is given exactly by

$$\text{power} = \Phi \left[ \frac{\sqrt{\pi_0(1 - \pi_0)}}{\sqrt{\pi_A(1 - \pi_A)}} \left( -Z_{1-\alpha/2} + \frac{(|\pi_0 - \pi_A|)}{\sqrt{\pi_0(1 - \pi_0)/n}} \right) \right]. \quad (26)$$

# Power

1. The power formula has nothing to do with observed sample statistic  $z$ , however, the power depends on  $\pi_A$  and its variance  $\pi_A(1 - \pi_A)$ .
2. If we consider  $H_A : \pi = \hat{\pi}$ , that is, we calculated the power after the study, this is sometime called **post-hoc power**. It is not recommend by many statisticians.
3. For one-sided test, we use  $Z_{1-\alpha}$  (instead of  $Z_{1-\alpha/2}$ ).

## DM-TKA Example

1. Suppose  $H_A : \pi = \pi_A$ , for example,  $\pi_A = 0.05$ . We have power

$$\Phi\left(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}}\left(\frac{(|\pi_0 - \pi_A|)}{\sqrt{\pi_0(1-\pi_0)/n}}\right)\right) = 0.76612.$$

2. If we assume  $H_A : \pi_A = 0.09$ , then the power is

$$\Phi\left(\frac{\sqrt{\pi_0(1-\pi_0)}}{\sqrt{\pi_A(1-\pi_A)}}\left(-Z_{1-\alpha/2} + \frac{(|\pi_0 - \pi_A|)}{\sqrt{\pi_0(1-\pi_0)/n}}\right)\right) = 0.9630.$$

# DM-TKA Example

The power depends on the variance of  $\pi_A$ , so the direction of power of the approximate Z test for proportion is not as clear as normal distribution.

Note: the power formula has nothing to do with observed sample statistic  $z$ . If we consider  $H_A : \pi = \hat{\pi}$ , that is, we calculated the power after the study, this is sometime called **post-hoc power**. It is not recommend by many statisticians.

## DM-TKA Example: R

```
> # two-tailed power
> power.prop.two.sided.Z.test<-function(pi0,pia,alpha,n)
  { power<-sqrt((pi0*(1-pi0))/(pia*(1-pia)) )
    *(-qnorm(1-alpha/2)
      +abs(pi0-pia)*sqrt(n)/sqrt(pi0*(1-pi0)))
    power<-pnorm(power)
    cat("power = ",power,"\n")
  }
> power.prop.two.sided.Z.test(0.01,0.05,0.05,78)
power = 0.7661206
> power.prop.two.sided.Z.test(0.01,0.09,0.05,78)
power = 0.9630652
```

# **Exact Small-Sample Inference**

## **Exact Test for Proportion**

# Exact Test for Proportion

1. The approximate  $Z$  test procedure to test the hypothesis  $H_0 : \pi = \pi_0$  versus  $H_A : \pi \neq \pi_0$  depends on the assumptions is only true if  $n\pi_0(1 - \pi_0) \geq 5$ .
2. With modern computational power, it is not necessary to rely on large-sample approximation for the distribution of statistics such as  $\hat{\pi}$ .
3. Tests and confidence intervals can use the binomial distribution directly rather than its normal approximation. Such inferences occur naturally for small samples, but apply for any  $n$ .



# Exact Test for Proportion

1. We will base our test on **exact** binomial probabilities.
2. Let  $Y \sim \text{Bin}(n, \pi_0)$  under  $H_0$ .
3. Let  $\hat{\pi} = y/n$ , be the observed sample proportion.
4. The computation of the  $p$ -value depends on whether  $\hat{\pi} \leq \pi_0$  or  $\hat{\pi} \geq \pi_0$ .

# $p$ -value of the Exact Test for Proportion

1. If  $\hat{\pi} \leq \pi_0$ , then

$$p - \text{value}/2 = P(Y \leq y \text{ successes in } n \text{ trials} \mid H_0) \quad (27)$$

$$= \sum_{k=0}^y \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k} \quad (28)$$

2. If  $\hat{\pi} \geq \pi_0$ , then

$$p - \text{value}/2 = P(Y \geq y \text{ successes in } n \text{ trials} \mid H_0) \quad (29)$$

$$= \sum_{k=y}^n \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k} \quad (30)$$

# $p$ -value of the Exact Test for Proportion

1. If  $\hat{\pi} \leq \pi_0$ , then

$$p - \text{value} = 2 \times P(Y \leq y) \quad (31)$$

$$= \min \left[ 2 \sum_{k=0}^y \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k}, \quad 1 \right] \quad (32)$$

2. If  $\hat{\pi} \geq \pi_0$ , then

$$p - \text{value} = 2 \times P(Y \geq y) \quad (33)$$

$$= \min \left[ 2 \sum_{k=y}^n \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k}, \quad 1 \right] \quad (34)$$

## ***p*-value of the Exact Test for Proportion**

We illustrate by testing  $H_0 : \pi = 0.5$  against  $H_A : \pi \neq 0.5$  for the survey results,  $y = 0$ , with  $n = 25$ . We noted that the score statistic equals  $z = -5.0$ . The exact *p*-value for this statistic, based on the null  $\text{Bin}(25, 0.5)$  distribution, is

$$P(|z| \geq 5.0) = P(Y = 0 \text{ or } Y = 25) = 0.5^{25} + 0.5^{25} = 0.000000006.$$

# C.I. of the Exact Test for Proportion

1. The exact  $100(1 - \alpha)\%$  confidence intervals consists of all  $\pi$  for which  $p$ -values exceed  $\alpha$  in exact binomial tests.
2. The best known interval (**Clopper and Person**, 1934) uses the tail method for forming confidence intervals.
3. It requires each one-sided  $p$ -value to exceed  $\alpha/2$ .

## C.I. of the Exact Test for Proportion

The lower and upper endpoints are the solutions in  $\pi_0$  to the equations

$$\sum_{k=y}^n \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k} = \alpha/2$$

$$\text{and} \quad \sum_{k=0}^y \binom{n}{k} \pi_0^k (1 - \pi_0)^{n-k} = \alpha/2, \quad (35)$$

except that the lower bound is 0 when  $y = 0$  and the upper bound is 1 when  $y = n$ .

# C.I. of the Exact Test for Proportion

1. The **Clopper and Person** confidence interval equals

$$\left[1 + \frac{n - y + 1}{y F_{2y, 2(n-y+1), \alpha/2}}\right]^{-1} < \pi < \left[1 + \frac{n - y + 1}{(y + 1) F_{2(y+1), 2(n-y), (1-\alpha/2)}}\right]^{-1}, \quad (36)$$

where  $F_{a,b,c}$  denotes the  $c$  quantile from the  $F$  distribution with degrees of freedom  $a$  and  $b$ .

2. Example,

When  $y = 0$  with  $n = 25$ , the **Clopper-Pearson 95% confidence interval** for  $\pi$  is  $(0, 0.137)$ .

## DM-TKA Example: Exact Test

The exact  $(1 - \alpha) \times 100\%$  confident interval is (0.0211, 0.1433). The exact two-sided test  $p$ -value is  $0.001152 < \alpha = 0.05$  (SAS:  $p = 0.0023$ ). We reject the null hypothesis  $H_0$  based on the exact confidence interval and  $p$ -value.



# DM-TKA Example: Exact Test with R `binom.test`

```
> binom.test(x=5,n=78,p=0.01,alternative = c("two.sided"),  
  conf.level = 0.95)
```

Exact binomial test

data: 5 and 78

number of successes=5, number of trials=78, p-value=0.001152

alternative hypothesis:

true probability of success is not equal to 0.01

95 percent confidence interval:

0.02113972 0.14328760

sample estimates:

probability of success

0.06410256

# DM-TKA Example: Exact C.I. and Asymptotic C.I. with R

```
> library(Hmisc)
> help(binconf)
> binconf(x=5,n=78,alpha=0.05,method="all")
```

	PointEst	Lower	Upper
Exact	0.06410256	0.021139720	0.1432876
Wilson	0.06410256	0.027689315	0.1414360
Asymptotic	0.06410256	0.009745922	0.1184592

# DM-TKA Example: R

```
> help(prop.test)
> prop.test(x=5,n=78,p=0.01,alternative = c("two.sided"),
  correct=F,conf.level = 0.95)

1-sample proportions test without continuity correction
data: 5 out of 78, null probability 0.01
X-squared = 23.0619, df = 1, p-value = 1.569e-06
alternative hypothesis: true p is not equal to 0.01
95 percent confidence interval:
 0.02768932 0.14143595
sample estimates:
      p
0.06410256
```

# DM-TKA Example: Exact Test with R

Warning message:

Chi-squared approximation may be incorrect in:

```
prop.test(x=5, n=78, p=0.01, alternative=c("two.sided"),
```

# DM-TKA Example: SAS

```
title "FREQ: One-sample Z test  
      for proportion with 95% C.I." ;  
proc freq data=dmtakcsv order=data ;  
    exact binomial ;  
    tables infect / bin( p=0.01) ;  
run;
```

# DM-TKA Example: SAS

```
title "Categorical Data:  
    Graphics of One-sample";  
proc gchart data=dmtakcsv ;  
    vbar infect / discrete ;  
    hbar infect / discrete ;  
    pie infect / discrete ;  
    pie infect / discrete  
        explode=1 slice=arrow percent=inside ;  
run;
```

# DM-TKA Example: SAS

The FREQ Procedure

infect			Cumulative	Cumulative
	Frequency	Percent	Frequency	Percent
-----				
1	5	6.41	5	6.41
0	73	93.59	78	100.00

# DM-TKA Example: SAS

Binomial Proportion for infect = 1

-----

Proportion (P)	0.0641
----------------	--------

ASE	0.0277
-----	--------

95% Lower Conf Limit	0.0097
----------------------	--------

95% Upper Conf Limit	0.1185
----------------------	--------



# DM-TKA Example: SAS

Exact Conf Limits

95% Lower Conf Limit	0.0211
----------------------	--------

95% Upper Conf Limit	0.1433
----------------------	--------

# DM-TKA Example: SAS

Test of  $H_0$ : Proportion = 0.01

ASE under $H_0$	0.0113
Z	4.8023
One-sided Pr > Z	<.0001
Two-sided Pr >  Z	<.0001

# DM-TKA Example: SAS

Exact Test

One-sided Pr  $\geq$  P 0.0012

Two-sided = 2 \* One-sided 0.0023

Sample Size = 78