

Two-sided Exact Tests and Matching Confidence Intervals for Discrete Data

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Motivating Example 1: Fisher's exact Test for 2×2 Table

	Homozygous for CCR5 Δ 32 mutation		Wild Type or Heterozygous for CCR5 Δ 32 mutation	
Abdominal Pain	4	(26.7%)	50	(8.1%)
No Abdom. Pain	11	(73.3%)	569	(91.9%)

Relationship of CCR5 Δ 32 mutation (genetic recessive model) to Early Symptoms with West Nile Virus Infection
(from Lim, et al, J Infectious Diseases, 2010, 178-185)

Step 1: Create 2 by 2 Table

```
> abdpain<-matrix(c(4,50,11,569),2,2,  
+   dimnames=list(c("Abdominal Pain","No Abdom. Pain"),  
+   c("Homo","WT/Hetero")))  
> abdpain
```

	Homo	WT/Hetero
Abdominal Pain	4	50
No Abdom. Pain	11	569

Analysis in R 2.11.1, stats package

Step 2: Run test

```
> fisher.test(abdpain)
```

```
Fisher's Exact Test for Count Data
```

```
data:  abdpain
```

```
p-value = 0.03166
```

```
alternative hypothesis: true odds ratio is not equal to 1
```

```
95 percent confidence interval:
```

```
0.9235364 14.5759712
```

```
sample estimates:
```

```
odds ratio
```

```
4.122741
```

Test-CI Inconsistency

Problem: Test rejects but confidence interval includes odds ratio of 1.

- ▶ Same problem in:
 - ▶ R (fisher.test), Version 2.11.1,
 - ▶ SAS (Proc Freq), Version 9.2 and
 - ▶ StatXact, (StatXact 8 Procs).
- ▶ In all 3: One and only one exact confidence for odds ratio for the 2 by 2 table is given, AND
- ▶ the confidence interval is not an **inversion** of the usual two-sided Fisher's exact test.
 - ▶ (Test defined the same way in all 3 programs).

Example 2: One Sample Binomial Test

Observe 10 out of 100 from a simulation. Is this significantly different from a true proportion of 0.05?

```
> binom.test(10,100,p=0.05)
```

```
Exact binomial test
```

```
data: 10 and 100
```

```
number of successes = 10, number of trials = 100, p-value = 0.03411
```

```
alternative hypothesis: true probability of success is not equal to 0.05
```

```
95 percent confidence interval:
```

```
0.04900469 0.17622260
```

```
sample estimates:
```

```
probability of success
```

```
0.1
```

Example 3: Two Sample Poisson Test

If we observe rates 2/17887 (about 11.2 per 100,000) for the standard treatment and 10/20000 (50 per 100,000) for new treatment, do these two groups significantly differ by exact Poisson rate test?

```
> poisson.test(c(10,2),c(20000,17877))
```

Comparison of Poisson rates

```
data: c(10, 2) time base: c(20000, 17877)
count1 = 10, expected count1 = 6.336, p-value = 0.04213
alternative hypothesis: true rate ratio is not equal to 1
95 percent confidence interval:
 0.952422 41.950915
sample estimates:
rate ratio
 4.46925
```

What is happening in the examples?

- ▶ In each example, we used an exact test and an exact confidence interval, **but**,
- ▶ the confidence interval is **not** an inversion of the test.

What is happening in the examples?

- ▶ In each example, we used an exact test and an exact confidence interval, **but**,
- ▶ the confidence interval is **not** an inversion of the test.
- ▶ Definition: **confidence interval by inversion of (a series of) tests** = all parameter values that fail to reject point null hypothesis.

Definition: Inversion of Family of Tests

- ▶ Consider a series of tests, indexed by β_0
- ▶ Let \mathbf{x} be data.
- ▶ Let $p_{\beta_0}(\mathbf{x})$ be p-value for testing the following hypotheses:

$$H_0 : \quad \beta = \beta_0$$

$$H_1 : \quad \beta \neq \beta_0$$

Then the inversion confidence set is

$$C(\mathbf{x}, 1 - \alpha) = \{\beta : p_{\beta}(\mathbf{x}) > \alpha\}$$

Cannot have test-confidence set inconsistency with inversion confidence set.

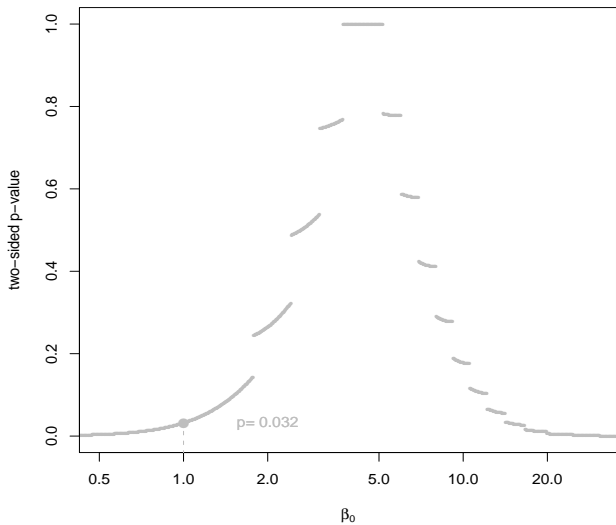


Figure: CCR5 data: Abdominal Pain, usual two-sided Fisher's exact p-values

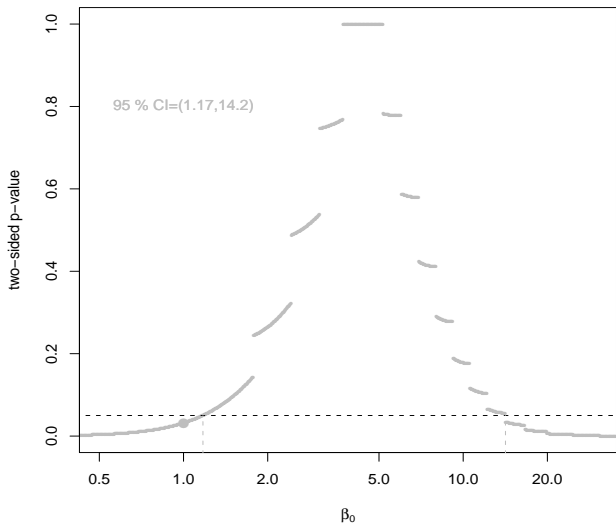


Figure: CCR5 data: Abdominal Pain, 95 % inversion confidence interval to usual two-sided Fisher's exact

Another two-sided Fisher's exact Test

- ▶ Define p-value as 2 times minimum of the one-sided Fisher's exact p-values.
- ▶ Inversion of **that** two sided Fisher's exact is the usual exact confidence intervals.
- ▶ Call it **Central Fisher's exact Test**

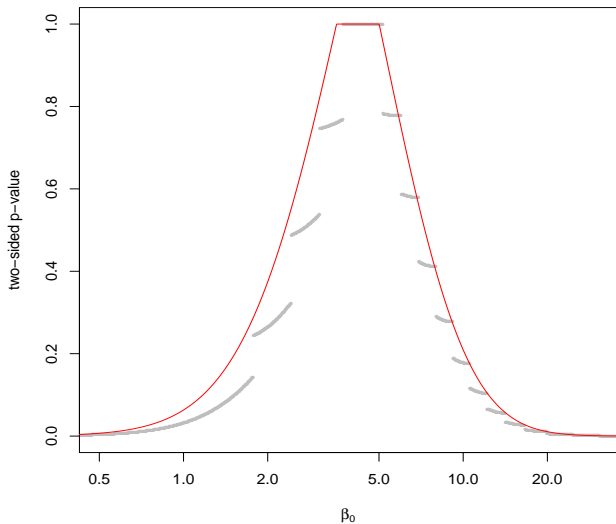


Figure: CCR5 data: Abdominal Pain, gray= usual two-sided Fisher's exact p-values, red=twice minimum one-sided p-values

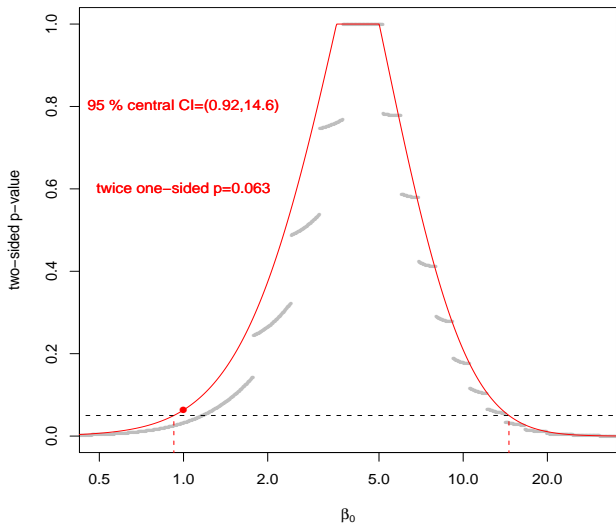


Figure: CCR5 data: Abdominal Pain, 95 % central confidence intervals

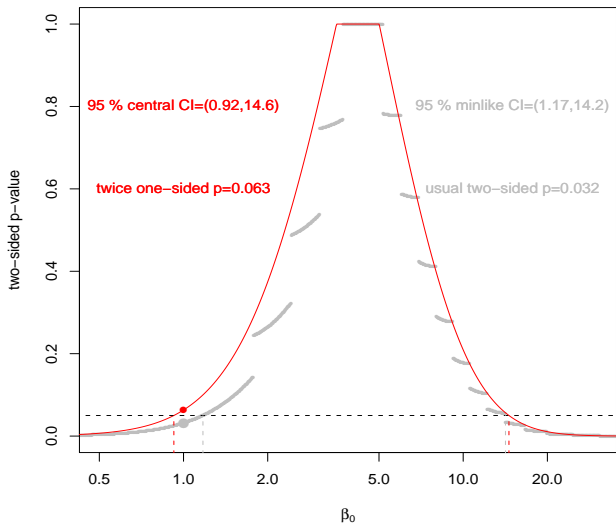


Figure: CCR5 data: Abdominal Pain, 95 % central confidence intervals

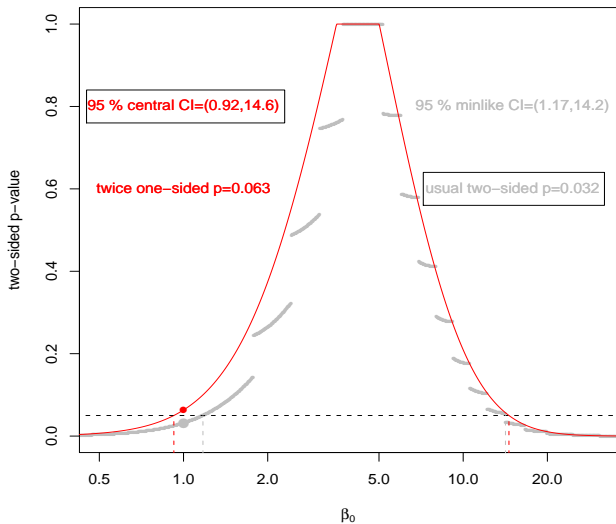


Figure: CCR5 data: Abdominal Pain, 95 % central confidence intervals

3 Ways to Calculate Two-sided p-values

central: 2 times minimum of one-sided p-values,

minlike: sum of probabilities of outcomes with likelihoods less than or equal to observed.

$$p_m(x) = \sum_{X:f(X)\leq f(x)} f(X)$$

blaker: take smaller observed tail and add largest probability on the opposite tail that does not exceed observed tail.

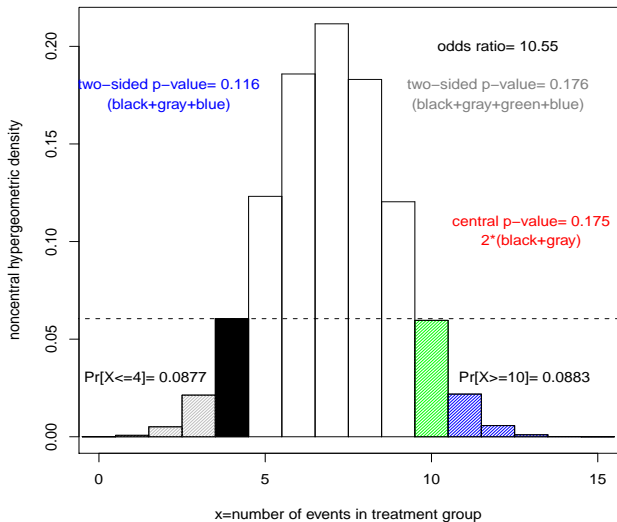


Figure: CCR5 data: Abdominal Pain

Solution: Use “Matching” Confidence Intervals

Smallest confidence interval that contains all parameters that fail to reject.

```
> library(exact2x2)
Loading required package: exactci
> fisher.exact(abdpain)
```

```
Two-sided Fisher's Exact Test (usual method using minimum likelihood)
```

```
data: abdpain
p-value = 0.03166
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 1.1734 14.1659
sample estimates:
odds ratio
 4.122741
```

Solution: Use “Matching” Confidence Intervals

```
> fisher.exact(abdpain,tsmethod="central")
```

```
Central Fisher's Exact Test
```

```
data: abdpain  
p-value = 0.06332  
alternative hypothesis: true odds ratio is not equal to 1  
95 percent confidence interval:  
 0.9235364 14.5759712  
sample estimates:  
odds ratio  
 4.122741
```

Solution: Use “Matching” Confidence Intervals

```
> blaker.exact(abdpain)
```

```
Blaker's Exact Test
```

```
data: abdpain  
p-value = 0.03166  
alternative hypothesis: true odds ratio is not equal to 1  
95 percent confidence interval:  
 1.1734 14.2183  
sample estimates:  
odds ratio  
 4.122741
```

Example 2: One Sample Binomial

```
> library(exactci)
> binom.exact(10,100,p=0.05)
```

```
Exact two-sided binomial test (central method)
```

```
data: 10 and 100
number of successes = 10, number of trials = 100, p-value = 0.05638
alternative hypothesis: true probability of success is not equal to 0.05
95 percent confidence interval:
 0.04900469 0.17622260
sample estimates:
probability of success
      0.1
```

Example 2: One Sample Binomial

```
> binom.exact(10,100,p=0.05,tsmethod="minlike")
```

```
Exact two-sided binomial test (sum of minimum likelihood method)
```

```
data: 10 and 100
```

```
number of successes = 10, number of trials = 100, p-value = 0.03411
```

```
alternative hypothesis: true probability of success is not equal to 0.05
```

```
95 percent confidence interval:
```

```
0.0534 0.1740
```

```
sample estimates:
```

```
probability of success
```

```
0.1
```


Example 2: One Sample Binomial

```
> binom.exact(10,100,p=0.05,tsmethod="blaker")
```

```
Exact two-sided binomial test (Blaker's method)
```

```
data: 10 and 100
```

```
number of successes = 10, number of trials = 100, p-value = 0.03411
```

```
alternative hypothesis: true probability of success is not equal to 0.05
```

```
95 percent confidence interval:
```

```
0.0513 0.1723
```

```
sample estimates:
```

```
probability of success
```

```
0.1
```

Example 3: Two Sample Poisson

```
> poisson.exact(c(10,2),c(20000,17877))
```

```
Exact two-sided Poisson test (central method)
```

```
data: c(10, 2) time base: c(20000, 17877)
```

```
count1 = 10, expected count1 = 6.336, p-value = 0.06056
```

```
alternative hypothesis: true rate ratio is not equal to 1
```

```
95 percent confidence interval:
```

```
0.952422 41.950915
```

```
sample estimates:
```

```
rate ratio
```

```
4.46925
```

Example 3: Two Sample Poisson

```
> poisson.exact(c(10,2),c(20000,17877),tsmethod="minlike")
```

```
Exact two-sided Poisson test (sum of minimum likelihood)
```

```
data: c(10, 2) time base: c(20000, 17877)
```

```
count1 = 10, expected count1 = 6.336, p-value = 0.04213
```

```
alternative hypothesis: true rate ratio is not equal to 1
```

```
95 percent confidence interval:
```

```
1.061630 28.412707
```

```
sample estimates:
```

```
rate ratio
```

```
4.46925
```

Example 3: Two Sample Poisson

```
> poisson.exact(c(10,2),c(20000,17877),tsmethod="blaker")
```

```
Exact two-sided Poisson test (Blaker's method)
```

```
data: c(10, 2) time base: c(20000, 17877)
```

```
count1 = 10, expected count1 = 6.336, p-value = 0.04213
```

```
alternative hypothesis: true rate ratio is not equal to 1
```

```
95 percent confidence interval:
```

```
1.068068 28.412707
```

```
sample estimates:
```

```
rate ratio
```

```
4.46925
```

An Anomaly: Unavoidable Test-CI Inconsistency

Made-Up Example:

	Group A	Group B
Event	7 (2.67 %)	30 (6.07%)
No Event	255 (97.33 %)	464 (93.93%)

- ▶ usual two-sided Fisher's exact test $p = 0.04996$
- ▶ 95% inversion confidence set:

$$\{\beta : \beta \in (0.177, 0.993) \text{ or } \beta \in (1.006, 1.014)\}$$

Matching CI defined as smallest interval that contains all elements of inversion confidence set:

$$(0.177, 1.014)$$

Unavoidable test-CI inconsistency!

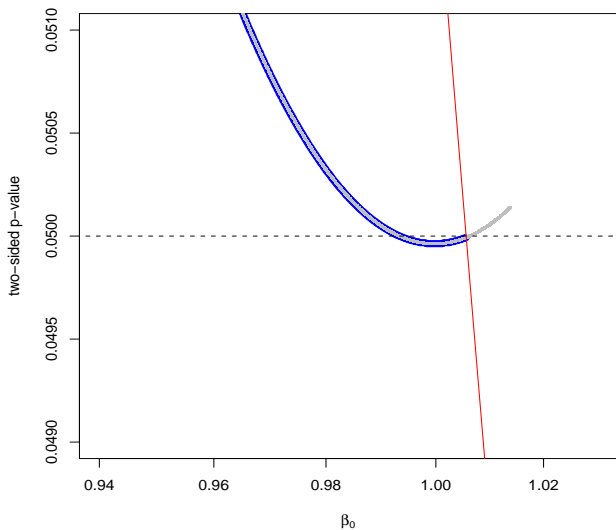


Figure: Made-up example, gray=usual two-sided Fisher's exact, blue=Blaker's exact p-values, red=twice minimum one-sided p-values

References

- ▶ Fay (2010) Biostatistics 373-374
- ▶ Fay (2010) R Journal, 2(1): 53-58.
- ▶ R package: exact2x2
- ▶ R package: exactci