

Fuzzy Statistics

Dr. Mohammed Jasim Mohammed

Estimate p , Binomial Population

Chapter 5

5.1 Fuzzy Estimator of p

We have an experiment in mind in which we are interested in only two possible outcomes labeled "success" and "failure". Let p be the probability of a success so that $q = 1 - p$ will be the probability of a failure. We want to estimate the value of p . let we have a random sample which here is running the experiment n independent times and counting the number of times we had a success.

5.1 Fuzzy Estimator of p

Let x be the number of times we observed a success in n independent repetitions of this experiment. Then our point estimate of p is: $\hat{p} = \frac{x}{n}$

We know that $\frac{(\hat{p} - p)}{\sqrt{p(1-p)/n}}$ is approximately $N(0,1)$ if n is sufficiently large. Then

$$P(z_{\beta/2} \leq \frac{\hat{p} - p}{\sqrt{p(1-p)/n}} \leq z_{\beta/2}) \approx 1 - \beta$$

5.1 Fuzzy Estimator of p

where $z_{\beta/2}$ is defined as:

$$\int_{-\infty}^{z_{\beta/2}} N(0, 1)dx = 1 - \beta/2,$$

Solving the inequality for the p in the numerator we

have $P(\hat{p} - z_{\beta/2}\sqrt{p(1-p)/n} \leq p \leq \hat{p} + z_{\beta/2}\sqrt{p(1-p)/n}) \approx 1 - \beta$.

This leads directly to the $(1-\beta)\%100$ confidence interval for p .

$$[\hat{p} - z_{\beta/2}\sqrt{p(1-p)/n}, \hat{p} + z_{\beta/2}\sqrt{p(1-p)/n}]$$

5.1 Fuzzy Estimator of p

However, we have no value for p to use in this confidence interval. So, still assuming that n is sufficiently large, we substitute \hat{p} for p in the equation using $\hat{q} = 1 - \hat{p}$, and we get the final $(1-\beta)\%100$ approximate confidence interval

$$[\hat{p} - z_{\beta/2} \sqrt{\hat{p}\hat{q}/n}, \hat{p} + z_{\beta/2} \sqrt{\hat{p}\hat{q}/n}]$$

Put these confidence intervals together we get \hat{p} our triangular shaped fuzzy number estimator of p .

Example 5.1.1

Assume that $n = 350$, $x = 180$ so that $p = 0.5143$. The confidence intervals become.

$$[0.5143 - 0.0267z_{\beta/2}, 0.5143 + 0.0267z_{\beta/2}]$$

To obtain a graph of fuzzy p , or \bar{p} , first assume that $0.01 \leq \beta \leq 1$.

Example 5.1.1

```
>> x=linspace(0,1);  
>> y=linspace(0.1,1);  
>> f1=0.5143-0.0267*icdf('Normal',(1-y/2));  
>> f2=0.5143+0.0267*icdf('Normal',(1-y/2));  
>> plot(f1,y,f2,y)  
>> ylabel ('alpha')  
>> xlabel('x')
```


Example 5.1.1

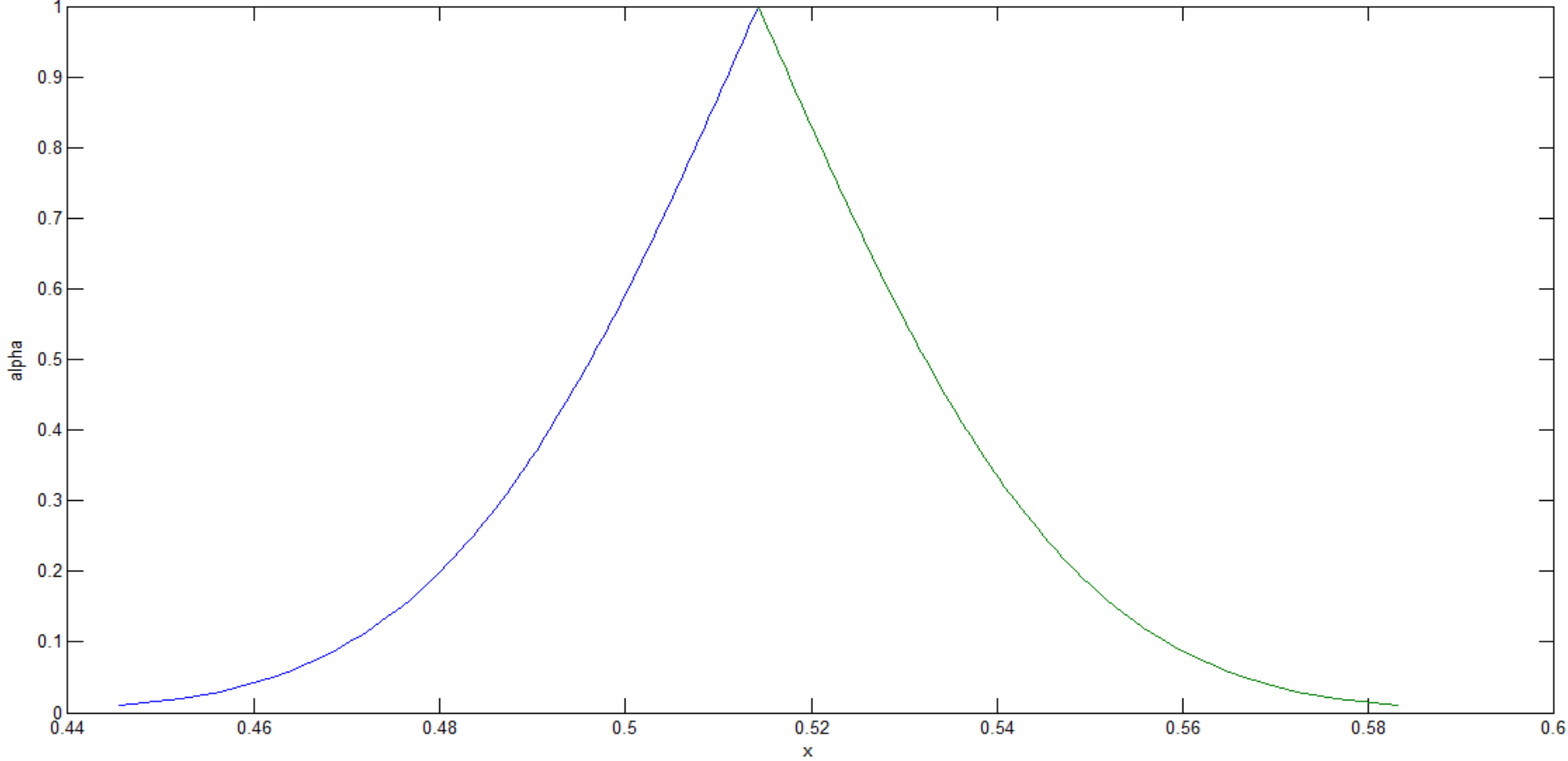


Figure 5.3: Fuzzy Estimator \bar{p} in Example 5.1.1, $0.01 \leq \beta \leq 1$

Example 5.1.1

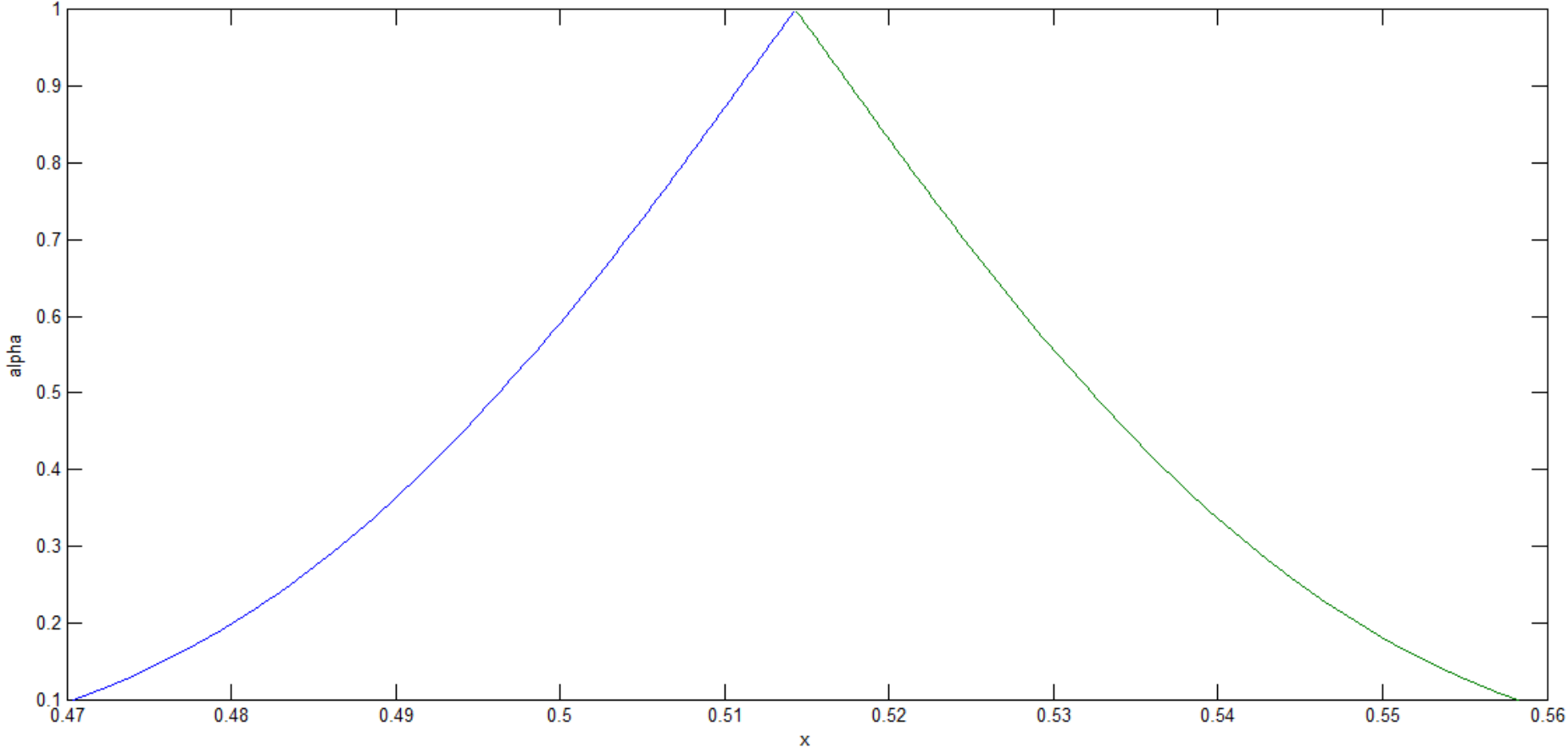


Figure 5.2: Fuzzy Estimator \bar{p} in Example 5.1.1, $0.10 \leq \beta \leq 1$