

Approximating the Binomial Distribution by the Normal Distribution – Error and Accuracy

Peder Hansen

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Department of Mathematics
Uppsala University

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Peder Hansen Uppsala University

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Abstract

Different rules of thumb are used when approximating the binomial distribution by the normal distribution. In this paper an examination is made regarding the size of the approximations errors. The exact probabilities of the binomial distribution is derived and then compared to the approximated value of the normal distribution. In addition a regression model is done. The result is that the different rules indeed gives rise to errors of different sizes. Furthermore, the regression model can be used in order to get guidance of the maximum size of the error.

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Thank you Professor Sven Erick Alm!

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1 Introduction

Neither is any extensive examination found, regarding the rules of thumb used when approximating the binomial distribution by the normal distribution, nor of the accuracy and the error which they result in. The scope of this paper is the most common approximation of a Binomial distributed random variable by the normal distribution. We let $X \sim \text{Bin}(n, p)$, with expectation E(X) = np and variance V(X) = np(1-p), be approximated by Y, where $Y \sim N(np, np(1-p))$. We denote, $X \approx Y$.

The rules of thumb, is a set of different guidelines, minimum values or limits, here denoted L for np(1-p), in order to get a good approximation, that is, $np(1-p) \ge L$. There are various kinds of rules found in the literature and any extensive examination of the error and accuracy has not been found. Reasonable approaches when comparing the errors are, the maximum error and the relative error, which both are investigated.

The main focus lies on two related topics. First, there is a shorter section, where the origin of the rules, where they come from and who is the originator, is discussed. Next comes an empirical part, where the error affected by the different rules of thumb is studied. The result is both plotted and tabled. An analysis of regression is also made, which might be useful as a guideline when estimating the error in situations not covered here. In addition to the main topics, a section dealing with the preliminaries, notation and definitions of probability theory and mathematical statistics is found. Each of the sections will be more explanatory themselves regarding their topics. I presume the reader to be familiar with some basic concepts of mathematical statistics and probability theory, otherwise the theoretical part would range way to far. Therefor, also proofs and theorems are just referred to. Finally there is a summarizing section, where the results of the empirical part are discussed.

2 Theory and methodology

First of all, the reader is assumed to be familiar with basic concepts in mathematical statistics and probability theory. Furthermore there are, as stated above, some theory that instead of being explicitly explained, only is referred to. Regarding the former, I suggest the reader to view for instance [1] or [4] and concerning the latter the reader may want to read [7].

2.1 Characteristics of the distributions

As the approximation of a binomial distributed random variable by a normal distributed random variable is the main subject, a brief theoretical introduction about them is made. We start with a binomial distributed random

variable, X and denote,

$$X \sim \text{Bin}(n, p)$$
, where $n \in \mathbb{N}$ and $p \in [0, 1]$.

The parameters p and n are the probability of an outcome and the number of trials. The expected value and variance of X are,

$$E(X) = np$$
 and $V(X) = np(1-p)$,

respectively. In addition, X has got the probability function

$$p_X(k) = P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$$
, where $0 \le k \le n$,

and the cumulative probability function, or distribution function,

$$F_X(k) = P(X \le k) = \sum_{i=0}^k \binom{n}{i} p^i (1-p)^{k-i}.$$
 (1)

The variable X is approximated by a normal distributed random variable, call it Y, we write,

$$Y \sim N(\mu, \sigma^2)$$
, where $\mu \in \mathbb{R}$ and $\sigma^2 < \infty$.

The parameters μ and σ^2 are the mean value and variance, E(Y) and V(Y), respectively. The density function of Y is

$$f_Y(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-(x-\mu)^2/2\sigma^2}$$

and the distribution function is defined by,

$$F_Y(k) = P(Y \le x) = \int_{-\infty}^x \frac{1}{\sigma\sqrt{2\pi}} e^{-(t-\mu)^2/2\sigma^2} dt.$$
 (2)

2.2 Approximation

Thanks to De Moivre, among others, we know by the central limit theorem that a sum of random variables converges to the normal distribution. A binomial distributed random variable X may be considered as a sum of Bernoulli distributed random variables. That is, let Z be a Bernoulli distributed random variable,

$$Z \sim \text{Be}(p)$$
 where $p \in [0, 1]$,

with probability distribution,

$$p_Z = P(Z = k) = \begin{cases} p & \text{for } k = 1\\ 1 - p & \text{for } k = 0 \end{cases}$$

Consider the sum of n independent identically distributed Z_i 's, i.e.

$$X = \sum_{i=0}^{n} Z_i$$

and note that $X \sim \text{Bin}(n,p)$. For instance one can realize that the probability of the sum being equal to k, $P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$. Hence, we know that when $n \to \infty$, the distribution of X will be normal and for large n approximately normal. How large n should be in order to get a good approximation also depends, to some extent, on p. Because of this, it seems reasonable to define the following approximations. Again, let $X \sim \text{Bin}(n,p)$ and $Y \sim \text{N}(\mu, \sigma^2)$. The most common approximation, $X \approx Y$, is the one where $\mu = np$ and $\sigma^2 = np(1-p)$, this is also the one used here. Regarding the distribution function we get

$$F_X(k) \approx \Phi\left(\frac{k - np}{\sqrt{np(1 - p)}}\right),$$
 (3)

where $F_X(k)$ is defined in (1) and Φ is the standard normal distribution function. We extend the expression above and get that,

$$F_X(b) - F_X(a) = P(a < X \le b) \approx \Phi\left(\frac{b - np}{\sqrt{np(1 - p)}}\right) - \Phi\left(\frac{a - np}{\sqrt{np(1 - p)}}\right). \tag{4}$$

2.3 Continuity correction

We proceed with the use of continuity correction, which is recommended by [1], suggested by [4] and advised by [9], in order to decrease the error, the approximation (3) will then be replaced by

$$F_X(k) \approx \Phi\left(\frac{k + 0.5 - np}{\sqrt{np(1-p)}}\right)$$
 (5)

and hence (4) is written as

$$F_X(b) - F_X(a) = P(a < X \le b) \approx \Phi\left(\frac{b + 0.5 - np}{\sqrt{np(1-p)}}\right) - \Phi\left(\frac{a + 0.5 - np}{\sqrt{np(1-p)}}\right).$$
(6)

This gives, for a single probability, with the use of continuity correction, the approximation,

$$p_X(k) = F_X(k) - F_X(k-1) \approx \Phi\left(\frac{k+0.5-np}{\sqrt{np(1-p)}}\right) - \Phi\left(\frac{(k-1)+0.5-np}{\sqrt{np(1-p)}}\right)$$
(7)

and further we note that it can be written

$$F_X(k) - F_X(k-1) \approx \int_{k-0.5}^{k+0.5} f_Y(t)dt.$$
 (8)

2.4 Error

There are two common ways of measuring an error, the absolute error and the relative error. In addition another usual measure of how close, so to speak, two distributions are to each other, is the *supremum norm*

$$\sup_{A} |P(X \in A) - P(Y \in A)|.$$

However, from a practical point of view, we will study the absolute error and relative error of the distribution function. Let a denote the exact value and \bar{a} the approximated value. The absolute error is the difference between them, the real value and the one approximated. The following notation is used,

$$\varepsilon_{abs} = |a - \bar{a}|$$
.

Therefor, the absolute error of the distribution function, denoted $\varepsilon_{F_{abs}}(k)$, for any fixed p and n, where $k \in \mathbb{N} : 0 \le k \le n$, without use of continuity correction, is

$$\varepsilon_{F_{abs}}(k) = \left| F_X(k) - \Phi\left(\frac{k - np}{\sqrt{np(1 - p)}}\right) \right|.$$
(9)

Regarding the relative error, in the same way as before, let a be the exact value and \bar{a} the approximated value. Then the relative error is defined as

$$\varepsilon_{rel} = \left| \frac{a - \bar{a}}{a} \right|.$$

This gives the relative error of the distribution function, denoted $\varepsilon_{F_{rel}}(k)$, for any fixed p and n, where $k \in \mathbb{N} : 0 \le k \le n$, without use of continuity correction, is

$$\varepsilon_{F_{rel}}(k) = \frac{\varepsilon_{F_{abs}}(k)}{F_X(k)},$$

or equivalently, inserting $\varepsilon_{F_{abs}}(k)$ from (9),

$$\varepsilon_{F_{rel}}(k) = \frac{\left| F_X(k) - \Phi\left(\frac{k - np}{\sqrt{np(1 - p)}}\right) \right|}{F_X(k)}.$$

2.5 Method

The examination is done in the statistical software R. The software provides predefined functions for deriving the distribution function and probability function of the normal and binomial distributions. The examination is split into two parts, where the first part deals with the absolute error of the approximation of the distribution function and the second part concerns the relative error. The conditions under which the calculations are made, are those found as guidelines in [4]. The calculations will be made with the help of a two-step algorithm. At the end of each section a linear model is fitted to the error. Finally, an overview, where a table and a plot of how the value of npq, where q = 1 - p, affects the maximum approximation error for different probabilities are presented.

2.5.1 Algorithm

The two step algorithm below is used. The values of npq mentioned in the literature are, in all cases said to be equal or larger than some limit, here denoted L. The worst case scenario, as to speak, is the case where they are equal, that is, npq = L. Therefor equalities are chosen as limits. We know that $n \in \mathbb{N}$, which means that p must be semi-fixed if the equality should hold, this means that the values of p are adjusted, but still remain close to the ones initially chosen. The way of doing this is a two-step algorithm. First a reasonable set of different initial probabilities, \tilde{p}_i 's are chosen, whereafter the corresponding \tilde{n}_i values, which in turn will be rounded to n_i , are derived. These are used to adjust \tilde{p}_i to p_i so that the equality will hold.

- 1. (a) Chose a set $\tilde{\mathbf{P}}$ of different initial probabilities, $\tilde{p}_i \in [0, 0.5]$, where $i \in \mathbb{N} : 0 < i < |\tilde{\mathbf{P}}|$.
 - (b) Derive the corresponding $\tilde{n}_i \in \mathbb{R}^+$ so that $\tilde{n}_i \tilde{p}_i (1 \tilde{p}_i) = L$,
 - (c) continue by deriving $n_i \in \mathbb{N}$, in order to get a integer,

$$n_i(p_i) := \min\{n \in \mathbb{N} : n\tilde{p}_i(1 - \tilde{p}_i) \ge L\}. \tag{10}$$

Now we got a set of $n_i \in \mathbb{N}$, denote it **N**.

2. Chose a set **P** so that for every $p_i \in \mathbf{P}$,

$$n_i p_i (1 - p_i) = L.$$

The result is that we always keep the limit L fixed. Let us take a look at an example. Let L = 10, use continuity correction and the initial $\tilde{\mathbf{P}} = 0.1(0.1)0.5$,

Exe	emplifyin	g table	of algori	ithm val	lues
\overline{i}	1	2	3	4	5
$ ilde{p}_i$	0.1	0.2	0.3	0.4	0.5
\tilde{n}_i	111.11	62.50	47.62	41.67	40.00
n_i	112	63	48	42	40
p_{i}	0.099	0.198	0.296	0.391	0.500

Different rules of thumb are suggested by [4]. Using approximation (3) the authors say that $np(1-p) \geq 10$ gives reasonable approximations and in addition, using (5), it may even be sufficient using $np(1-p) \geq 3$. The investigation takes place under three different conditions,

- np(1-p) = 10 without continuity correction, suggested in [4],
- np(1-p) = 10 with continuity correction, suggested in [2],
- np(1-p) = 3 with continuity correction, suggested in [4].

The investigation of the rules is made only for $p_i \in [0, 0.5]$ due to symmetry. As we see, np(1-p) simply gets the same values for $p \in [0,0.5]$ as for $p \in [0.5,1]$. So, for every p, $n_i(p_i)$ is derived, this in turn, means that we get $n_i(p_i) + 1$ approximations. For every $n_i(p_i)$, and of course p_i as well, we define the maximum absolute error of the approximation of the distribution function,

$$M_{F_{abs}} = \max\{\varepsilon_{F_{abs}}(k) : 0 \le k \le n_i(p_i)\},\tag{11}$$

and in addition the maximum relative error

$$M_{F_{rel}} = \max\{\varepsilon_{F_{rel}}(k) : 0 \le k \le n_i(p_i)\}. \tag{12}$$

The results are both tabled and plotted.

2.5.2 Regression

Beforehand, some plots where made which indicated that the maximum absolute error could be a linear function of p. Regarding the relative maximum error, a quadratic or cubic function of p seemed plausible. Because of that, a regression is made. The model assumed to explain the absolute error is

$$M_{\varepsilon} = \alpha + \beta p + \epsilon_l, \tag{13}$$

where M_{ε} is the maximum error, α is the intercept, β the slope and ϵ_l the error of the linear model. For the relative error, the two additional regression models are,

$$M_{\varepsilon} = \alpha + \beta p + \gamma p^2 + \epsilon_l \tag{14}$$

and

$$M_{\varepsilon} = \alpha + \beta p + \gamma p^2 + \delta p^3 + \epsilon_l. \tag{15}$$

3 Background

In the first basic courses in mathematical statistics, the approximations (3) and (5) are taught. Students have learned some kind of rules of thumb they should use when applying the approximations, myself included, for example the rules suggested by Blom [4],

$$np(1-p) \ge 10$$
, $np(1-p) \ge 3$ with continuity correction.

Any motivation why the limit L is set to be L = 10 and L = 3 respectively is not found in the book. On the other hand, in 1989 Blom claims that the approximation "gives decent accuracy if npq is approximately larger than 10" with continuity correction [2]. Further, it is interesting, that Blom changes the suggestion between the first edition of [3] from 1970, where it says, similarly as above, that it "gives decent accuracy if np(1-p) is approximately larger than 10" with continuity correction, and in the second edition from 1984 the same should yield, but now instead without use of continuity correction, the conclusion is that there has been some fuzziness regarding the rules. Neither have I, nor my advisor Sven Erick Alm, found any examination of the accuracy of these rules anywhere else. With Blom [4] as starting-point, I begun backtracking, hoping that I could find the source of the rules of thumb. It is worth mentioning that among authors, slightly different rules have been used. For instance Alm himself and Britton, present a schema with rules for approximating distributions, in which np(1-p) > 5with continuity correction is suggested [1]. Even between countries, or from an international point of view, so to speak, differences are found. Schader and Schmid [10] says that "by far the most popular are"

$$np(1-p) > 9$$

and

$$np > 5 \text{ for } 0
 $n(1-p) > 5 \text{ for } 0.5$$$

which I am not familiar with and I have not found in any Swedish literature. In the mid-twentieth century, more precise 1952, Hald [9] wrote,

An exhaustive examination on the accuracy of the approximation formulas has not yet been made, and we can therefore only give rough rules for the applicability of the formulas.

With these words in mind, the conclusion is that there probably does not exist any earlier work made about the accuracy of the approximation. However, Hald himself made an examination in the same work for npq > 9. Further he also points out that in cases where the binomial distribution is very skew, $p < \frac{1}{n+1}$ and $p > \frac{n}{n+1}$, the approximation cannot be applied. Some articles have been found that briefly discuss the accuracy and error of the distributions. Mainly, the focus of the articles lies on some more advanced method of approximating than (3) or (5). An update of [2] has been made by Enger, Englund, Grandell and Holst in 2005, [4]. The writers have been contacted and Enger was said to be the one that assigned the rules. Hearing this made me believe that the source could be found. However, Enger could not recall from where he had got it [6]. That is how far I could get. Nevertheless, the examination remains as interesting as beforehand.

Discussing rules for approximating, one can not avoid at least mentioning the Berry-Esseen theorem. The theorem gives a conservative estimate, in the sense that it gives the largest possible size of the error. It is based upon the rate of convergence of the approximation to the normal distribution. The Berry-Esseen theorem will not be further examined here, but there are several interesting articles due to that the theorem is improved every now and then, most recently in May 2010 [11].

4 The approximation error of the distribution function

The errors of the approximations, $M_{F_{abs}}$ and $M_{F_{rel}}$, defined in (11), and (12) respectively, are plotted and tabled. The cases that are examined are those mentioned earlier, suggested by [4].

4.1 Absolute error

We examine the absolute maximum errors of the approximation of the distribution function, $M_{F_{abs}}$ defined in (11), here in the first part. In addition to

that a regression is made, defined in (13), to see if we might find any linear trend.

Case 1: npq = 10, without continuity correction

First, the case where L=10=npq, without continuity correction. $\tilde{\mathbf{P}}$, the set of different initial probabilities is chosen to be $\tilde{p}_i=0.01(0.01)0.50$. This means that we use 50 equidistant \tilde{p}_i . The smallest probability is $p_1=0.0100$ and it has the largest error $M_{F_{abs}}=0.0831$. $M_{F_{abs}}$ decreases the closer to 0.5 we get, which is natural since the binomial distribution tends to be skew. The points make a pattern which is a bit curvy, but still the points are close to the straight line in Figure 1. Another remark made, is that the distance between the probabilities decreases the closer to 0.5 we get. The fact that there are several \tilde{n}_i rounded to the same value of n_i , which in turn gives equal values on p_i , makes several $M_{F_{abs}}$ the same, and plotted in the same spot. So they are all there, but not visible due to that reason. Next we try to fit a linear model for $M_{F_{abs}}$. The result is

$$M_{F_{abs}} = 0.0836 - 0.0417p + \epsilon_l.$$

The regression line is the straight line in Figure 1. The slope of the line shows that the size of $M_{F_{abs}}$ changes moderately. Note that the sum of the errors of the regression line, $\sum |\epsilon_l|$, is relatively small, the result should be somewhat precise estimates of $M_{F_{abs}}$ for probabilities which are not taken in consideration here.

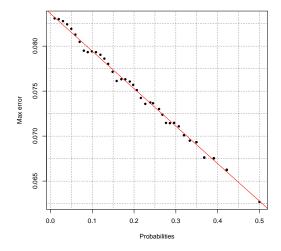


Figure 1: Maximum absolute error for npq = 10 without continuity correction. The straight line is the regression line, $M_{F_{abs}} = 0.0836 - 0.0417p$.

Case 2: npq = 10, with continuity correction

Under these circumstances $M_{F_{abs}}$ decreases and is about four times smaller than without continuity correction. The regression line,

$$M_{F_{abs}} = 0.0209 - 0.0416p + \epsilon_l, \tag{16}$$

also has got a four times smaller intercept than in the first case. What is interesting is that, the slope is approximately the same in both cases, this in turn, means that for every $\tilde{p}_i = 0.01(0.01)0.50$, it holds that $M_{F_{abs}}$ also is four times smaller. This can be seen in Figure 2.

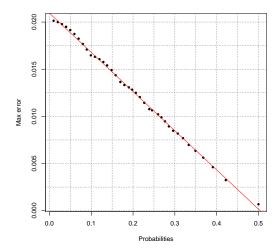


Figure 2: Maximum absolute error for npq = 10 with continuity correction. The straight line is the regression line, $M_{F_{abs}} = 0.0209 - 0.0416p + \epsilon_l$.

Case 3: npq = 3, with continuity correction

Finally we take a look at the last case, regarding the absolute error, where L=3=npq and continuity correction is used. The plot is seen in Figure 3. $\tilde{\mathbf{P}}$ is the same as above. In this case the regression line is

$$M_{F_{abs}} = 0.0373 - 0.0720p + \epsilon_l.$$

The largest error, $M_{F_{abs}} = 0.0355$ appears at $p_1 = 0.0100$ and is about twice the size compared to the largest $M_{F_{abs}}$ for L = 10. The slope of the line is more aggressive here, which in turn results in errors, one order of magnitude less than in the Case 1 for probabilities close to 0.5. Also here the sum of

discrepancy from the regression line is relatively small which should result in fairly good estimations of $M_{F_{abs}}$.

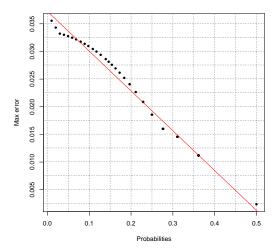


Figure 3: Maximum absolute error for npq=3, with continuity correction. The straight line is the regression line, $M_{F_{abs}}=0.0373-0.0720p$.

4.2 Relative Error

Here, the maximum relative error of the approximation of the distribution function, $M_{F_{rel}}$, defined in (12) is examined. The regression models (14) and (15) are both tested.

Case 1: npq = 10, without continuity correction

In the first case we perform the calculations under, L=10=npq without continuity correction. The result is shown in Figure 4. As we see $M_{F_{rel}}$ increases very rapidly. The smallest value of $M_{F_{rel}}$, 16.97317 is at p_1 . The largest 138.61756 at p_{50} . As we see in Table 4, it is k=0 that gives the largest error. For other values of k the error is much smaller. Furthermore we note that $M_{F_{rel}}$ is very large. If we look at a specific example where p=0.2269, which means that n=57, then $X\sim \text{Bin}(57,0.2269)$. Let X be approximated, according to (3), by $Y\sim N(12.933,3.162078)$. We get that $P(X\leq 1)=7.55\cdot 10^{-6}$ and $P(Y\leq 1)=8.04\cdot 10^{-5}$. Under these circumstances we get,

$$M_{F_{rel}} = \frac{|P(X \le 1) - P(Y \le 1)|}{P(X \le 1)} = 9.64.$$

The result is shown in Table 4. So the relative error is, as we also can see, large, for small k and small probabilities. The regression curves, defined in (14) and (15) are,

$$M_{F_{rel}} = 14.66 + 69.86p + 416.14p^2 + \epsilon_l$$

and

$$M_{F_{red}} = 21.53 - 92.26p + 1246.60p^2 - 1136.07p^3 + \epsilon_l$$

respectively. We note that there are not any larger differences in accuracy depending on the choice of model. Naturally, the discrepancy of the second model is lower.

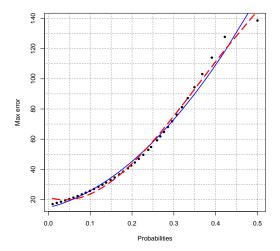


Figure 4: Maximum relative error for npq=10 without continuity correction. The solid line is the regression curve, $M_{F_{rel}}=14.66+69.86p+416.14p^2$ and the dashed line, $M_{F_{rel}}=21.53-92.26p+1246.60p^2-1136.07p^3$.

Case 2: npq = 10, with continuity correction

We continue by looking at the same case as above, but here continuity correction is used. This gives somewhat remarkable results, $M_{F_{rel}}$ is actually about two times larger than without continuity correction. Let us study the same numeric example as above, except that we use continuity correction. We got p = 0.2269 which again means that n = 57, then $X \sim \text{Bin}(57, 0.2269)$. We let X be approximated, according to (5), by $Y \sim N(12.933, 3.162078)$. It results in, $P(X \le 1) = 7.55 \cdot 10^{-6}$ and $P(Y \le 1 + 0.5) = 0.000150$. Under

these circumstances we get,

$$M_{F_{rel}} = \frac{|P(X \le 1) - P(Y \le 1 + 0.5)|}{P(X \le 1)} = 18.84,$$

which fits the values in Table 5. $M_{F_{abs}}$ gets dramatically worse when we use continuity correction than without. Hence, also $M_{F_{rel}}$ becomes worse. In Figure 5 one can judge that the results gets worse as we get closer to probabilities near 0.5. The regression curves, defined in (14) and (15) are,

$$M_{F_{rel}} = 34.9 - 69.8p + 1597.1p^2 + \epsilon_l$$

and

$$M_{F_{rel}} = 37.4 - 127.3p + 1891.8p^2 - 403.2p^3 + \epsilon_l,$$

respectively. Looking at Figure 5, we see that the difference between the two models is insignificant.

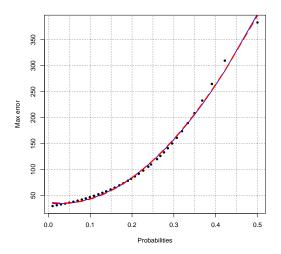


Figure 5: Maximum relative error for npq=10 with continuity correction. The solid line is the regression curve, $M_{F_{rel}}=34.9-69.8p+1597.1p^2$ and the dashed line, $M_{F_{rel}}=37.4-127.3p+1891.8p^2-403.2p^3$.

Case 3: npq = 3 with continuity correction

Here, in the last case npq = 3 and continuity correction is used, see Figure 6. This gives the curves of regression, defined in (14) and (15),

$$M_{F_{rel}} = 0.473 + 2.204p + 2.123p^2 + \epsilon_l$$

and

$$M_{F_{rol}} = 0.514 + 1.155p + 7.858p^2 - 7.885p^3 + \epsilon_l,$$

respectively. As we see $M_{F_{rel}}$ actually get the smallest value here, where npq=3 and continuity correction is used. As well as in the two other cases regarding the relative error the difference between the quadratic and cubic regression model is minimal.

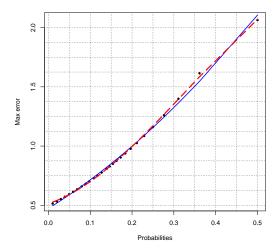


Figure 6: Maximum relative error for npq=3 with continuity correction. The solid line is the regression curve, $M_{F_{rel}}=0.473+2.204p+2.123p^2$ and the dashed line, $M_{F_{rel}}=0.514+1.155p+7.858p^2-7.885p^3$.

5 Summary and conclusions

The three different rules of thumbs that are focused on turned out to give approximation errors of different sizes. Regarding the absolute errors, the largest difference is found between the case where L=10 without continuity correction and L=10 with continuity correction. The largest error decreases from ~ 0.08 to about ~ 0.02 , which is approximately four times smaller, a relatively large difference. Letting L=3 and using continuity correction we end up with the largest error ~ 0.035 , closer to the latter case, but still between them. When using this common and simple way of approximating, depending on the problem, different levels of tolerance usually are accepted. A common level in many cases may be 0.01. If we look deeper, we see that the probabilities for getting such a small $M_{F_{abs}}$ differs from between the rules of thumb. Using npq=10 without continuity correction does not even reach to the 0.01 level of accepted accuracy. Comparing this to the other

two cases which in contrast reach the 0.01 level for probabilities ~ 0.25 in the same case as above but in addition with continuity correction, and for probabilities ~ 0.35 in the case where npq=3. Further, it would be interesting to investigate how the relationship between k and n affects the error. In addition, another interesting part would be some tables indicating how large n should be in order to get sufficiently small errors, for different probabilities.

Concerning the relative errors I would say that the applicability may be somewhat uncertain, due to the fact that $M_{F_{rel}}$ is very large for small values of k but rapidly decrease. This fact, I may say, make the plots look a bit extreme and there are other values of k that give much better approximations. Judging by Tables 4, 5 and 6 indeed this seems to be the case. We know that the approximation is motivated by the central limit theorem, however, what we also know, is that it does not hold the same accuracy for small probabilities, that is, the tails of the distributions. This is also the direct reason why the accuracy gets worse when using continuity correction, it puts extra mass on the already too large approximated value. In a similar way we get the explanation why the relative error increases when the value of npg changes from 10 to 3, (as one maybe would expect the opposite), the mean value of the normal distribution, np, gets closer to 0 which in turn gives additional mass. The conclusion is, one should remember that due to the fluctuations depending on k, of the relative errors, what we also can see in Tables 4, 5 and 6, that the regression model also provides conservative estimates of the errors. As a natural alternative, and most likely better, Poisson approximation is recommended for small probabilities. Like in the previous case concerning the absolute errors, some more exhaustive examination of the relative error would be interesting. How large should n be to get acceptable levels of the error, for instance 10% or 5% and so on.

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Tables

Regarding the plotted probabilities, that is the set \mathbf{P} , only the maximum error is plotted. One can not tell from which k the error comes from, neither can one tell if the error is of similar size for other values of k. To get a more detailed picture this section contains tables both for the absolute errors and the relative errors. It would have been possible to table all the errors for all values of k, but due to the fact that the cardinality of \mathbf{N} at times, that is for small probabilities, is relatively large, it would have taken too much place. This made me table only the 10 values of k which resulted in the largest errors. The columns in the tables, that contains the values of k is in descending order. What this means is that the first value of k in each column is the maximum error that is plotted. On the side of every column of k, there is a column where the corresponding error is written. These two sub columns, got a common header which tells the value of p in the specific case.

0.1899	$\varepsilon_{F_{abs}}$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	0.5	$\varepsilon_{F_{abs}}$	0.0627 0.0614 0.058 0.0544 0.0487 0.04373 0.0373 0.0371 0.026
	$\varepsilon_{Fab_S} \mid k \ \varepsilon_{Fab_S} \mid $	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		$\varepsilon_{Fabs} \mid k \ \varepsilon_{Fabs} \mid k \ \varepsilon_{F$	
0.1792	k ε_{F_a}	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	0.4219	k ε_{F_a}	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
969	F_{abs}	0.0761 12 0.0763 12 0.0761 10 0.0747 11 0.0665 13 0.0683 13 0.0662 10 0.0636 10 0.051 14 0.0536 14 0.044 9 0.0368 15 0.0342 15 0.0368 15 0.0308 8 0.0281 8 0.0138 16 0.0219 16 0.0138 16 0.0219 16	606	Fabs	0.0676 17 0.0647 17 0.0675 17 0.0645 16 0.0603 15 0.0652 18 0.048 14 0.0522 15 0.047 19 0.0424 20 0.0329 0.0293 21 0.032 12 0.0293 21 0.022 12 0.0293 12
0.1696	k 6.	12 0. 11 0 0. 10 0. 14 0. 9 0. 15 0. 7 0.	0.3909	<i>3</i>	16 0. 17 0. 18 0. 14 0. 13 0. 20 0. 21 0.
0.1584	$^{\varepsilon}F_{abs}$	0.0761 0.0655 0.0665 0.0652 0.0494 0.0342 0.0308 0.0198	0.3679	$\varepsilon_{F_{abs}}$	0.0676 0.0675 0.0603 0.0601 0.0475 0.0329 0.0329 0.022
	$b_s \mid k$	771 11 884 13 84 10 21 14 88 9 32 15 83 15 83 15 84 16 69 7		bs k	993 15 56 16 52 14 42 17 18 13 96 19 96 19 17 11
0.1487	k ε_{F_a}	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	0.3492	k ε_{F_a}	0.0715 14 0.0774 14 0.0715 14 0.0771 14 0.07071 15 0.0693 15 0.0693 15 0.0711 13 0.0702 13 0.685 15 0.0657 15 0.068 16 0.0653 15 0.0663 15 0.0645 15 0.0665 15 0.0653 15 0.0653 15 0.0653 15 0.0653 15 0.0653 15 0.0653 15 0.0653 17 0.0652 17 0.0652 17 0.0652 17 0.0652 17 0.0652 17 0.0652 17 0.0653 17 0.0652 17 0.0652 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0654 17 0.0655 11 0.0655 11 0.0655 12 0.0655 12 0.0655 12 0.0655 12 0.0655 12 0.0655 12 0.0655 13 0.0655
81	Fabs	078 1 0.075 1 0.0706 1 0.0531 1 0.0549 1 0.0359 1 0.0187 1	0.3333	aps	0695 1 0682 1 0629 1 0593 1 0508 1 0453 1 0366 1 0301 1 0235 1
0.1381	k E1	11 0 12 0 10 0.0 13 0.0 9 0.0 14 0.0 8 0.0 7 0.0		k ε1	15 0.0 14 0.0 13 0.0 17 0.0 18 0.0 19 0.0
0.129	ε_{Fabs}	0.0786 0.0724 0.0724 0.0614 0.0573 0.0382 0.0382 0.0277 0.0204	0.3194	ε_{Fabs}	0.0701 0.0688 0.0632 0.0502 0.0466 0.0347 0.0322 0.0205
	k	79 11 11 12 34 10 37 13 36 9 58 15 58 15 58 15 58 15 58 15 58 15 58 15		- k	11 14 75 15 52 13 57 16 13 12 28 17 88 10 10 10
0.1196	$\varepsilon_{F_a b}$	0.07 0.074 0.055 0.055 0.056 0.046 0.046 0.026	0.307	$\varepsilon_{F_{ab}}$	0.0228
	3 k	793 10 758 10 723 2 623 9 623 14 8433 14 8398 8 8244 15 7 7 7		1 k	715 14 685 15 685 15 578 16 541 12 425 17 269 18 196 19
0.109	k $\varepsilon_{F_{q}}$	111 0.0 100 0.0 112 0.0 100 0.0 113 0.0 144 0.0 17 0.0 16 0.0 17 0.0 16 0.0	0.2959	k EF	14 0.0 0 113 0.0 0 115 0.0 0 116 0.0 0 111 0.0 0 111 0.0 0 111 0.0 0 118 0.0
0.0991	$ F_{abs} $.0794 .0772 .0711 .0647 .0555 .0458 .0377 .0264	0.2857	Fabs	.0714 .0702 .0643 .0607 .0514 .0459 .0364 .0229
0.0	, k	8 12 0 8 12 0 9 9 9 0 6 13 0 14 0 17 0 18 0 18 0 19 0 10	0.2	- ×	2 14 0 3 15 0 6 12 0 6 12 0 8 11 0 7 17 0 6 10 0 6 18 0
0.0893	$^{\varepsilon}F_{ab}$	0.078 0.069 0.069 0.053 0.048 0.028 0.028 0.020	0.2764	$\varepsilon_{F_ab_b}$	0.071 0.063 0.063 0.048 0.048 0.033 0.033 0.020
	$_{lbs} \mid k$	795 791 791 791 791 793 793 793 793 793 793 793 793		. k	724 13 705 144 705 144 705 145 508 15 508 15 516 11 164 16 852 17 872 18 873 18
0.0799	k ε_{F_G}	11 0.0795 11 11 0.0791 10 0.0791 10 0.0691 12 0.0685 13 0.0516 13 8 0.0507 8 14 0.0337 14 0.0335 7 0.0305 7 0.0305 6 0.0138 6	0.2678	k EF	13 0.0724 13 0.0715 10 0.0726 14 0.0711 12 0.0654 12 0.0636 15 0.0608 15 0.0626 11 0.0516 11 0.0488 10 0.0352 17 0.0337 17 0.0314 10 0.0326 9 0.0202 18 0.0206 18 0.0187 9 0.0182
8698	F_{abs}	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	2598	F_{abs}	0.073 13 0.0724 13 0.0715 0.0698 14 0.0705 14 0.0711 0.0672 12 0.0654 12 0.0638 0.0554 11 0.0516 11 0.0489 0.0421 11 0.0516 11 0.0488 0.0376 10 0.0352 17 0.0337 0.0292 17 0.0314 10 0.0326 0.0292 17 0.0314 10 0.0326 0.017 18 0.0187
0.0	k		0.2	~	
0.0597	$\varepsilon_{F_{abs}}$	0.0828 10 0.0824 10 0.0819 10 0.0818 0.0778 11 0.0776 11 0.0778 10 0.0782 0.0729 9 0.0765 9 0.0748 9 0.0754 0.0628 12 0.0638 12 0.0654 0.0419 13 0.0435 18 0.0457 0.0419 13 0.0435 7 0.0378 7 0.0375 0.0244 7 0.0396 7 0.0373 7 0.0375 14 0.0297 0.0243 14 0.026 14 0.0277 14 0.0297 0.0158 6 0.0168 0.0217 6 0.0184 6 0.0168 0.0168 0.0168	0.2454	ε_{Fabs}	0.0734 12 0.0736 13 0.0737 13 0.0737 13 0.0737 13 0.0731 12 0.0731 12 0.0731 12 0.0771 14 0.0664 14 0.0663 14 0.0672 14 0.0672 14 0.0681 15 0.0634 11 0.0654 11 0.0654 15 0.0541 15 0.0541 15 0.0547 11 0.0484 10 0.0476 10 0.0444 10 0.042 16 0.0335 16 0.0353 16 0.0353 16 0.0354 17 0.0256 17 0.0191 17 0.0213 17 0.024 17 0.0256 17 0.0179 8 0.0159 8 0.0138 18 0.0142 18 18
	bs k	319 76 11 748 938 128 81 85 13 7 7 7 7 84 6 84 6 15 15 16 17 17 18 18 18 18 19 10 10 10 10 11 12		bs k	7.37 13 7.12 12 7.12 12 7.12 14 802 11 7.44 10 884 16 873 9 924 17 38 18
0.0499	k ε_{F_C}	10 0.00 11 0.00 12 0.00 8 0.00 13 0.00 7 0.00 14 0.00 15 0.00	0.2389	k EF	13 0.07 12 0.07 11 0.06 15 0.06 16 0.07 8 0.00
0.0399	F_{abs}	.0824 .0768 .0765 .0621 .0605 .0435 .0396 0.026	0.2269	F_{abs}	0.0736 0.073 0.0634 0.0511 0.0476 0.0353 0.0353 0.0139
0.0	_ k	1100 11100 11200 11200 11300 11400 11500	0.2	- ×	13 0 12 1 12 0 11 0 11 0 10 0 10 0 10 0 10
0.03	k ε_{Fabs} $\mid k$ ε_{Fabs} $\mid k$ ε_{Fabs} $\mid k$ ε_{Fabs} $\mid k$ ε_{Fabs}	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	0.2163	$k arepsilon_{Fabs} ig k arepsilon_{Fabs} ig k arepsilon_{Fabs} ig k arepsilon_{Fabs} ig k arepsilon_{Fabs} ig $	
	$_{bs}\mid k$	083 10 795 9 749 11 887 12 887 7 142 7 135 14 135 14 03 15		.bs k	0.0751 12 0.0652 13 0.0681 11 0.0615 14 0.0536 10 0.036 9 0.036 9 0.0301 16 0.0199 17
0.02	k ε_{F_a}	0.0831 10 0.083 10 0.0808 10 0.00808 10 0.0708 9 0.0775 9 0.0752 8 0.0659 12 0.0587 12 0.0587 13 0.0377 13 0.0387 13 0.0351 14 0.0253 14 0.0251 14 0.0252 15 0.0081	0.2066	$k \ \varepsilon_{F_a}$	12 0.0751 12 11 0.0551 13 0.0725 13 11 0.0651 11 11 0.0651 11 11 0.0536 10 0.0536 10 0.0536 10 0.036 15 0.036 15 0.036 15 0.036 15 0.036 17 0.0172 8 0.0172 8
0.01	F_{abs}	0.0831 0.0808 0.0738 1 0.0672 0.0569 1 0.0377 1 0.0253 0.0092 1	0.1979	F_{abs}	0.0757 12 0.0717 13 0.0597 14 0.0597 14 0.0581 15 0.0384 9 0.0281 16 0.0217 8
p = 0.01	k e	10 0. 11 0. 8 0. 12 0. 7 0. 13 0. 14 0.	0.18	k e	12 0. 13 0. 11 14 0. 10 0. 15 0. 16 0. 17 0.

Table 1: Table of the 10 largest errors, $\varepsilon_{F_{abs}}$ and which k is comes from, for every p_i , under npq = 10 without continuity correction.

668	Fabs	0.0131 0.0121 0.0104 0.0077 0.0068 0.0055 0.0055 0.0047	امر	F_{abs} $7e-04$ $7e-04$ $6e-04$ $6e-04$ $6e-04$ $6e-04$ $6e-04$ $6e-04$	4e-04 4e-04
0.1899	$k \in$	13 0.1 14 0.1 11 0.1 7 C 8 0.1 6 0.1 15 0.1 19 0.1	0.5	k E ₁ 27 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7	
92	Tabs	0.0136 13 0.0133 13 0.0134 12 0.0134 12 0.0013 14 0.0009 14 0.0074 17 0.0066 8 0.0066 8 0.0066 6 0.0066 6 0.0067 18 0.0057 18 0.0057 19 0.0047 15 0.0048 19 0.0047 10 0.0047 1	19	$\begin{bmatrix} \tau_{abs} \\ \tau_{abs} \\ 0032 \\ 0024 \\ 0022 \\ 0019 \\ 0019 \\ 0013 \\$	1100.
0.1792	$k \in E_I$	13 0.0 14 11 0.0 7 0.0 8 0.0 6 0 6 0 18 0 15 0.0	0.4219	$\begin{array}{c cccc} \varepsilon_{Fabs} & k & \varepsilon_{Fabs} \\ \hline 0.0046 & 18 & 0.003 \\ 0.0048 & 17 & 0.0032 \\ 0.0038 & 19 & 0.0024 \\ 0.0028 & 16 & 0.0022 \\ 0.0027 & 12 & 0.0029 \\ 0.0027 & 11 & 0.0019 \\ 0.0023 & 11 & 0.0019 \\ 0.0013 & 11 & 0.0019 \\ 0.0019 & 10 & 0.0013 \\ \end{array}$	20 0.0 14 0
96	abs	1136 1134 1099 1097 1076 1064 1052 1047	60	abs	015
0.1696	k ε_F	2 5 1 4 7 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0.3909	8 6 7 8 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9	9 0.C
4.	abs	144 1135 1111 1 1111 1 1 1 1 1	6.	$\begin{array}{c c} abs & \\ \hline & abs & \\ \hline & 0.056 & \\ \hline & 0.025 & \\ \hline & 0.032 & \\ \hline & 0.032 & \\ \hline & 0.024 & \\ \hline & 0.023 & \\ \hline$	0.22
0.1584	k s _F	2 0.0 2 0.0 0.0 1 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0	0.3679	\$\epsilon \$\epsilon \text{\$\epsilon \tex	0.0
7	l squ	149 1 134 1 121 1 121 1 121 1 134 1 134 1 134 1 134 1 135 1 1 1 1 1 1 1 1 1		$\begin{array}{c c c c c c c c c c c c c c c c c c c $	002 2
0.1487	EF.	2 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.	0.3492	\$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$	0.0
	1 89.	254 12 33 13 11 13 83 14 1008 14 87 18 171 8 171 8 165 165 165		Page	24 2.
0.1381	ε_{F_a}	0.00	0.3333	EF a 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0	0.00
	$b_s \mid k$	58 13 39 13 31 11 31 11 32 14 40 10 50 1		bs ks ks ks ks ks ks ks	03 21 26 13
0.129	$arepsilon_{abs} \mid k \ arepsilon_{abs} \mid k \ arepsilon_{abs} \mid k \ arepsilon_{Fabs} \mid k \ arepsi$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	0.3194	$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	- k	112 113 114 115 117 117 118 118 119 119 119 119 119 119	٥	* * * * * * * * * * * * * * * * * * *	21 21
0.1196	ε_{F_a}	0.015 0.012 0.005 0.006 0.006 0.006	0.307	$^{\varepsilon}F_{ab}$	0.007
0	s k	3 12 7 11 7 11 10 9 6 9 6 11 11 14 17 18 18 18 18 18 18 18 18 19 19 19 19 19 19 19 19 19 19		$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	3 20 8 21
0.109	ε_{F_ab}	0.016 0.015 0.012 0.010 0.008 0.007 0.006 0.005	0.2959	$\varepsilon F_a b_b$ 0.008 0.008 0.006 0.005 0.004 0.004	0.00
0	_ k	12 11 11 11 11 11 11 11 11 11 12 13 14 14 15 17 18 18 18 18 19 19 19 19 19 19 19 19 19 19 19 19 19	0.	k 1 1 1 1 1 1 1 1 2 3 4 5 6 7 8 8 8 8 8 8 8 8 8 8 9 9 10	3 21
0.0991	ε_{F_abs}	0.0168 0.012 0.012 0.009 0.009 0.005 0.005 0.005	0.2857	ε_{Fabs} 0.008(0.003
0.0	_ k	112 111 113 10 6 6 7 7 7 7 117 118 118 114	0.:	7 11 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	21 (
0.0893	$^{\varepsilon}F_{abs}$	0.0171 0.0126 0.0126 0.0096 0.008 0.008 0.0064 0.0058	0.2764	EFabs 1.0095 1.0087 1.0075 1.0052 1.0052 1.0048 1.0036	0.0034
0.0	, k	111 100 113 100 100 100 100 100 100 100	0.2	* 4111111111111111111111111111111111111	20 (
0.0799	F_{abs}	0.0066 0.0066 0.0084 0.0088 0.0088 0.0066 0.0066 0.0066	0.2678	Fabs .0099 .0086 .0083 .0053 .0053 .0044	.0037
0.0	*	11 0 12 0 10 0 13 0 6 0 6 0 7 7 7 0 17 0 9 0	0.2		20 0
969	F_{abs}	0.0182 0.0165 0.0147 0.0107 0.0089 0.0077 0.0071 0.0067	298	$\begin{array}{c c} F_{abs} \\ \hline \\ 0.0102 \\ 0.0085 \\ 0.0054 \\ 0.0053 \\ 0.0041 \\ 0.0041 \\ \end{array}$	0035
0.0698		0000 00000	0.2598		> 이
269	Fabs	0.0187 0.0164 0.0157 0.0102 0.0094 0.0083 0.0068 0.0068	154	$ \begin{array}{c c} k & \varepsilon_{Fabs} \\ \hline 14 & 0.0106 \\ \hline 13 & 0.0102 \\ \hline 15 & 0.0082 \\ \hline 10 & 0.0068 \\ \hline 8 & 0.0059 \\ \hline 9 & 0.0054 \\ \hline 7 & 0.0048 \\ \hline 16 & 0.004 \\ \hline 17 & 0.0048 \\ \hline 17 & 0.0048 \\ \hline 10 & 0.0048 \\ \hline 10 & 0.0048 \\ \hline 11 & 0.0048 \\ \hline 12 & 0.0048 \\ \hline 13 & 0.0048 \\ \hline 14 & 0.0048 \\ \hline 15 & 0.0048 \\ \hline 15 & 0.0048 \\ \hline 16 & 0.0048 \\ \hline 17 & 0.0048 \\ \hline 17 & 0.0048 \\ \hline 18 & 0.$	0039
0.0597	k e	11 0.0187 12 0.0164 10 0.0157 6 0.0102 13 0.0101 5 0.0094 7 0.0074 17 0.0068 4 0.0066	0.2454	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	19 0. 20 0.
66	Tabs	0.0195 11 0.0192 11 0.0187 11 0.0197 10 0.0176 10 0.0167 12 0.0164 12 0.0155 12 0.0155 10 0.0157 10 0.0155 10 0.0187 11 0.0108 12 0.0108 13 0.0101 13 0.0095 13 0.0094 14 0.0083 14 0.0097 15 0.0094 15 0.0097 15 0.0071 17 0.0074 17 0.0075 17 0.0075 17 0.0068 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0069 17 0.0075 17 0.0069	68	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$)037)036
0.0499	$k \in E_I$	11 0.0192 10 0.0167 12 0.0162 6 0.0103 13 0.0098 9 0.0094 4 0.007 17 0.0069	0.2389	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	20 0.0 16 0.0
66	abs	0.0195 0.0176 0.0159 0.0106 0.0104 0.0089 0.0076 0.007	69	$\begin{array}{c c} \varepsilon F_{abs} \\ \hline 0.0114 \\ 0.0109 \\ 0.0087 \\ 0.0063 \\ 0.0056 \\ 0.0056 \\ 0.0052 \\ 0.0042 \\ \end{array}$	038
0.0399	k ε_F	11 0.0195 10 0.0176 12 0.0159 9 0.0106 6 0.0104 5 0.0103 13 0.008 4 0.0076 17 0.007	0.2269	k e _F 113 0.0 114 0.0 112 0.0 115 0.0 7 0.0 9 0.0	0.0
	$k \ \varepsilon_{Fabs} \mid k \ \varepsilon_{Fabs} \mid k$		33	$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	19 0.0046 9 0.0046 19 0.0044 6 0.0038 20 0.0037 19 0.0039 20 18 0.0045 19 0.0045 6 0.0043 20 0.0037 16 0.0036 20 0.0037 19
0.03	k EF	111 0.0198 110 0.0185 12 0.0156 9 0.0118 5 0.0107 6 0.0105 13 0.0082 4 0.0081 16 0.0071 17 0.007	0.2163	$k \varepsilon_{Fabs}$ 113 0.012 114 0.0109 112 0.0097 115 0.006 7 0.006 9 0.0049 111 0.0048	0.0 9
	. 891	0.02 11 0.0193 11 0.0129 9 0.0129 0.0105 0.00086 11 0.0073 11 0.0073 11 0.0073		$ \begin{array}{c cccc} & & & & & & & & & & & & & & & & & $	046 I
0.02	EF.	11 0.02 10 0.0193 12 0.0153 9 0.0129 5 0.011 6 0.0105 13 0.0075 16 0.0073 17 0.0071	0.2066	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	9 0.0
17	1 89.	10.0201 11 0.02 11 0.0198 10 0.02 10 0.0193 10 0.0155 12 0.0156 9 0.0114 5 0.0115 9 0.0118 10 0.0016 10 0.0016 10 0.0016 10 0.0075 10 0.0071 10 0.0071 11 0.0071 12 0.0071 12 0.0071 13 0.0071 14 0.0068 13 0.0071 13 0.0071 14 0.0071 15 0.0071 1		$\begin{array}{c ccccccccccccccccccccccccccccccccccc$)46)45 19
p = 0.01	ε_{F_a}	1 0.0201 0 0.02 2 0.0149 9 0.0141 5 0.0114 6 0.0104 4 0.0091 6 0.0076 7 0.0071	0.1979	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	0.00
_ d	k	111 122 123 124 146 161 171 171	-	4 122 112 123 124 125 126 136 136 136 136 136 136 136 136 136 13	31 8

Table 2: Table of the 10 largest errors, $\varepsilon_{F_{abs}}$ and which k is comes from, for every p_i , under npq = 10 with continuity correction.

p = 0.01	0.0199	0.0297	0.0395	0.0493	0.059	0.0685	0.0795	0.089	0.0978	0.1086	0.1172	0.1273	0.1394
$k \ \varepsilon_{F_{abs}} _k$	$\varepsilon_{F_{abs}} _k$	$\varepsilon_{F_{abs}}$	$ k \ \varepsilon_{F_{abs}} $	k \in F_{abs} $ k$ \in F_{abs}	EFabs	t EFabs h	$k \in \mathbb{F}_{abs} \mid k$	EFabs K	EFabs k	$\varepsilon_{F_{abs}} _k$	$\epsilon_{F_{abs}}$	k $\varepsilon_{F_{abs}}$	$ k \ \varepsilon_{F_a} $
2 0.0356 2	0.0356 2 0.0343 3 0.0332 3 0.033	0.0332	0.033	3 0.0328 3 0.0325 3 0.0322 3 0.0318 3 0.0314 3 0.031 3 0.0304 3 0.031	0.0325 3	3 0.0322 3	3 0.0318 3	0.0314 3	0.031 3	0.0304 3		3 0.0294	3 0.0286
0 0.0245 C	0.0245 0 0.0242 0 0.0239 0 0.0236	0.0239	0 0.0236	0 0.0233 0	0.0229 0	0 0.0225 0	0 0.0221 0	0.0216	0.0212 0 0	0.0207		0 0.0196	2 0.0181
6 0.0117 6	0.0117 6 0.0116 6 0.0114	0.0114	6 0.0112	4 0.0111 4	0.0116 4	4 0.0121 4	4 0.0126 4	0.0131 4	0.0135 4	0.014 4		4 0.0147	4
4 0.009 4	0.009 4 0.0096 4 0.0101 4 0.0106	0.0101	_	6 0.0111 6			$^{\circ}$		0.0101 6	0.0098		6 0.0094	9
5 0.009 E	0.009 5 0.0086 5	0.0082	0.0082 5 0.0077	5 0.0072 7	0.0071	7 0.007 7		7 0.007 7		0.0069 7		7 0.0068	Н
7 0.0072 7	0.0072 7 0.0072 7 0.0071 7 0.0071	0.0071	_	7 0.0071 5	0.0068 5	5 0.0063 5	5 0.0058 5	0.0053 5	0.0048 1	0.0053 1	0.0059	1 0.0067	7
1 0.0047 1	0.0047 1 0.0035 8	0.003 8		8 0.003 8	0.003 8	8 0.003 8	8 0.003 1	0.0036 1	0.0044 5	0.0042 5	0.0038	5 0.0032	8 0.003
8 0.0031 8		0.0024	0.003 1 0.0024 1 0.0013	9 0.001 9		1 0.0017 1	0.001 1 0.0017 1 0.0028 8	8 0.003 8	0.003 8	0.003 8	8 0.003	8 0.003 5	5 0.0025
9 0.001 9	0.001 9	0.001	9 0.001 10	10 3e-04 1		9 0.001 9	9 0.001 9	0.001 9		0.001 9		9 0.001	9 0.001
0.1464	0.1542	0.1629	0.1727	0.1838	0.1965 0.2113		0.2288	0.25	0.2764	0.311	0.3613	0.5	
$k \ \varepsilon_{F_{abs}} _k$	$\varepsilon_{F_{abs}} _k$	EFabs	$ k \in F_{abs} $	$k \varepsilon_{Fabs} k $	EFabs	k EFabs h	$k \in \mathcal{E}_{Fabs} \mid k$	$\varepsilon_{F_{abs}} _k$	$\varepsilon_{F_{abs}} k$	$\varepsilon_{F_{abs}} _k$	$\epsilon_{F_{abs}}$	k $\varepsilon_{F_{abs}}$	_
3 0.0281 3	0.0276 3	0.0269	3 0.0262	$0.0281 3 \ 0.0276 3 \ 0.0269 3 \ 0.0262 \ 3 \ 0.0252 3 \ 0.0241 3 \ 0.0227 3 \ 0.0209 3 \ 0.0186 4$	0.0241	3 0.0227 3	3 0.0209 3	0.0186		0.0146	0.016 4 0.0146 4 0.0112 9 0.0024	9 0.0024	_
0 0.0185 0	0.018 0	0.0175	0.018 0 0.0175 0 0.0169	4 0.0164 4 0.0166 4 0.0167 4 0.0168 4 0.0166 3	0.0166	4 0.0167	4 0.0168 4	0.0166		0.0155 3 0.0114 1	0.0079	2 0.0024	
2 0.0172 2	0.0172 2 0.0161 4 0.0159 4 0.0161	0.0159		0 0.0161 0 0.0153 0 0.0143 0 0.0131 0 0.0116 1	0.0153	0 0.0143 (0 0.0131 0	0.0116	0.0108	0.0099 5	0.0068	10 0.0015	
4 0.0154 4	4 0.0156 2 0.0149 2 0.0135	0.0149	2 0.0135	2 0.012 2	0.0104	1 0.0106 1	1 0.0109 1	0.011 0	0 8600.0	0.0076 3	3 0.0057	1 0.0015	
6 0.0088 6	0.0088 6 0.0086 1 0.0088 1	0.0088	1 0.0093	1 0.0098 1	0.0102 2		0.0085 2 0.0063 7	0.0055 7	0.005 5	0.0061 0		3 0.0015	
1 0.0079 1	0.0079 1 0.0084 6 0.0083 6	0.0083	0.008	0.0077	0.0072 6	9 20000 9	9 900.0 9	0.0051 5	0.0048 7	0.0042 2	0.0043	8 0.0015	
7 0.0066 7	0.0066 7 0.0066 7 0.0065 7 0.0064	0.0065	_	7 0.0063 7	0.0062 7	7 0.006 7	7 0.0058 2	0.0039 6	8 0.0039 8	0.0024 7	7 0.0027	6 8e-04	
8 0.003 8	3 0.003 8	0.003 8	0.003	8 0.003 8	0.0029	0.003 8 0.0029 8 0.0029 8	8 0.0029 5	0.0029 5 0.0036 8	0.0026 6	0.0026 6 0.0024 8	3 0.0017	5 8e-04	
5 0.0021 5	0.0021 5 0.0017 5 0.0012 9	0.0012	9 9e-04	9 9e-04 9		5 0.0015	9e-04 5 0.0015 5 0.0025 8 0.0028 2 0.0012 2 0.0017 9	0.0028 2	0.0012 2	0.0017	5e-04	7 6e-04	
0 000	0 000	1	1	1	1					, ,			

Table 3: Table of the 10 largest errors, $\varepsilon_{F_{abs}}$ and which k is comes from, for every p_i , under npq = 3 with continuity correction.

0.1696 0.1792 0.1899	k eFrel k eFre	34.6254 0 36.684 038.5519 0 40.7878 6.9494 1 7.3262 1 7.6559 1 8.0701 9 2524 9 2 3758 9 2 4865 9 2 6176	2 8 8 8 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	10 0 11 0 0	10 88 89 10 10 10 10 10 10 10 10 10 10 10 10 10
87 0.1584	$ ec{r}_{rel} k arepsilon_{Fel} $		1004084	1 0.5	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
0.1381 0.1487	k $\varepsilon_{F_{rel}} k$ ε_{I}	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	4 0.2127 8 0.219 9 0.2111 9 0.212 7 0.2047 7 0.201 0 0.187410 0.190 1 0.1553 10 159 0.3333 0.3492	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
0.129	$_{el} \ k\ arepsilon_{rel} \ ^{l}$	04 029.8146 05 1 6.0582 18 2 1.9578 229 3 0.7009 97 8 0.2199	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	92 9 0.2094 76 7 0.207 16 4 0.1931 3710 0.184611 23 6 0.156811 0.3194	92 9 0.2094 76 7 0.207 3710 0.34611 23 6 0.156811 0.3194
0.109 0.1196	$\varepsilon_{F_{rel}} k$ $\varepsilon_{F_{r}}$	5.9868 028.44 5.5263 1 5.80 1.7799 2 1.87 0.6222 3 0.66 0.2194 8 0.21	0.2055 9 0.20 0.178210 0.18 0.1681 4 0.17 0.1528 6 0.16	0.2959 0.1752 0.1782 0.1782 0.1782 0.1737 1.1528 0.1623 0.2959 0.307	$\mathcal{E}_{rel} = \frac{1}{k} \mathcal{E}_{rel} = \frac{1}{k} \mathcal{E}_{rel}$
0.0991	k $\varepsilon_{F_{rel}} k$	025.7144 020 1 5.2848 1 8 2 1.6985 2 3 3 0.5861 3 0 7 0.5139 8 0.219 8 0	0 0.2034 9 0 0.2034 9 0 0 0.17510 0 0 0.1731 6 0 0.1401 4 0	0.2857 0.2233 / 0.22857 0.2	$ \begin{array}{c cccc} & 0.2153 & 0.0 \\ & 0.2034 & 0.0 \\ & 0.1731 & 0.0 \\ \hline & 0.1401 & 4.0 \\ \hline & 0.2857 & 0.2 \\ \hline & & & & & & & & & & & & & & & & & &$
0.0799 0.0893	$_{il} \mid k$ $\varepsilon_{F_{rel}} \mid$	37 0 24.515 37 1 5.0558 37 2 1.621 38 0 5.516 39 0 5.516 30	0.2103 (0.2149 (0.1991 9 0.2012 9 0.1821 6 0.177910 0.168710 0.1718 6 0.132811 0.136411	0.01903 (0.2149) 0.01901 (0.01779) 0.0168710 (0.1778) 1.0132811 (0.13641) 0.2678 (0.2764)	0.2764 0.2764 0.2764 0.2764 0.2764 0.2764
0.0698 0.0799	$F_{rel} k$ ε_{Fre}	115 0.22.3354 0.23.4367 0.24.515 0.25.7144 0.26.9868 0.28.4404 0.29.8146 135 1.4.636 1.4.8487 1.5.0558 1.5.2848 1.5.283 1.5.8005 1.6.0582 2.1.4781 2.1.5677 2.1.621 2.1.6815 2.1.7799 2.1.8718 2.1.9578 688 3.0.4877 3.0.5202 3.0.5516 3.0.5861 3.0.622 3.0.62	44 9 0.1968 9 0.199 04 6 0.1864 6 0.182 6210 0.165410 0.168 5311 0.12911 0.132	2103 (2.210 1968 9 0.199 1864 6 0.182 165410 0.168 0.12911 0.132 698 0.2678	1968 0.1510 1964 0.182 185410 0.182 165410 0.168 1.129 11 0.132 198 0.2678 Frel k & Fre
0.0597 0.06	$\varepsilon_{F_{rel}} k$	21.3015 0.22.3 4.435 1 4 1.4092 2 1.4 0.4568 3 0.4 0.2186 7 0.2			$ \begin{array}{c} 0.1944 \\ 0.1904 \\ 0.1904 \\ 0.16210 \\ 0.125311 \\ 0.2454 \\ 0.25 \\ \hline \end{array} $
0.0499	$\mid k \mid \varepsilon_{F_{rel}} \mid k$	0 16.9732 0 17.7468 0 18.5666 0 19.4284 0 20.3435 0 21.30 1 3.5794 1 3.7342 1 3.8973 1 4.40578 1 4.2478 1 4.44 2 1.1123 2 1.1065 2 1.2232 2 1.3247 2 1.40 3 0.3227 3 0.3473 3 0.373 3 0.3998 3 0.4278 3 0.45 7 0.221 7 0.2213 7 0.2209 7 0.2203 7 0.2195 7 0.21 6 0.2026 6 0.3141 6 0.375 8 0.3203 7 0.2195 7 0.21	6 0.194 9 0.19 6 0.194 9 0.19 7 0.1919 6 0.19 7 0.158810 0.11 111 0.121711 0.12	6 0.134 9 6 0.194 9 6 0.194 9 6 0.194 9 6 0.1919 6 6 110 0.158810 111 0.1217111 0.2389 C	0 0.2139 0 0.0199
0.0399	$_{rel} \ k\ arepsilon_{Frel}$	018.5666 019.4284 1 3.8973 1 4.0678 2 1.2233 2 1.2825 3 0.373 3 0.3998 7 0.2203 7 0.2203	0.2097 8 0.2111 8 0.2129 8 0.2137 8 0.2064 6 0.2006 6 0.2075 6 0.1975 6 0.1977 9 0.1843 9 0.1869 9 0.1895 9 0.145710 0.148910 0.152210 0.155510 0.1454 5 0.139 5 0.1322 5 0.125111	125 8 0.2137 307 6 0.1975 869 9 0.1895 522 10 0.1555 322 5 0.1251 3 0.2269	25 8 0.2134 107 6 0.1975 108 20.1895 109 0.1555 109 0.155
0.02 0.03	$\varepsilon_{F_{rel}} _{k}$	7.7468 018.56 3.7342 1 3.89 3.1665 2 1.22 3.3473 3 0.32 3.2213 7 0.22	2.2111 8 0.2125 8 2.2036 6 0.2007 6 3.1843 9 0.1869 9 3.148910 0.152210 0.139 5 0.1322 5	8 0.2111 6 0.2036 9 0.1843 9 0.1489 10 0.1489 5 0.139 5 0.139 7 0.2066 0 0.2163	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
p = 0.01 0.	k $\varepsilon_{F_{rel}} k$	016.9732 017.7468 018.5666 019.4284 020.3435 021.30 1 3.5794 1 3.7342 1 3.8973 1 4.40678 1 4.2478 1 4.4 2 1.1123 2 1.1665 2 1.2233 2 1.2825 2 1.3447 2 1.40 3 0.3227 3 0.3473 0.373 3 0.3998 3 0.4278 3 0.45 7 0.2216 7 0.2213 7 0.2203 7 0.2203 7 0.2195 7 0.21	8 0.2097 8 0.2111 8 0.2125 8 0.2157 8 0.2144 8 0.2 9 0.2064 6 0.2036 6 0.2007 6 0.1975 6 0.194 9 0.19 9 0.1817 9 0.1843 9 0.1869 9 0.1895 9 0.1919 6 0.19 10 0.145710 0.148910 0.152210 0.155510 0.158810 0.1 5 0.1454 5 0.139 5 0.1322 5 0.125111 0.121711 0.12	8 0.2097 8 0 6 0.2064 6 0 9 0.1817 9 0 10 0.145710 0 5 0.1454 5 0.1979 0.2	$\begin{array}{c} 0.2097 \\ 0.2094 \\ 0.2064 \\ 0.9 \\ 0.1817 \\ 0.0 \\ 0.145710 \\ 0.1454 \\ \hline 0.1979 \\ 0.1979 \\ 0.25 \\ 0.285 \\ 0.285 \\ 0.280 \\ $

Table 4: Table of the 10 largest errors, $\varepsilon_{F_{rel}}$ and which k is comes from, for every p_i , under npq = 10 without continuity correction.

													1				
p = 0.01 0.02	0.03	0.0399	0.0499	0.0597	0.0698	0.0799 0	0.0893	0.0991	0.109	0.1196	0.129	0.1381	0.1487	0.1584 (0.1696	0.1792	0.1899
rel k EFre	$_{l} \ k\ \varepsilon_{F_{rel}} $	k $\varepsilon_{F_{rel}}$	EFrel k EFrel k EFrel k EFrel k EFrel k EFrel k EFrel	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} _{k}$	ε $\varepsilon_{F_{rel}} k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} k$	ε_{Frel}
0 29.7208	0 31.1973 0 32.7709 1 6.7619 1 7.0685 2 2.3926 2 2.4984 3 1.0166 3 1.0448	2.7709 034.4347 (7.0685 1 7.3907 2.4984 2 2.6091 1.0648 3 1.1151 2.4747 4 0.614 1.0648 3 1.1151 2.4747 4 0.614	0 36.212 0 1 7.7327 1 2 2.7261 2 3 1.1681 3	38.084 0 40.11 8.0908 1 8.47 2.848 2 2.97 11.232 3 1.28	27.7.2 89.22 80.22 80.22	0 42.297 0 1 8.8893 1 2 3.1178 2 3 1.3445 3	44.4454 0 9.2928 1 3.2532 2 1.4051 3	46.851 0 9.7419 1 3.4032 2 1.4721 3	49.4206 0 10.2188 1 3.5617 2 1.5425 3	52.3782 0 10.7642 1 3.7421 2 1.6224 3	55.1954 0 11.2805 1 3.912 2 1.6975 3	58.0921 0 11.8083 1 4.0849 2 1.7735 3	61.6752 0 12.4572 1 4.2964 1 1.8663 1 0 0 0 0 0 0	65.2141 0 13.0941 1 4.503 2 1.9566 3	69.5729 0 13.8735 1 4.7545 2 2.066 3	73.5645 0 14.5828 1 4.9822 2 2.1647 3	78.3871 15.4344 5.2542 2.2822
4 7 0 0 8 0	7.4431 4 0.4747 7.1804 5 0.1951 7.0468 6 0.0553 7.0447 9 0.0442 0.042 8 0.0398 7.035910 0.0364	0.4/4/f 4 0.5014 4 0.01951 5 0.2104 5 0.0553 6 0.0436 9 0.0398 8 0.037510 0.036410 0.0368 8	0.5295 0.2265 0.0736 0.0428 0.0371		0.2612 0.0941 0.0408 0.03751		0.2986 5 0.2986 5 0.1164 6 0.03810 0.0374 9	0.3189 5 0.3189 5 0.1286 6 0.0372 7 0.036310	0.7269 4 0.3403 5 0.1414 6 0.0373 7 0.0368 10 0.0344 9	$\begin{array}{c} 0.7689 \\ 0.3645 \\ 0.156 \\ 0.0461 \\ \hline 0.036210 \\ 0.032111 \end{array}$	0.3873 5 0.1698 6 0.0544 7 0.0356 10 0.0304 11	0.8479 4 0.4102 5 0.1837 6 0.0629 7 0.034810	0.5363 0.4382 0.2008 0.0733 7 0.033610 0.0308111		1.0002 0.4982 0.2374 0.096 7 0.0306110 0.0306110	0.5278 0.2555 0.1073 0.0289	0.5629 0.2771 0.1208 0.0354 0.0298
0.1979 0.2066	0.2163	0.2269	0.2389	0.2454	0.2598	0.2678 0	0.2764 C	0.2857 (0.2959	0.307	0.3194 (0.3333	0.3492	0.3679	0.3909	0.4219	0.5
$k \ \varepsilon_{Frel} \mid k \ \varepsilon_{Frel} \mid k$	$_{l} \ k\ \varepsilon_{F_{rel}} $	k ε_{Frel}	k $\varepsilon_{F_{rel}} k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} _{k}$	$\varepsilon = \varepsilon_{F_{rel}} k$	$\varepsilon_{F_rel} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_rel} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} _{k}$	$\varepsilon_{F_{rel}} k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \ k$	$\varepsilon_{F_{rel}} \mid k$	$\varepsilon_{F_{rel}}$
0.82.1989 0.86.6195 0.91.8044 0.97.9661 0.10.5.478 0.119.9526 0.12.5.805<	5 0 91.8044 1 1 17.7774 5 2 5.9954 7 3 2.6001 4 4 1.2758 4 5 0.6573 1 6 0.3352 5 7 0.1576 5 8 0.0584	1.8044 097.9661 17.777 118.8412 5.956001 3 2.7423 1.2758 4 1.3487 0.6575 5 0.6993 0.1576 7 0.1741 0.1576 7 0.1741 0.05741 0.0581	0 105.4031 0 1 20.1171 1 2 6.727 2 3 2.9111 3 5 0.749 5 6 0.3919 6 7 0.1938 7 8 0.0815 9	20.8552 1 6.9561 2 3.0079 3 1.4844 4 0.7774 5 0.4094 6 0.205 7 0.0888	22.5896 7.4919 3.2335 3.2335 3.2335 0.8433 0.8433 0.2314 0.03348	0126.053 01 123.6179 1 2 7.8079 2 3 3.366 3 4 1.666 3 5 0.8819 5 6 0.4742 6 7 0.2468 7 7 0.2468 7 9 0.0412 9	24.7821 1 8.1642 2 3.515 3 1.7419 4 0.9252 5 0.2642 7 0.1274 6 0.2642 7 0.1274 8 9	26.1109 1 8.5694 2 3.6839 3 1.8273 4 0.974 5 0.5314 6 0.2839 7 0.1405	27.6424 1 9.0342 2 3.8769 3 1.9246 4 1 1.0297 5 0.5659 6 0.3065 7 0.1557 9 0.1554	160.9731 9.5734 2.0368 4.1001 3.2.0368 1.0938 5.06057 6.0325 7.0077 8.001728	173.8113 31.5357 10.2072 2.1677 4.3612 3.1686 1.1686 0.6552 0.6552 0.0552 0.0953	189.3227 34.0669 10.964 2.323 4.6718 2.323 1.257 0.7072 6 0.399 7	208.4603 37.1685 11.886 5.0485 3.25107 1.3638 6.0.4428 0.2478 0.2478 0.2478	232.7278 0. 41.0727 1 13.0394 2 5.5174 3 2.7436 4 1.4962 5 0.856 6 0.4973 7 0.2848 0	264.7054 0 46.177 1 14.5375 2 6.1233 3 3.0436 4 1.6663 5 0.962 6 0.5676 7 0.3334 0	2309.5134 16.6055 6.9555 3.4543 1.1074 6.06647 7.006647	382.9711 64.8288 19.9613 8.304 4.1207 2.2788 1.3466 0.8267 0.5164

Table 5: Table of the 10 largest errors, $\varepsilon_{F_{rel}}$ and which k is comes from, for every p_i , under npq = 10 with continuity correction.

p = 0.01	0.0199	0.0199 0.0297		0.0395	0.0493 0.059	0.0	920	0.0685		0.0795	0.089		826	0.1086	0.0978 0.1086 0.1172		0.1273	0.13	0.1394
k $\varepsilon_{F_{rel}}$	k EFrel k	k ε_{Frel}	$ k \varepsilon_F$	$_{rel}$ $ k $	$\varepsilon_{F_{rel}}$	k	F_{rel}	& EFre	1 k E	$F_{rel} \mid k$	ε_{Frel}	k E	$F_{rel} \mid k$	ε_{Frel}	$ k \in E_{I_{\eta}}$	rel k	$\varepsilon_{F_{rel}}$	k E	F_{rel}
$\begin{array}{ccc c} 0 & 0.514 & 0 \\ 2 & 0.0856 & 2 \end{array}$	_	$0.533 \mid 0.0.5523 \mid 0.0.5721 \mid 0.0.5924 \mid 0.0.6132 \mid 0.0.634 \mid 0.0.6588 \mid 0.0.6809 \mid 0.0.7019 \mid 0.0.7283 \mid 0.0.7501 \mid 0.0.842 \mid 2.0.0833 \mid 2.0.0797 \mid 2.0.0781 \mid 2.0.0762 \mid 2.0.0741 \mid 2.0.0731 \mid 2.0.0792 \mid 2.0.0678 \mid 2.0.0657 \mid 0.0657 \mid 0.0657 \mid 0.0657 \mid 0.0678 \mid 0.0678 \mid 0.0657 \mid 0.0678 \mid 0.0678 \mid 0.0678 \mid 0.0678 \mid 0.0677 \mid $	0 0.5	721 0 813 2	0.5924 0.0797	0 0.	6132	0.63	2 2 2 0 0 0 0	6588 C	0.6809	0 0.7	7019 0 0702 2	0.7283	0 0.75		$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	0 0.8	8088
	3 0.0527	0.0527 3 0.053 3 0.0533 3 0.0536 3 0.0538 3 0.054 3 0.0541 3 0.0542 3 0.0542 3 0.0542 3 0.0542	3 0.0	533 3	0.0536	3 0.	0538	3 0.05	4 3 0.	0541 3	0.0542	3 0.0	0542 3	0.0542	3 0.05		3 0.0541 2 0.0599	2 0.0	0599
$\begin{vmatrix} 1 & 0.0243 \end{vmatrix} 1 \\ 6 & 0.0121 \end{vmatrix} 6$	_	0.0189 1 0.0132 4 0.0133 4 0.0141 4 0.0148 4 0.0155 1 0.0181 1 0.0247 1 0.031 1 0.039 1 0.0455 0.012 4 0.0126 6 0.0117 6 0.0115 6 0.0113 6 0.0111 4 0.0163 4 0.017 4 0.0177 4 0.0184 4 0.0191	4 0.0 6 0.0	133 4	0.0141	4 9 0 0	0148	$\frac{4}{5}$ 0.015	1 0. 1 4 0.	$0181 1 \\ 0163 4$	0.0247	1 0 4	0.031 1	0.039	1 0.04 4 0.01		$\begin{array}{cccccccccccccccccccccccccccccccccccc$	ω 4 0.0	0538
4 0.0112	_	6 0.0118	5 0.0	085 5	0.008	5 0.	0075	0.010	8 6 0.	0109	0.0107	0.0	0105 6	0.0103	6 0.01		6 0.0098 6 0.0095	0.0	000
5 0.0099	$0.0099 \ 5\ 0.0094 \ 5\ 0.009 \ 1\ 0.0075 \ 7\ 0.0072 \ 7\ 0.0072 \ 7\ 0.0071 \ 7\ 0$	5 0.009	1 0.0	075 7	0.0072	7 0.	0072	7 0.007	1 7 0.	0071 7	7 0.0071	7 0	7 200.0	0.007	7 0.00		7 0.0069 7 0.0068	7 0.0	.006
	$0.0073 \mid 7 0.0073 \mid 7 0.0072 \mid 7 0.0072 \mid 8 0.003 \mid 1 0.0046 \mid 5 0.007 \mid 5 0.0064 \mid 5 0.0069 \mid 5 0.0059 \mid 5 0.0054 \mid 5 0.0048 \mid 5 0.0043 \mid 5 0.0043 \mid 5 0.0074 \mid 5 \mid $	7 0.0072	7 0.0	072 8	0.003	1 0.	0046	0.00	7 5 0.	0064 5	0.0059	5 0.0	0054 5	0.0048	5 0.00		_	0	0.00
8 0.0031	8 0.0031	0.0031 8 0.0031 8 0.0031 1 0.0015 8	8 0.0	031 1	0.0015	∞ c	003	8 0.00	8 -	3 800.0	0.003 8 0.003 8 0.003 8 0.003 8 0.003 8 0.003 8 0.003	о о «	0.003 8	0.003	8 0.0	003	0.003 5 0.0028	υ 0.0	005
9 0.001	8 0.001	9 0.001	9	2001	0.001	- 1	100.	0.00	6 1	3 700.	0.001	8	100.	0.001	9 0.1	4		2	
0.1464	0.1542	0.1542 0.1629 0.1727 0.1838 0.1965 0.2113 0.2288	0.17	27	0.1838	0.1	965	0.2113	0.2	288	0.25	0.2	0.2764	0.311	0.311 0.3613	65	0.5		
k $\varepsilon_{F_{rel}}$	k cFrel	k ε_{Frel}	$ k \varepsilon_F$	$_{rel}$ $ k $	$\varepsilon_{F_{rel}}$	k	F_{rel}	& EFre	$_{l}$ $ k $	$F_{rel} \mid k$	ε_{Frel}	k E	$F_{rel} \mid k$	ε_{Frel}	$ k \in F_{\eta}$	rel k	$\varepsilon_{F_{rel}}$		
0 0.8281	$0.08281 \mid 0.08499 \mid 0.08748 \mid 0.0936 \mid 0.09373 \mid 0.09773 \mid 0.10255 \mid 0.1085 \mid 0.11605 \mid 0.12602 \mid 0.13994 \mid 0.16142 \mid 0.1614$	0 0.8748	0.0	036 0	0.9373	0 0	9773 0	0 1.025	5 0 1	085 0	1.1605	0 1.5	2602 0	1.3994	0 1.61	142 C	2.0641		
1 0.0692	1 0.0759	1 0.0836	1 0.0	925 1	0.1029	1 0.	1153	0.130	4 1 0.	1491 1	0.173	1 0.5	2048 1	0.2498	1 0.32	203 1	0.4769		
2 0.0579	$0.0579 \ \ 2 \ 0.0557 \ \ 3 \ 0.0557 \ \ 3 \ 0.0531 \ \ 3 \ 0.0526 \ \ 3 \ 0.052 \ \ 3 \ 0.0512 \ \ 3 \ 0.0512 \ \ 3 \ 0.0484 \ \ 3 \ 0.0459 \ \ 3 \ 0.042 \ \ 3 \ 0.0424 \ \ 3 \ 0.0444$	3 0.0531	3 0.0	526 3	0.052	3 0.	0512	3 0.0	5 3 0.	0484 5	0.0459	3 0	0.042 3	0.0354	2 0.04		2 0.1227		
3 0.0536	$0.0536 \begin{vmatrix} 3 & 0.0534 \end{vmatrix} 2 & 0.053 \end{vmatrix} 2 & 0.0499 \begin{vmatrix} 2 & 0.0462 \end{vmatrix} 2 & 0.0417 \end{vmatrix} 2 & 0.0361 \begin{vmatrix} 2 & 0.029 \end{vmatrix} 4 & 0.0263 \end{vmatrix} 4 & 0.0263 \end{vmatrix} 4 & 0.0268 \end{vmatrix} 4 & 0.0266 \end{vmatrix} 4 & 0.0246 \end{vmatrix} = 0.0246 \end{vmatrix} 4 & 0.0268 \end{vmatrix} + 0.0266 \end{vmatrix} 4 & 0.0268 \end{vmatrix} + 0.0266 \end{vmatrix} + 0.0268 \end{vmatrix} + 0.0266 \end{vmatrix} + 0.0268 \end{aligned} + 0.0268 \end{vmatrix} + 0.026$	2 0.053	2 0.0	499 2	0.0462	2 0.	0417	2 0.036	1 2 0	0.029	0.0263	4 0.0	0268 4	0.0266	4 0.0	34 3	0.02		
4 0.0211	$0.0211 \ \ 4 \ 0.0216 \ \ 4 \ 0.0222 \ \ 4 \ 0.0228 \ \ 4 \ 0.0234 \ \ 4 \ 0.0241 \ \ 4 \ 0.0249 \ \ 4 \ 0.0256 \ \ 2 \ 0.0196 \ \ 2 \ 0.0068 \ \ 2 \ 0.0122 \ \ 3 \ 0.0226 \ \ 0.0068 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ \ 0.0026 \ \ 0.0068 \ $	4 0.0222	4 0.0	228 4	0.0234	4 0.	0241	4 0.024	9 4 0.	0256 2	0.0196	2 0.0	0068 2	0.0122	3 0.02		4 0.0031		
6 0.0093	$0.0093 \mid 6.0.0091 \mid 6.0.0088 \mid 6.0.0085 \mid 6.0.0081 \mid 6.0.0077 \mid 6.0.0071 \mid 6.0.0064 \mid 7.0.0056 \mid 5.0.0061 \mid 5.0.0081 \mid 5.0.0099 \mid 9.0.0099 \mid $	6 0.0088	6 0.0	085 6	0.0081	6 0.	0077	3 0.007	.1 6 0.	0064 7	0.0056	20.0	0061 5	0.0081	5 0.00	5 660	9 0.0024		
7 0.0068	0.0068 7 0.0066 7 0.0066 7 0.0066 7 0.0063 7 0.0063 7 0.0065 7 0.0051 7 0.0059 7 0.0059	7 0.0066	7 0.0	066 7	0.0065	0,	0063	7 0.006	<u> </u>	0.006	0.0055	7 0.1	0051 7	0.0043	7 0.00	029	5 0.002		
8 0.003 8	8 0.003	0.003 8 0.003 8 0.003 8 0.003 8 0.0029 5 0.003 5 0.0044 6 0.0043 8 0.0018 8 0.0018	× 0	003	0.003	× 0	003	5 0.002	000	7.003	0.0044	9 0	0043 6	0.0026	× .	210	8 0.0016		

Table 6: Table of the 10 largest errors, $\varepsilon_{F_{rel}}$ and which k is comes from, for every p_i , under npq=3 with continuity correction.