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A note on Wilson coverage intervals for proportions estimated from complex samples

Phillip S. Kott¹

Abstract

This note discusses the theoretical foundations for the extension of the Wilson two-sided coverage interval to an estimated proportion computed from complex survey data. The interval is shown to be asymptotically equivalent to an interval derived from a logistic transformation. A mildly better version is discussed, but users may prefer constructing a one-sided interval already in the literature.

Key Words: Effective sample size; Confidence interval; Logistic transformation.

1 Introduction

Brown, Cai and Dasgupta (2001) show that a method proposed by Wilson (1927) can produce reasonably well-behaved two-sided coverage intervals for a proportion under simple random sampling *with* replacement. Section 2 of this note discusses the theoretical foundations for extending this interval-construction method to estimated proportions computed from a complex survey. Section 3 shows that such a Wilson-type interval can be asymptotically equivalent to an interval derived from a logistic transformation. Section 4 offers some concluding remarks.

The term "coverage interval" is used here in place of the more common "confidence interval" because a 95% Wilson coverage interval does not attempt to cover the true proportion at least 95% of the time no matter what that proportion is. Instead, it merely tries to cover the true proportion 95% of the time for reasonable values of the true proportion. For some values it overcovers, for others it undercovers as shown in Brown et al. (2001). By limiting its applicability to two-sided coverage intervals, the Wilson methodology is (mostly) able to ignore the asymmetry of the distribution of an estimated proportion.

2 The extension

It is not hard to generalize Wilson coverage intervals (also called "score intervals") to complex survey data. See, for example, Kott and Carr (1997). As with the Wilson itself, one simply solves this equation for the true proportion P:

$$\frac{(p-P)^2}{\left[\frac{P(1-P)}{n^*}\right]} \le z_{1-\alpha/2}^2,$$
(2.1)

where p is a consistent estimator for P under probability-sampling theory, and $z_{1-\alpha/2}$ is the Normal z score for $(1 - \alpha/2)$ given the goal is to produce a $(1 - \alpha)$ % coverage interval (α is often set at 0.05). The missing piece to equation (2.1) is n^* , the so-called "effective sample size", which in the standard Wilson formulation is the sample size n. In our more general context, $n^* = p(1-p)/\operatorname{var}(p)$, where $\operatorname{var}(p)$ is a consistent estimator for the variance of p, $\operatorname{Var}(p)$.

In order to calculate n^* , we need both p(1-p), and $\operatorname{var}(p)$ to be positive. In addition, let us assume that $1/n^* = O_p(1/n^a)$ for some positive $a \le 1$, $p - P = O_p(1/\sqrt{n^*})$, $0 < \operatorname{Var}(p) = O(1/n^*)$, and $\operatorname{var}(p)/\operatorname{Var}(p)$ is $1 + O_p(1/\sqrt{n^*})$. Note that the last three are always true under simple random sampling with replacement so long as $P(1-P) \ge B > 0$.

Dropping $O_P(1/[n*]^{3/2})$ terms, but allowing p(1-p) to be small (effectively $o_P(1)$), one can derive this Wilson-like interval for P from equation (2.1):

$$p + \frac{1-2p}{n^*} \frac{z_{1-\alpha/2}^2}{2} - z_{1-\alpha/2} \left(\frac{p(1-p)}{n^*} + \frac{z_{1-\alpha/2}^2}{4(n^*)^2} \right)^{1/2} \le P$$
$$\le p + \frac{1-2p}{n^*} \frac{z_{1-\alpha/2}^2}{2} + z_{1-\alpha/2} \left(\frac{p(1-p)}{n^*} + \frac{z_{1-\alpha/2}^2}{4(n^*)^2} \right)^{1/2}.$$
(2.2)

We can call this the "complex-sampling Wilson coverage interval". WesVar (2007) computes a variant of this interval that does not drop $O_P(1/[n^*]^{3/2})$ terms. It is dropped here because other terms of that size will be dropped later in this note.

If it is reasonable to drop $O_p(1/[n^*]^{3/2})$ terms in deriving equation (2.2), one can also safely ignore the difference between 1/n and 1/(n-1). Under simple random sampling *without* replacement, $n^* = n/(1-f)$ (or (n-1)/(1-f)) where f is the sampling fraction. When f is very small, the distinction between with and without replacement sampling can be ignored.

Observe that under simple random sampling with replacement, the denominator of the pivotal appearing on the left-hand side of equation (2.1) has no variance at all. By contrast, the denominator in the traditional Wald pivotal, var (p) = p (1-p)/(n-1), can have considerable variance, especially when p or 1-p is small. That is why Wilson intervals have superior performance under simple random sampling, whether with or without replacement.

That superiority carries over to complex sampling (see, for example, Kott, Andersson and Nerman, 2001), where the pivotal's denominator is

$$\frac{P(1-P)}{n^*} = \operatorname{var}(p) \frac{P(1-P)}{p(1-p)} = \operatorname{var}(p) \left[1 - \frac{(p-P) - (p^2 - P^2)}{p(1-p)} \right]$$
$$= \operatorname{var}(p) \left[1 - \frac{(p-P) - (p-P)(p+P)}{p(1-p)} \right]$$
$$= \operatorname{var}(p) - \frac{1-2P}{n^*}(p-P) + \operatorname{O}_p(1/[n^*]^2),$$

which is likely to have less variance than var (p) in most applications. For an intuition into why this is so, observe that a putative variance estimator of the form var₁ (p) = var(p) - b(p-P) is minimized when b = Cov [var(p), p]/Var(p). Under simple random sampling, whether with or without replacement, b is exactly $(1-2P)/n^*$.

Although the minimizing b is not exactly equal to $(1-2P)/n^*$, under more complex sampling designs, the optimal b is likely to be closer to $(1-2P)/n^*$ than to 0. It is thus not surprising that the variance of var $(p) - [(1-2P)/n^*](p-P)$ will usually be less than the variance of var (p). Nevertheless, a slight improvement on the complex-sampling Wilson coverage interval can be made by replacing n^* in equation (2.2) by

$$\tilde{n} = [(1-2p) \operatorname{var}(p)]/\operatorname{cov}[\operatorname{var}(p), p]$$

when $\operatorname{cov}[\operatorname{var}(p), p]$, a consistent estimator for $\operatorname{Cov}[\operatorname{var}(p), p]$, exists (see Kott et al., 2001).

As with the standard Wilson, the center of the complex-sample Wilson interval in equation (2.2) is slightly different from p when p is not $\frac{1}{2}$:

$$C = p + \frac{1 - 2p}{n^*} \frac{z_{1 - \alpha/2}}{2}$$

Its length L appears longer than the Wald's:

$$L = z_{1-\alpha/2} \left(\frac{p(1-p)}{n*} + \frac{z_{1-\alpha/2}^2}{4(n*)^2} \right)^{1/2} > z_{1-\alpha/2} \left(\frac{p(1-p)}{n*} \right)^{1/2}.$$

When $P(1-P) \ge B > 0$, however,

$$\left(\frac{p(1-p)}{n^*} + \frac{z_{1-\alpha/2}^2}{4(n^*)^2}\right)^{1/2} = \left(\frac{p(1-p)}{n^*}\right)^{1/2} \left(1 + \frac{\frac{1}{4}z_{1-\alpha/2}^2}{n^*p(1-p)}\right)^{1/2} \\ = \left(\frac{p(1-p)}{n^*}\right)^{1/2} + o_p\left(\frac{1}{n^*}\right).$$
(2.3)

3 The logistic transformation

The complex-sampling Wilson coverage interval turns out to be very similar to this two-sided coverage interval derived using a logistic transformation (see Brown et al., 2001):

$$f^{-1}\left\{f(p) - z_{1-\alpha/2}\sqrt{\operatorname{var}\left[f(p)\right]}\right\} \le P \le f^{-1}\left\{f(p) + z_{1-\alpha/2}\sqrt{\operatorname{var}\left[f(p)\right]}\right\},$$
(3.1)

where $f(p) = \log(p) - \log(1-p)$, and $\operatorname{var}[f(p)] = [f'(p)]^2 \operatorname{var}(p) = [1/p + 1/(1-p)]^2 p(1-p)/n^* = 1/[n^*p(1-p)]$. The original rationale for this interval appears to be that it has this desirable property: it cannot contain values less than 0 or greater than 1, which would be nonsensical for a proportion.

The left-hand side of equation (3.1) can be rewritten as g(x-h), where

$$g(y) = f^{-1}(y) = [1 + \exp(-y)]^{-1}, x = f(p) = \log\left(\frac{p}{1-p}\right),$$

and

$$h = \frac{z_{1-\alpha/2}}{\sqrt{n*p(1-p)}} \,.$$

The first and second derivatives of g(y) are g'(y) = g(y)[1 - g(y)], and g''(y) = g(y)[1 - g(y)][1 - 2g(y)]. Invoking the mean value theorem, there is an h^* between 0 and h such that

$$g(x-h) = g(x) - g'(x)h + \frac{1}{2}g''(x-h^*)h^2$$

$$= p - p(1-p)\frac{z_{1-\alpha/2}}{\sqrt{n^*p(1-p)}}$$

$$+ \frac{1}{2}\left[1 + \left(\frac{1-p}{p}\right)e^{h^*}\right]^{-1}\left\{1 - \left[1 + \left(\frac{1-p}{p}\right)e^{h^*}\right]^{-1}\right\}\left\{1 - 2\left[1 + \left(\frac{1-p}{p}\right)e^{h^*}\right]^{-1}\right\}\frac{z_{1-\alpha/2}^2}{n^*p(1-p)}$$

$$= p - p(1-p)\frac{z_{1-\alpha/2}}{\sqrt{n^*p(1-p)}}$$

$$+ \frac{1}{2}\frac{p}{1+(1-p)(e^{-h^*}-1)}\frac{(1-p) - (1-p)(e^{h^*}-1)}{1+(1-p)(e^{h^*}-1)}\frac{(1-2p) - (1-p)(e^{h^*}-1)}{1+(1-p)(e^{h^*}-1)}\frac{z_{1-\alpha/2}^2}{n^*p(1-p)}$$

using

$$\left[1 + \left(\frac{1-p}{p}\right)e^{h^*}\right]^{-1} = \frac{p}{1 + (1-p)(e^{h^*} - 1)}$$

An analogous derivation can be made for the right-hand side of equation (3.1).

Consequently,

$$p + \frac{1-2p}{n^*} \frac{z_{1-\alpha/2}^2}{2} - z_{1-\alpha/2} \left(\frac{p(1-p)}{n^*}\right)^{1/2} + o_p\left(\frac{1}{n^*}\right) \le P$$
$$\le p + \frac{1-2p}{n^*} \frac{z_{1-\alpha/2}^2}{2} + z_{1-\alpha/2} \left(\frac{p(1-p)}{n^*}\right)^{1/2} + o_p\left(\frac{1}{n^*}\right).$$

After invoking the asymptotic equality in equation (2.3) and dropping $o_P(1/n^*)$ terms, the last set of inequalities is equivalent to Wilson interval in equation (2.2) so long as n^* is sufficiently large and P(1-P) > 0, the latter meaning that the true proportion is neither 0 or 1.

4 Some concluding remarks

The asymptotic equivalence of a coverage interval based on a logistic transformation to the theoretically grounded Wilson interval is the main contribution of this paper. Although in the asymptotic framework, P(1-P) is fixed and positive as n^* grows large, in practice it is the size of $p(1-p)n^*$ that matters when comparing the Wilson-type and logistic-transformation intervals. This requires that P(1-P) not be too small.

Brown et al. (2001) show empirically that under simple random sampling (with n = 50), coverage intervals derived from the logistic transformation tend to be larger than corresponding Wilson intervals for small values of P(1-P). Kott and Liu (2009) make the same observation for one-sided intervals based on complex samples, supporting the notion that it is a better choice.

The asymptotic equivalence of the logistic-transformation interval with the Wilson interval explains the former's empirical superiority in the literature (e.g., in Brown et al., 2001) to an analogous interval constructed using an arcsine transformation. Because $\arcsin(p)$ has a constant large-sample variance under simple random sampling no matter the true value of P (so long as P(1-P) > 0), it has been hoped that the arcsine transformation would be ideal for interval construction.

Better than a Wilson interval, but not yet incorporated into any software package I know of, is the onesided coverage intervals for P derived using an Edgeworth expansion on p - P in Kott and Liu (2009). That method produces this two-sided interval:

$$p + \frac{1-2p}{\tilde{n}} \left(\frac{1}{6} + \frac{z_{1-\alpha/2}^2}{3} \right) - z_{1-\alpha/2} \left(\operatorname{var}(p) + \left[\frac{1-2p}{\tilde{n}} \left(\frac{1}{6} + \frac{z_{1-\alpha/2}^2}{3} \right) \right]^2 \right)^{1/2} \le P$$
$$\le p + \frac{1-2p}{\tilde{n}} \left(\frac{1}{6} + \frac{z_{1-\alpha/2}^2}{3} \right) + z_{1-\alpha/2} \left(\operatorname{var}(p) + \left[\frac{1-2p}{\tilde{n}} \left(\frac{1}{6} + \frac{z_{1-\alpha/2}^2}{3} \right) \right]^2 \right)^{1/2},$$

. . .

where $\tilde{n} = [(1-2p) \operatorname{var}(p)] / \operatorname{cov}[\operatorname{var}(p), p]$, and $\operatorname{cov}[\operatorname{var}(p), p]$, a consistent estimator for $\operatorname{Cov}[\operatorname{var}(p), p]$, exists and equals a consistent estimator for the third moment of p. Note that $\operatorname{cov}[\operatorname{var}(p), p]$ doesn't exist for designs with only two primary sampling units per stratum. Moreover, it is not a consistent estimator for the third moment of p when finite population correction matters.

Observe that \tilde{n} again replaces n^* . In addition, $1/6 + z_{1-\alpha/2}/3$ replaces $z_{1-\alpha/2}/2$, which means that the center will often be closer to the p using this interval rather than the Wilson. The good coverage properties of this interval, like the Wilson, breaks down when the skewness coefficient of $p\left(E\left[(p-P)^3\right]/[Var(p)]^{3/2}\right)$ gets too large in absolute value, how large has yet to be determined.

Finally, SAS/STAT (SAS Institute Inc., 2010) offers a Wilson coverage interval for estimated proportions in its SURVEYFREQ procedure. The procedure's method of adjusting the effective sample size, which can – and should – be turned off, is not related to the \tilde{n} discussed here. Instead, it is based on an ad-hoc t – adjustment that sadly is not related to the variance of the denominator variance of the Wilson pivotal.

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