

# Required sample sizes for estimating means and areal fractions of soil fertility parameters for districts in Andhra Pradesh

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## 1 Introduction

The Soil Health Card (SHC) project in India involves soil sampling at a very high density every two years. For example, Andhra Pradesh state ( $162\,975\text{ km}^2$ ) in Cycle 2 (2017/18 - 2018/19) recorded  $2\,393\,8875^{[1]}$  observations, a density of  $14.7\text{ km}^{-1}$ , i.e., one per 6.8 ha. These data are used for soil fertilization recommendations at the field level. Due to the very high sampling density the current soil sampling survey is expensive: labour costs for collecting the soil samples and analyzing these soil samples in laboratories are high. The question is whether this high investment in soil survey pays. Would a reduction of the number of sampling locations also suffice? This document describes statistical methods for adapting the existing sampling campaign.

In this document we focus on estimation of means and areal fractions of soil fertility variables within an administrative unit, a district in the state Andhra Pradesh. For this aim a *design-based* sampling approach is recommended (Brus and de Gruijter, 1997). In this approach sampling locations are selected by *probability sampling*. Probability sampling involves the use of a random number generator. Probability samples can be selected in many ways. The simplest selection method, also called a *sampling design*, is *simple random sampling* (SRS). In SRS each possible sample of  $n$  sampling locations ( $n$  is referred to as the *size* of the sample) has equal probability of being selected.

As an illustration we compute the required sample size for estimating the means of Zn and areal fractions with Zn-deficiency within each district of Andhra Pradesh. These estimated population parameters are of practical importance. They can, for example, be used to prioritize districts for policy interventions. The quality criterion that is used to compute these sample sizes is the width of a confidence interval of the population parameter.

The required sample sizes are computed with three approaches: the frequentist approach, the fully Bayesian approach and the mixed Bayesian-likelihood approach (Joseph et al., 1995, Joseph and Belisle, 1997). To compute the required sample sizes for estimating the mean of Zn within districts, the Zn concentrations are log-transformed. The probability distribution of log-transformed Zn concentrations are much closer to a normal distribution, which is required for the Bayesian and mixed Bayesian-likelihood approach. As a critical

Zn-concentration we use 0.9, i.e., if the Zn-concentration at a location is less than 0.9, we consider that this location is deficient of Zn, so that the application of Zn fertilizer is recommended.

With the fully Bayesian and mixed Bayesian-likelihood approach required sample sizes are computed for three criteria: the average width criterion, the average coverage criterion and the worst outcome criterion (Joseph et al., 1995, Joseph and Belisle, 1997)..

## 2 Reading the data

The cycle 1 SHC data collected in 2015-2017 are used to compute summary statistics for each district.

##	district	n	mean	var	fraction
## 1	Anantapur	49114	-0.725297326	0.9556266	0.7679277
## 2	Chittoor	37978	-0.060070260	0.4066594	0.4946021
## 3	East Godavari	30353	0.239702652	0.6963413	0.3251408
## 4	Guntur	63956	-0.368286248	0.8161529	0.6118269
## 5	Kadapa	21739	-0.658311838	0.5992403	0.7723446
## 6	Krishna	30481	-0.050333924	0.7972063	0.3915554
## 7	Kurnool	79775	-0.388149609	1.1241939	0.5891319
## 8	Nellore	48053	-1.217143951	1.1920288	0.8576155
## 9	Prakasam	50392	-0.638136119	1.3453333	0.6716741
## 10	Srikakulam	40823	0.008696663	0.4371163	0.4009994
## 11	Visakhapatnam	8678	-0.405452646	0.9664817	0.5746716
## 12	Vizianagaram	28321	-0.350607401	0.4598608	0.6402316
## 13	West Godavari	20211	0.368697951	0.7655733	0.2586215

## 3 Required sample sizes for estimating the mean of $\ln(\text{Zn})$

Given a maximum width  $w_{\max}$  of a  $100(1 - \alpha)\%$  confidence interval of the population mean, in the frequentist approach the required sample size can be computed with

$$n = \left( u_{(1-\alpha/2)} \frac{\tilde{\sigma}}{w_{\max}/2} \right)^2$$

The sample variance as computed with the cycle 1 SHC data of 2015-2017 (see output above) is used for  $\tilde{\sigma}^2$ .

As we are uncertain about the population standard deviation  $\sigma$ , in the fully Bayesian and mixed Bayesian-likelihood approach a prior distribution is assigned to this parameter. It is convenient to assign a gamma distribution as a prior distribution to the reciprocal of the population variance, referred to as the precision parameter  $\lambda = 1/\sigma^2$ . More precisely, a prior *bivariate* normal-gamma distribution is assigned to the population mean and the precision parameter. With this prior distribution, the *posterior* distribution of the population mean is fully defined, i.e. both the type of distribution and its parameters are known. We say that the prior distribution is *conjugate* with the normal distribution.

A prior gamma distribution is assigned to the reciprocal of the population variance  $\lambda = 1/\sigma^2$ . This gamma distribution has two parameters  $a$  and  $b$ . The mean of a gamma distribution equals  $a/b$ , the standard deviation equals  $\sqrt{a/b^2}$ . The mean of the gamma distribution was set equal to the reciprocal of the legacy sample variance of  $\ln(\text{Zn})$ ,  $a/b = 1/\sigma^2$  (Table ??). A second equation with  $a$  and  $b$  is needed to derive parameters  $a$  and  $b$ . In this second equation the coefficient of variation of the gamma distribution,  $cv(\lambda)$ , is set equal to a user-specified value. Solving the two equations with two unknowns gives  $a = 1/\{cv(\lambda)\}^2$  and  $b = a \sigma^2$ . Required sample sizes are computed for a coefficient of variation of 0.25 of the gamma distributions for the precision parameter.

```
library(SampleSizeMeans)
wmax <- 0.2 #maximum width (=length) of confidence interval
```

```

conflevel <- 0.95
worstlevel <- 0.80
lambda <- 1/sigma2 #prior estimate of precision parameter
cv <- 0.25 #coefficient of variation of gamma distribution for lambda
nreq.freq <- nreq.alc.bayes <- nreq.alc.mbl <- nreq.acc.bayes <- nreq.acc.mbl <- nreq.woc.bayes <- nreq.woc.mbl
for (i in 1:length(districts)) {
  a <- 1/cv^2
  b <- a/lambda[i]
  nreq.freq[i] <- mu.freq(len=wmax, lambda=lambda[i], level=conflevel)
  nreq.alc.bayes[i] <- mu.alc(len=wmax, alpha=a, beta=b, n0=0, level=conflevel)
  nreq.alc.mbl[i] <- mu.mblalc(len=wmax, alpha=a, beta=b, level=conflevel)
  nreq.acc.bayes[i] <- mu.acc(len=wmax, alpha=a, beta=b, n0=0, level=conflevel)
  nreq.acc.mbl[i] <- mu.mblacc(len=wmax, alpha=a, beta=b, level=conflevel)
  nreq.woc.bayes[i] <- mu.modwoc(len=wmax, alpha=a, beta=b, n0=0, level=conflevel, worst.level=worstlevel)
  nreq.woc.mbl[i] <- mu.mblmodwoc(len=wmax, alpha=a, beta=b, level=conflevel, worst.level=worstlevel)
}
(df <- data.frame(district=districts, lambda = round(lambda,2),
  freq = nreq.freq,
  alc = nreq.alc.bayes,
  alc.mbl = nreq.alc.mbl,
  acc = nreq.acc.bayes,
  acc.mbl = nreq.acc.mbl,
  woc = nreq.woc.bayes,
  woc.mbl = nreq.woc.mbl))

```

##	district	lambda	freq	alc	alc.mbl	acc	acc.mbl	woc	woc.mbl
## 1	Anantapur	1.05	368	386	388	397	399	466	473
## 2	Chittoor	2.46	157	165	166	169	173	197	204
## 3	East Godavari	1.44	268	282	283	289	292	339	346
## 4	Guntur	1.23	314	330	331	339	342	398	404
## 5	Kadapa	1.67	231	243	244	249	252	292	298
## 6	Krishna	1.25	307	323	324	331	334	388	396
## 7	Kurnool	0.89	432	454	455	467	470	548	556
## 8	Nellore	0.84	458	482	483	495	497	581	588
## 9	Prakasam	0.74	517	544	545	559	561	656	662
## 10	Srikakulam	2.29	168	177	179	182	185	212	219
## 11	Visakhapatnam	1.03	372	391	392	402	405	471	479
## 12	Vizianagaram	2.17	177	187	188	191	194	223	230
## 13	West Godavari	1.31	295	310	311	318	320	373	379

## 4 Required sample sizes for estimating the areal fraction with Zn deficiency

Given a maximum width  $w_{\max}$  of a  $100(1 - \alpha)\%$  confidence interval of the areal fraction, in the frequentist approach the required sample size can be computed with

$$n = \left( u_{(1-\alpha/2)} \frac{\sqrt{\hat{\pi}(1-\hat{\pi})}}{w_{\max}/2} \right)^2 + 1$$

In the fully Bayesian and mixed Bayesian-likelihood approach a prior beta distribution is assigned to the design-parameter  $\hat{\pi}$ . The beta distribution has two parameters  $\alpha$  and  $\beta$  which correspond to the number of “successes” (1) and “failures” (0) in the problem context. The larger these numbers, the more the prior

information, and the more sharply defined the probability distribution. By setting the mode of the prior beta distribution to the prior estimate of the areal fraction  $\tilde{\pi}$ , the parameters of the beta distribution can be computed by

$$\alpha = n_0 \tilde{\pi} + 1$$

$$\beta = n_0(1 - \tilde{\pi}) + 1 ,$$

with  $n_0$  the prior sample size.

## Loading required package: binom

##	district	f	wald	alc	alc.mbl	acc	acc.mbl	woc	woc.mbl
## 1	Anantapur	0.768	275	223	271	226	276	263	318
## 2	Chittoor	0.495	386	335	371	335	371	343	381
## 3	East Godavari	0.325	339	299	327	301	330	336	368
## 4	Guntur	0.612	366	294	356	295	357	311	376
## 5	Kadapa	0.772	272	249	269	256	278	310	335
## 6	Krishna	0.392	368	324	353	325	354	348	378
## 7	Kurnool	0.589	373	286	364	286	364	298	379
## 8	Nellore	0.858	189	144	192	149	201	187	247
## 9	Prakasam	0.672	340	282	331	284	333	310	364
## 10	Srikakulam	0.401	371	319	358	319	359	337	379
## 11	Visakhapatnam	0.575	377	333	340	337	345	370	378
## 12	Vizianagaram	0.640	355	315	341	317	344	347	375
## 13	West Godavari	0.259	296	271	289	277	297	328	352

## 5 References

Joseph, L., Wolfson, D.B. and Du Berger, R. 1995. Sample size calculations for binomial proportions via highest posterior density intervals. *Journal of the Royal Statistical Society. Series D (The Statistician)*: 44, 143-154.

Joseph, L. and Belisle, P. 1997. Bayesian sample size determination for normal means and differences between normal means. *Journal of the Royal Statistical Society. Series D (The Statistician)*: 46, 209-226.