

## A Laplace Approximation for Sums of Independent Random Variables

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### Introduction

In the well known Laplace approximation one considers integrals of the form  $\int \varphi(x) e^{nf(x)} dx$ , where  $\varphi$  and  $f$  are sufficiently nice real valued functions of  $x \in \mathbb{R}^1$ , and  $n$  is a large positive number. It is based on the following argument: Suppose that  $f(x)$  attains its maximum at  $x=0$  e.g., and that  $f(x) - f(0) \sim -A|x|^a$  for some  $A, a > 0$  when  $|x|$  is small. Then  $e^{nf(x)}$  will have a very sharp maximum at  $x=0$ , so the main contribution to the integral comes from a neighbourhood of 0, and we get:

$$\begin{aligned} \int \varphi(x) e^{nf(x)} dx &\sim \int_{|x| \leq n^{-1/2a}} \varphi(x) e^{nf(x)} dx \sim \varphi(0) \int_{|x| \leq n^{-1/2a}} e^{nf(0) - nA|x|^a} dx \\ &= \varphi(0) e^{nf(0)} (nA)^{-1/a} \int_{|y| \leq n^{1/2a} A^{1/a}} e^{-|y|^a} dy \\ &\sim \varphi(0) e^{nf(0)} (nA)^{-1/a} \int e^{-|y|^a} dy = \varphi(0) e^{nf(0)} (nA)^{-1/a} 2\Gamma(1+1/a). \end{aligned}$$

The meaning of  $\sim$  is that the ratio of the two expressions goes to 1 as  $n$  goes to  $\infty$ .

A complete proof is straightforward and is given e.g. in [4].

In some recent work on the asymptotics of an epidemic process [3] we met with the problem of deriving an analogous approximation for an expression of the form

$$E_n = E(e^{nf(S_n/n)}) = \int e^{nf(x/n)} F^{n*}(dx), \quad (1)$$

where  $\{S_n\}$  are the partial sums of i.i.d random variables  $\{X_i\}$  with a common distribution  $F(x)$ ,  $f(x)$  is a sufficiently nice function of  $x \in \mathbb{R}^1$ , and  $n \rightarrow \infty$ . In order to understand how the analogous approximation of  $E_n$  is formed let us first consider the situation when  $X_1$  has a density. Then under appropriate conditions the central limit theorem for large deviations tells us that  $\varphi_n(x)$ , the density of  $S_n/n$ , is approximatively given by the following formula:

$$\varphi_n(x) \sim (n/2\pi)^{1/2} (-h''(x))^{1/2} e^{nh(x)}. \quad (2)$$

Here  $h(x)$  is the entropy function of the distribution  $F(x)$ , which is defined below.

A proof of the above formula is given e.g. in [9]. Using it in the expression for  $E_n$  we can then make the Laplace approximation of the resulting integral as follows:

Suppose that  $f(x) + h(x)$  has its maximum at  $x=0$  e.g. and that

$$f(x) + h(x) - f(0) - h(0) \sim -A|x|^a \quad (3)$$

with  $A, a > 0$  as  $|x| \rightarrow 0$ , then we get

$$\begin{aligned} E_n &\sim (n/2\pi)^{1/2} \int (-h''(x))^{1/2} e^{n(f(x)+h(x))} dx \\ &\sim (2\pi)^{-1/2} 2\Gamma(1+1/a) A^{-1/a} (-h''(0))^{1/2} n^{1/2-1/a} e^{n(f(0)+h(0))}. \end{aligned} \quad (4)$$

When  $f(x)$  is twice continuously differentiable near 0 then  $2A = -f''(0) - h''(0)$ , and  $a=2$  if  $A > 0$ , so in this case the formula reduces to

$$E_n \sim (1 + f''(0)/h''(0))^{-1/2} e^{n(f(0)+h(0))}. \quad (5)$$

In the following we give a complete proof of the above asymptotic formulas with rather weak assumptions about  $F(x)$  and  $f(x)$ . (Theorem 2 and 3 below.)

Concerning  $f(x)$  it is enough to assume that it is Hölder continuous near the maximum at  $x=0$  and that it can be approximated by piecewise linear functions as explained in (26).

Concerning  $F(x)$  we have to assume that its generating function

$$e^{g(t)} = \int e^{tx} F(dx) = E(e^{tX_1}) \quad (6)$$

is defined in some open interval  $D_g \subset R^1$ , so that laws of large deviations can be established using saddlepoint methods. Then the entropy function  $h(x)$ , defined by  $h(x) = \inf_t (g(t) - tx)$ , is finite and very regular in some open interval  $D_h$ . It is assumed that  $f(x) + h(x)$  attains its maximum at a unique point  $\bar{x}$  in  $D_h$ , which we can take to be  $\bar{x} = 0$ .

It turns out that it is not necessary to assume that  $F(x)$  has a density. Either  $F$  is arithmetic, i.e.  $X_1$  takes only values  $\xi + m \cdot \delta$ , where  $m$  is integer and  $\delta > 0$ , and  $\text{Re } g(it)$  is periodic with period  $2\pi/\delta$  or if it is not, then it is enough to assume that the so called Cramér condition

$$\sup_{|t| \geq u} \text{Re } g(it) < 0 \quad \text{for any } u > 0 \quad (7)$$

is fulfilled.

Under these assumptions we can establish the formula (4) for any  $A, a > 0$ . (In the arithmetic case the formula has to be modified when  $0 < a \leq 1$ .) (Theorem 2 below.)

When  $a=2$ ,  $A > 0$  it turns out that (5) is true even without assuming the Cramér condition (7), and the arithmetic case need not be considered separately. (Theorem 3 below.)

We have treated explicitly only the case when  $f(x) + h(x)$  attains its maximal value at only one point. If it attains it at a finite number of points (4) is

of course replaced by a sum of similar terms, one for each maximum, and only those with the largest value of  $a$  will contribute as  $n \rightarrow \infty$ .

The formula (4) is an example of the use of “thermodynamic” probability estimates familiar in statistical mechanics. I.e. one considers a “macroscopic variable”  $S_n/n$  whose probability distribution is asymptotically determined by an entropy function through a formula like (2), and in the resulting formula (4) all information about the underlying distribution  $F$  is contained in  $h$  and its derivatives at the maximal point  $\bar{x}$ . The formula (4) ought to be useful in other statistical and physical applications as well.

(In [4] similar formulas are derived in a special situation when  $f(x) = \alpha x^2$ .)

The fact that  $S_n/n$  behaves as if it had a density  $\varphi_n(x)$  proportional to  $e^{nh(x)}$  so that  $\varphi_n(x) e^{nf(x)}$  is proportional to  $e^{-nA|x-\bar{x}|^a}$  near  $\bar{x}$  can be expressed more precisely in the form of a central limit theorem as follows: Let a modified probability distribution for  $S_n$  be defined by  $F_{n,f}(dx) = e^{nf(x/n)} F^{n*}(dx)/E_n$ . Then for this distribution the following central limit theorem holds: (Theorem 4 below). The distribution of  $n^{1/a}(S_n/n - \bar{x})$  defined by  $F_{n,f}$  converges weakly to the one defined by the density

$$e^{-A|x|^a} A^{1/a} dx / 2\Gamma(1 + 1/a) \quad \text{as } n \rightarrow \infty \quad \text{if } a > 1.$$

In particular when  $a=2$ ,  $A > 0$  this limit distribution is Gaussian with variance  $1/2A = -1/(f''(0) + h''(0))$ .

The problem of deriving logarithmic estimates for the distribution of  $S_n/n$  saying that  $\lim_{n \rightarrow \infty} n^{-1} \log P(S_n/n \in I) = \sup_{x \in I} h(x)$  has recently received much attention following the work of Lanford [7], where this formula is established for intervals  $I$  using an elegant subadditivity argument. A survey of such results and their statistical background is given in the recent article [2] by Bahadur and Zabell. In a much more general context Varadhan [11] has shown how to derive the corresponding logarithmic estimate

$$\lim_{n \rightarrow \infty} n^{-1} \log E_n = \sup_x (f(x) + h(x))$$

using the above logarithmic estimates for the distribution of  $S_n/n$  and arguments valid for distributions on more general metric spaces than  $R^1$ .

Our more accurate formula for  $E_n$ , valid in the sense of  $\sim$ , gives so to speak the accuracy of the central limit theorem, and it depends on the local behaviour of  $f(x) + h(x)$  near the maximal point  $\bar{x}$ . Analogous formulas for  $P(S_n/n \in I)$  have recently been derived by Höglund [6].

### 1. Preparations

Let us first recall the basic properties of  $g(t)$  defined by (6) and its Legendre transform or entropy function defined by

$$h(x) = \inf_t (g(t) - tx), \tag{8}$$

which we will need to know in the following (c.f. [8, 10]).

Let  $D_g$  be the maximal open interval in which  $g(t) < \infty$ , and let  $g(t) = +\infty$  outside of  $\bar{D}_g$ .  $D_g$  is assumed to be non empty. Then  $g(u+iv)$  is analytic in the vertical strip defined by  $u \in D_g$  and  $g(t)$  is continuous when  $t$  approaches the endpoints of  $D_g$ . In particular  $g(t)$  is infinitely differentiable when  $t \in D_g$ . Since  $g''(t) > 0$   $g(t)$  is strictly convex, and the mapping  $t \rightarrow x$  defined by  $x = g'(t)$  is strictly increasing and infinitely differentiable. Hence it is 1-1 from  $D_g$  onto an open interval  $D_h$ , and the inverse mapping is also strictly increasing and infinitely differentiable. When  $x \in D_h$  and  $x = g'(t)$  then inf in (8) is attained at  $t$ , so

$$h(x) = g(t) - tx, \quad \text{when } x = g'(t). \tag{9}$$

From (8) follows that  $h(x)$  is concave and closed (i.e. upper semicontinuous) and that  $g(t)$  is determined from  $h(x)$  by

$$g(t) = \sup_x (h(x) + tx). \tag{10}$$

From (9) it follows that  $h(x)$  is infinitely differentiable and strictly concave in  $D_h$  because we have

$$h'(x) = -t, \tag{11}$$

so the inverse mapping of  $x = g'(t)$  is given by  $t = -h'(x)$ . Therefore

$$\frac{dx}{dt} = g''(t) = -1/h''(x), \tag{12}$$

which shows that  $h''(x) < 0$ , and  $h(x)$  is strictly concave in  $D_h$ . Outside of  $D_h$  it is linear or  $-\infty$ .

We first derive a useful upper bound for  $E_n$  defined in (1) in the special case when  $f(x)$  is linear in an interval and  $-\infty$  outside it:

**Lemma 1.** *For any interval  $I$*

$$B_n = E(e^{tS_n}, S_n/n \in I) \leq \exp n \sup_{x \in I} (h(x) + tx). \tag{13}$$

*Proof.* Take first  $I$  to be open. If  $\sup_{x \in I} (h(x) + tx) = \sup_x (h(x) + tx) = g(t)$  then  $B_n \leq E(e^{tS_n}) = e^{ng(t)}$ , and (13) is true.

If  $-\infty < \sup_{x \in I} < \sup_x$  then e.g. the left endpoint  $l$  of  $I$  is finite,  $h(x)$  is finite to the right of  $l$ ,  $\sup_{x \in I} = h(l) + tl$ , and the right derivative  $h'_+(l) + t \leq 0$ . (A convex function is continuous and has one sided derivatives from the side where it is finite if it is closed as  $h$  is.)  $t_+ = -h'_+(l) > -\infty$  defines a supporting line to  $h(x)$  at  $l$ , i.e.  $h(x) \leq h(l) - t_+(x-l)$  for all  $x$ . This means that  $g(t_+) = \sup_x (h(x) + t_+ x) = h(l) + t_+ l$ , and since  $t(x-l) \leq t_+(x-l)$  for  $x \in I$  we have

$$B_n \leq E(e^{nt + (n^{-1}S_n - l) + nt}) = e^{ng(t_+) - nt + l + nt} = e^{n(h(l) + tl)},$$

and (13) is valid.

Finally, if  $\sup_{x \in I} (h(x) + tx) = -\infty$  then  $h(x) = -\infty$  for all  $x \in I$ , i.e.  $\inf (g(t) - tx) = -\infty$  for all  $x \in I$ . This means that  $I \cap D_h = \emptyset$ , so  $I$  lies e.g. to the right of  $D_h$ . I.e.  $g'(t) < x$  for any  $t \in D_g$  and any  $x \in I$ , so  $g(t) - tx$  is decreasing and

$$\inf_t (g(t) - tx) = \lim_{t \rightarrow +\infty} (g(t) - tx) = -\infty \quad \text{for any } x \in I.$$

Hence  $\lim_{t \rightarrow +\infty} E(e^{t(S_n - nx)}) = 0$  for any  $x \in I$ , so that  $P(S_n/n \geq x) = 0$ . If we now let  $x$  decrease to  $l$  we can conclude that  $P(n^{-1}S_n > l) = 0$ . Hence  $B_n = 0$  and (13) is again valid.

If  $I$  is not open and e.g.  $I = [l, r)$  then we consider the slightly larger interval  $I' = (l - \varepsilon, r)$  and let  $\varepsilon \rightarrow 0$ . Because  $h$  is closed we have that  $\limsup_{\varepsilon \rightarrow 0} \sup_{|x-l| \leq \varepsilon} h(x) \leq h(l)$ . This means that

$$\limsup_{\varepsilon \rightarrow 0} \sup_{x \in I'} (h(x) + tx) = \sup_{x \in I} (h(x) + tx),$$

and (13) follows by letting  $\varepsilon \rightarrow 0$  in (13'):

$$B_n \leq B'_n \leq \exp n \sup_{x \in I'} (h(x) + tx).$$

We next proceed to prove a local limit theorem for large deviations saying that if the Cramér condition (7) is fulfilled, then (2) is true for suitably defined smoothed density, which is defined as follows: Let  $k(x)$  be the usual Fejer kernel  $k(x) = \max(1 - |x|, 0)$  with Fourier transform  $\hat{k}(t) = 2(1 - \cos t)/t^2$ , and let for  $c > 0$  small  $k_c(x) = c^{-1}k(x/c)$  be the corresponding approximate delta function with  $\hat{k}_c(t) = \hat{k}(ct)$ . Since  $k(x)$  has compact support  $\hat{k}(t)$  is an entire function. Define the smoothed density of  $S_n$  as follows:

$$\varphi_{n,c}(x) = E(k_c(S_n - nx)) = \int k_c(y - nx) F^{n*}(dy). \tag{14}$$

Then the following asymptotic formula holds when  $n \rightarrow \infty, c \rightarrow 0$ :

**Theorem 1.** *If the Cramér condition (7) is fulfilled, then*

$$\varphi_{n,c}(x) = (2\pi n)^{-1/2} (-h''(x))^{1/2} e^{nh(x)} (1 + \mathcal{O}(n^{-1/2} + c + e^{-n\gamma}/c)) \tag{15}$$

*uniformly when  $x$  varies on a compact subset of  $D_h$  for some  $\gamma > 0$ . In the arithmetic case the same asymptotic formula is valid for  $P(S_n = nx)/\delta$  uniformly when  $x = \xi + m\delta/n$  varies on compact subsets of  $D_h$  and  $m$  is an integer. (Excluding of course the error terms involving  $c$ .)*

*Proof.* We use Parsevals relation and the usual change of the path of integration used in the saddlepoint method to express  $\varphi_{n,c}(x)$  as follows:

$$\varphi_{n,c}(x) = \int e^{n(g(iu) - iux)} \hat{k}_c(u) du / 2\pi.$$

If  $x = g'(t)$  with  $t \in D_g$  change the path of integration so that  $iu \rightarrow t + iu$ . (Here we use the fact that  $g(t + iu)$  is analytic when  $t \in D_g$  and that  $\hat{k}_c(u)$  is entire.)

Then  $g(t) - tx = h(x)$  by (9) so we get:

$$\begin{aligned}\varphi_{n,c}(x) &= \int e^{n(g(t+iu) - (t+iu)x)} \widehat{k}_c(u-it) du/2\pi \\ &= e^{nh(x)} \int e^{n(g(t+iu) - g(t) - iug'(t))} \widehat{k}_c(u-it) du/2\pi.\end{aligned}\quad (16)$$

We now split the integral into three parts  $I_1 + I_2 + I_3$  corresponding to the regions  $|u| \leq d_1/n^{1/2}$ ,  $d_1/n^{1/2} < |u| \leq d_2$ ,  $d_2 < |u|$  respectively, where  $d_1$  is large,  $d_1/n^{1/2}$  is small and  $d_2 > 0$  as  $n \rightarrow \infty$ . Each part is estimated separately:

$I_1$ : When  $|u|$  is small

$$g(t+iu) - g(t) - iug'(t) = -u^2 v/2 + \mathcal{O}|u|^3, \quad (17)$$

where  $v = g''(t) = -1/h''(x) > 0$ .

This holds uniformly when  $t \in K_g = \text{compact} \subset D_g$  and  $|u| \leq d_1/n^{1/2}$ . Replacing  $u$  by  $u/n^{1/2}$  in  $I_1$  we get

$$I_1 = \int_{|u| \leq d_1} e^{-vu^2/2} (1 + \mathcal{O}(|u|^3/n^{1/2})) \widehat{k}_c(u/n^{1/2} - it) du/n^{1/2} 2\pi.$$

Let us put

$$\begin{aligned}I_0 &= \int e^{-vu^2/2} \widehat{k}_c(u/n^{1/2} - it) du/n^{1/2} 2\pi \\ &= \int e^{-nvvu^2/2} \widehat{k}_c(u-it) du/2\pi \\ &= \int (2\pi nv)^{-1/2} e^{-x^2/2nv} e^{tx} k_c(x) dx.\end{aligned}$$

The last equation follows from Parseval's relation. We see that  $I_0 = (2\pi nv)^{-1/2} (1 + \mathcal{O}(c))$  as  $c \rightarrow 0$  uniformly when  $n \rightarrow \infty$  and  $t \in K_g$ . It is easy to check that

$$|\widehat{k}(u-it)| \leq K/(1+u^2) \quad (18)$$

for some  $K > 0$  uniformly when  $t \in K_g$ .

Hence

$$\begin{aligned}|I_1 - I_0| &= \mathcal{O}\left(\int_{|u| \leq d_1} |u|^3 e^{-vu^2/2} du/n + \int_{|u| > d_1} e^{-vu^2/2} du/n^{1/2}\right) \\ &= \mathcal{O}((nv)^{-1/2} (n^{-1/2} + e^{-vd_1^2/2})),\end{aligned}$$

so we have

$$I_1 = (2\pi nv)^{-1/2} (1 + \mathcal{O}(c + n^{-1/2} + e^{-vd_1^2/2})). \quad (19)$$

$I_2$ : From (17) we see that in  $I_2$  we have  $g(t+iu) - g(t) - iug'(t) \geq -u^2 v/4$  when  $|u| \leq d_2$  is small enough. Hence using (18) we have

$$I_2 = \mathcal{O}\left(\int_{|u| \geq d_1} e^{-vu^2/4} du/n^{1/2}\right) = \mathcal{O}((nv)^{-1/2} e^{-vd_1^2/4}). \quad (20)$$

$I_3$ : To estimate  $I_3$  we need the Cramér condition (7) and (18). If we can show that  $\text{Re } g(t+iu) - g(t) \leq \gamma < 0$  when  $|u| \geq d_2$  and  $t \in K_g$  then we have

$$I_3 = \mathcal{O}(e^{-n\gamma} \int du/(1+(cu)^2)) = \mathcal{O}(e^{-n\gamma}/c). \quad (21)$$

To see that the above bound for  $\text{Re } g$  holds we use the fact that

$$e^{g(t+iu)-g(t)} = \int e^{iux} e^{tx-g(t)} F(dx) = \int e^{iux} F_t(dx) \tag{22}$$

where  $F_t(dx)$  is absolutely continuous with respect to  $F(dx)$ . In [1], Lemma 4 it is shown that the Cramér condition is preserved when the distribution  $F$  is changed to one which is absolutely continuous with respect to  $F$ . Hence  $\sup_{|u| \geq d_2} \text{Re } g(t+iu)-g(t) < 0$  for each  $t$ . But it is easy to check that  $g(t+iu)-g(t)$  is continuous in  $t$  uniformly in  $u$ , so that  $\sup_{t \in K_g} \sup_{|u| \geq d_2} \text{Re } g(t+iu)-g(t) \leq \gamma < 0$  also.

If we now put  $v^{1/2} d_1 = n^{1/4}$  e.g. and add the expressions in (19), (20) and (21) we get the desired formula (15) because  $v = -1/h''(x) > 0$ . The proof in the arithmetic case follows the same lines starting from the inversion formula for the individual atoms of the distribution of  $S_n$ :

$$\begin{aligned} P(S_n = nx) &= \delta \int_{|u| \leq \pi/\delta} e^{n(g(iu)-ixu)} du / 2\pi \\ &= \delta e^{nh(x)} \int_{|u| \leq \pi/\delta} e^{n(g(t+iu)-g(t)-iug'(t))} du / 2\pi \end{aligned} \tag{23}$$

where  $x = \xi + m\delta/n$ ,  $m$  integer and  $g'(t) = x$ . In this case we take  $d_2 = \pi/\delta$  and the term  $I_3$  is absent.

### 2. First Proof of the Laplace Approximation

We now show that Theorem 1 can be used to establish the asymptotic formula (4) in the way indicated in the introduction. We assume that  $f(x) + h(x)$  attains its maximum at only one point,  $x=0$  e.g., and that  $0 \in D_h$ . Near this point we assume (3) and that  $f(x)$  is Hölder continuous:

$$|f(x+y) - f(x)| \leq A'|y|^{a'} \quad \text{for some } A', a' > 0. \tag{24}$$

Consider now  $E_n$  defined by (1) and split it into the contribution from a neighbourhood of 0 and the rest:

$$\begin{aligned} E_{n,1} &= E(e^{nf(S_n/n)} I_{2d}(S_n/n)) \\ E_{n,2} &= E_n - E_{n,1}, \end{aligned} \tag{25}$$

where  $I_d(x)$  is the indicator of the interval  $[-d, d]$ . Let us estimate  $E_{n,2}$  first with the help of Lemma 1. To this end we make the following regularity assumption:

For any  $d > 0$  there is a piecewise linear function  $f_d(x)$  with at most  $d^{-M}$  pieces for some  $M > 0$  such that

$$e^{f_d(x)-d} \leq e^{f(x)} \leq e^{f_d(x)+d} \quad \text{when } h(x) > -\infty. \tag{26}$$

$f(x)$  and  $f_d(x)$  are allowed to take the value  $-\infty$ . (This is true e.g. if  $f(x)$  is piecewise continuous and asymptotically linear at infinity.)

Take now  $b > 0$  such that  $f(x) + h(x) \leq f(0) + h(0) - 3b$  when  $|x| > d$  (this means that  $b = \mathcal{O}(d^a)$  when  $d$  is small), and take  $f_b(x)$  as in (26). In each interval  $J$  where  $f_b$  is linear we can use Lemma 1 to get the following bound:

$$\begin{aligned} E(e^{nf(S_n/n)}, S_n/n \in J) &\leq E(e^{nf_b(S_n/n) + nb}, S_n/n \in J) \\ &\leq \exp n \sup_{x \in J} (f_b(x) + h(x) + b) \\ &\leq \exp n \sup_{x \in J} (f(x) + h(x) + 2b) \\ &\leq \exp n (f(0) + h(0) - b) \end{aligned}$$

if  $|x| \geq 2d$  in  $J$ . We hence get the following bound for  $E_{n,2}$ :

$$E_{n,2} \leq e^{n(f(0) + h(0))} \cdot b^{-M} \cdot e^{-nb}. \quad (27)$$

(Intervals where  $h(x) = -\infty$  do not contribute to  $E_{n,2}$ .)

Coming now to  $E_{n,1}$  we consider the non arithmetic case first and want to use Theorem 1. Therefore we approximate the integrand by convolving it with  $k_{c/n}$ :

$$E_{n,3} = E(k_{c/n} * (e^{nf} I_{2d})(S_n/n)) \quad (28)$$

$$E_{n,4} = E_{n,1} - E_{n,3},$$

$$\begin{aligned} E_{n,3} &= \iint (n/c) k((n/c)(x/n - y)) e^{nf(y)} I_{2d}(y) dy F^{n*}(dx) \\ &= n \int e^{nf(y)} I_{2d}(y) \varphi_{n,c}(y) dx \end{aligned} \quad (29)$$

When we approximate  $\varphi_{n,c}(y)$  by using Theorem 1 the main term is:

$$\begin{aligned} E_{n,5} &= (n/2\pi)^{1/2} \int_{|x| \leq 2d} (-h''(x))^{1/2} e^{n(f(x) + h(x))} dx \\ &= (n/2\pi)^{1/2} (-h''(0))^{1/2} e^{n(f(0) + h(0))} \cdot \int_{|x| \leq 2d} (1 + \mathcal{O}(d)) e^{-n|x|^a(A + o(1))} dx. \end{aligned}$$

We have  $\int e^{-(nB)|x|^a} dx = (nB)^{-1/a} 2\Gamma(1 + 1/a)$ , and for the tail of this integral the following simple estimate holds:

**Lemma 2.**

$$c^{1/a} \int_d^\infty e^{-cx^a} dx \leq e^{-cd^a} (2/a) (cd^a)^{(1/a) - 1} \quad (30)$$

when  $cd^a \geq 2/a$ .

*Proof.*

$$\begin{aligned} \int_d^\infty e^{-cx^a} dx &= \int_d^\infty x e^{-cx^a} + \int_d^\infty acx^a e^{-cx^a} dx \\ &\geq -de^{-cd^a} + acd^a \int_d^\infty e^{-cx^a} dx. \end{aligned}$$

Hence  $(acd^a - 1) \int_d^\infty e^{-cx^a} dx \leq de^{-cd^a}$ , and (30) follows because  $acd^a - 1 \geq acd^a/2$ .



Using this estimate we see that the integral in  $E_{n,5}$  can be bounded above and below by

$$(nB)^{-1/a} 2\Gamma(1+1/a) (1 + \mathcal{O}(d) + \mathcal{O}(e^{-nB(2d)^a} (nB(2d)^a)^{(1/a)-1})),$$

where  $B$  can be made arbitrarily close to  $A$  by choosing  $d$  small. Hence, if  $d \rightarrow 0$ , but  $nd^a \rightarrow \infty$  we have

$$E_{n,5} = (2\pi)^{-1/2} 2\Gamma(1+1/a) A^{-1/2} (-h''(0))^{1/2} n^{1/2-1/a} e^{n(f(0)+h(0))} (1+o(1)) \quad (31)$$

and the same holds for  $E_{n,3}$  if the error terms in Theorem 1 go to zero, i.e. if  $c \rightarrow 0$  and  $e^{-\gamma n}/c \rightarrow 0$ .

It now remains to bound the term  $E_{n,4}$  in (28), and we split it into two pieces as follows:

$$\begin{aligned} E_{n,4} &= E(I_d(e^{nf} I_{2d} - k_{c/n} * (e^{nf} I_{2d})) (S_n/n)) \\ &\quad + E((1-I_d)(e^{nf} I_{2d} - k_{c/n} * (e^{nf} I_{2d})) (S_n/n)) = E_{n,6} + E_{n,7}. \end{aligned}$$

In  $E_{n,6}$  the argument of the integrand lies in the interval  $|x| \leq d$ . If also  $c/n < d$  then in the integral defining  $k_{c/n} * (e^{nf} I_{2d})$  the argument of the last factor lies in the interval  $|y| \leq 2d$ , so the integrand in  $E_{n,6}$  is given by:

$$\begin{aligned} (e^{nf} - k_{c/n} * e^{nf})(x) &= \int (e^{nf(x)} - e^{nf(x-y)}) k_{c/n}(y) dy \\ &= e^{nf(x)} \int (1 - e^{n(f(x-y)-f(x))}) k_{c/n}(y) dy, \end{aligned}$$

so using (24) it is bounded by:

$$|(e^{nf} - k_{c/n} * e^{nf})(x)| \leq e^{nf(x)} 2A' n(c/n)^{a'} \quad (32)$$

when  $|x| \leq d$  and  $d$  is small.

Hence we have:

$$|E_{n,6}| \leq 2A' n(c/n)^{a'} E_n \quad (33)$$

when  $d$  is small and  $c/n < d$ .

For  $E_{n,7}$  we have the estimate

$$|E_{n,7}| \leq E((1-I_d) e^{nf} (S_n/n)) + E((1-I_d) k_{c/n} * (e^{nf} I_{2d}) (S_n/n)).$$

The first term is bounded just as  $E_{n,2}$  above by (27). In the second term  $k_{c/n} * (e^{nf} I_{2d})(x) = \int k_{c/n}(y) e^{nf(x-y)} I_{2d}(x-y) dy$  is zero if  $|x| > 2d + c/n$  because  $k_{c/n}$  has compact support. Hence in the second term it can be bounded using (24) and (32) by e.g.  $2e^{nf(x)}$  when  $d$  is small, so the second term can be bounded just as the first, and

$$|E_{n,7}| \leq 3b^{-M} e^{-nb} e^{n(f(0)+h(0))} \quad (34)$$

when  $d$  is small and  $c/n < d$ .

Collecting all our estimates (27), (31), (33) and (34) we get the desired formula for  $E_n$

$$E_n = (2\pi)^{-1/2} 2\Gamma(1+1/a) A^{-1/2} (-h''(0))^{1/2} n^{1/2-1/a} e^{n(f(0)+h(0))} (1+o(1)) \quad (35)$$

if  $c$  and  $d$  can be chosen suitably. We need to have  $d \rightarrow 0$ ,  $nd^a \rightarrow \infty$ ,  $c/n < d$ ,  $n(c/n)^{a'} \rightarrow 0$ ,  $e^{-\gamma n}/c \rightarrow 0$ ,  $n^{1/a-1/2} b^{-M} e^{-nb} \rightarrow 0$  with  $b = \mathcal{O}(d^a)$ . This can easily be achieved if e.g.  $d = n^{-1/2a}$  and  $c/n = n^{-l}$  with  $l$  large enough.

It finally remains to estimate  $E_{n,1}$  in the arithmetic case. Now we can use Theorem 1 to estimate the atoms of the distribution of  $S_n$  directly and get:

$$E_{n,1} = (2\pi n)^{-1/2} \delta \sum_{|x| \leq d} (-h''(x))^{1/2} e^{n(f(x)+h(x))} (1 + \mathcal{O}(n^{-1/2})), \quad (36)$$

where the summation is taken over  $x = \xi + m\delta/n$ ,  $m$  integer. In estimating the sum in (36) we use (3) and the fact that  $h''(x)$  is continuous at  $x=0$ . We have to consider the cases  $a > 1$ ,  $a = 1$ ,  $a < 1$  separately however.

When  $a > 1$  the sum can be approximated from above and below by

$$(2\pi n)^{1/2} (-h''(0))^{1/2} e^{n(f(0)+h(0))} \sum_{|x| \leq d} e^{-nB|x|^a} (\delta/n) (1 + \mathcal{O}(d))$$

with  $B$  arbitrarily close to  $A$  if  $d$  is small. The sum is a Riemann sum approximating the integral  $\int_{|x| \leq d} e^{-nB|x|^a} dx$ , so it follows that (35) holds as above if  $d$  is chosen as before.

When  $a \leq 1$  the sum is dominated by its largest terms as follows: Let  $x_0 \leq 0 < x_1$  be those  $x$ -values in the sum which are closest to  $x=0$ . For a tail of the sum we then have the estimate

$$\sum_{y \geq x} e^{-nB|y|^a} \leq e^{-nB|x|^a} + \int_x^\infty e^{-nB|y|^a} dy \leq 2e^{-nB|x|^a}$$

as  $n \rightarrow \infty$  using Lemma 2.

Hence  $\sum_{y > x_1} e^{-nB|y|^a} \leq 2e^{-nB|x_1 + \delta/n|^a}$  e.g., and since  $x_1 = \theta\delta/n$  with  $0 \leq \theta \leq 1$  this is negligible compared to  $e^{-nB|x_1|^a}$  because

$$n(|\theta\delta/n + \delta/n|^a - |\theta\delta/n|^a) = n^{1-a} \delta^a (|\theta+1|^a - |\theta|^a) > n^{1-a} \delta^a (2^a - 1).$$

We conclude that the two terms corresponding to  $x_0$  and  $x_1$  dominate in the sum, so we have:

$$E_n = (2\pi n)^{-1/2} (-h''(0))^{1/2} \delta (e^{n(f(x_0)+h(x_0))} + e^{n(f(x_1)+h(x_1))}) (1 + o(1)) \quad (37)$$

when  $0 < a < 1$ .

When  $a = 1$  all the terms contribute however:

$$f(x) + h(x) = f(x_1) + h(x_1) - A(x - x_1) + o(x_1 + x)$$

for  $x = x_1 + m\delta/n$  gives

$$\begin{aligned} \sum_{x \geq x_1} e^{n(f(x)+h(x))} &= e^{n(f(x_1)+h(x_1))} \sum_0^\infty e^{-Am\delta} (1 + o(1)) \\ &= e^{n(f(x_1)+h(x_1))} / (1 - e^{-A\delta}) (1 + o(1)), \end{aligned}$$

so in this case we see that

$$E_n = (2\pi n)^{-1/2} (-h''(0))^{1/2} (\delta/(1 - e^{-A\delta})) \cdot (e^{n(f(x_0)+h(x_0))} + e^{n(f(x_1)+h(x_1))}) (1 + o(1)) \tag{38}$$

and we have finally proved the following result:

**Theorem 2.** *If  $f(x) + h(x)$  has a single maximum at  $x=0 \in D_h$  and if (3) and (24) hold near  $x=0$  and the regularity condition (26) holds for  $f(x)$  and the Cramér condition (7) holds for the distribution  $F(x)$  then the asymptotic formula (35) holds for  $E_n$  defined by (1) in the non arithmetic case. In the arithmetic case (7) does not hold but (35) is true if  $a > 1$ , whereas (37) holds if  $a < 1$  and (38) holds if  $a = 1$ .*

### 3. Second Proof of the Laplace Approximation

In this section we show that the asymptotic formula (5) for  $E_n$  holds when  $f(x)$  is twice continuously differentiable near 0 and  $2A = -(f''(0) + h''(0)) > 0$ ,  $a = 2$  without assuming that the Cramér condition (7) holds.

From (3) it follows that

$$t \equiv f'(0) = -h'(0), \quad \text{so } g'(t) = 0, \tag{39}$$

and we have with  $B = f''(0)$

$$f(x) = f(0) + tx + Bx^2/2 + o(x^2). \tag{40}$$

Again we split  $E_n$  into  $E_{n,1} + E_{n,2}$  with

$$E_{n,1} = E(e^{nf(S_n/n)}, |S_n/n| \leq d) \tag{41}$$

and estimate  $E_{n,2}$  as before by (27) with  $b = O(d^2)$ . Because of (40) we approximate  $E_{n,1}$  by

$$E_{n,3} = E(e^{n(f(0) + t(S_n/n) + B(S_n/n)^2/2)}, |S_n/n| \leq d) \tag{42}$$

and have

$$|E_{n,1} - E_{n,3}| \leq |e^{nd^2r(d)} - 1| E_{n,3} \tag{43}$$

where  $r(d) \rightarrow 0$  when  $d \rightarrow 0$ .

In estimating  $E_{n,3}$  we have to consider the case  $B > 0$  and  $B \leq 0$  separately. When  $B \leq 0$  we use the fact that

$$e^{-z^2/2} = E(e^{izU}) \tag{44}$$

where  $U$  is a standard Gaussian random variable independent of  $\{X_i\}$ .

Hence

$$\begin{aligned}
 E_{n,3} &= e^{nf(0)} E(e^{t+icU} S_n, |S_n/n| \leq d) \\
 &= e^{nf(0)} (E(e^{t+icU} S_n) - E(e^{t+icU} S_n, |S_n/n| > d)) \\
 &= E_{n,4} - E_{n,5}
 \end{aligned} \tag{45}$$

with  $c \equiv (|B|/n)^{1/2} \rightarrow 0$  as  $n \rightarrow \infty$ .

Since  $|e^{t+icU} S_n| = e^{tS_n} E_{n,5}$  can be bounded using Lemma 1 by

$$|E_{n,5}| \leq e^{nf(0)} (\exp n \sup_{x>d} (h(x) + tx) + \exp n \sup_{x<-d} (h(x) + tx)).$$

From (10) and the fact that  $t = -h'(0)$  we see that

$$g(t) = \max_x (h(x) + tx) = h(0) + 0. \tag{46}$$

Since  $h$  is concave it follows that  $\sup_{x>d} = h(d) + td$  and  $\sup_{x<-d} = h(-d) - td = h(0) + d^2 h''(0)/2 + O(d^3)$ , so we have

$$|E_{n,5}| \leq 2e^{n(f(0)+h(0))} e^{-nd^2 h''(0)/4} \quad \text{e.g.} \tag{47}$$

when  $d$  is small enough.

Consider now  $E_{n,4}$ . Using (46) we see that

$$E_{n,4} = e^{n(f(0)+h(0))} E(e^{n(g(t+icU)-g(t))}). \tag{48}$$

Since  $|e^{g(t+icU)-g(t)}| \leq 1$  it is no problem to truncate at  $|U|=D$ ,  $D \rightarrow \infty$  in (48). We have  $P(|U|>D) = O(e^{-D^2/2})$ . When  $|U| \leq D$  we use the Taylor expansion of  $g$  and get (with  $g'(t)=0$ ,  $v \equiv g''(t) = -1/h''(0)$ ):

$$n(g(t+icU) - g(t)) = n(-c^2 U^2 v/2 + O(cD)^3) = BvU^2/2 + O(D^3/n^{1/2}),$$

and

$$\begin{aligned}
 E_{n,6} &\equiv E(e^{n(g(t+icU)-g(t))}) \\
 &= E(e^{BvU^2}, |U| \leq D) (1 + O(D^3/n^{1/2})) + O(e^{-D^2/2}) \\
 &= E(e^{BvU^2/2}) (1 + O(D^3/n^{1/2})) + O(e^{-D^2/2}) \\
 &= (1 - Bv)^{-1/2} (1 + O(D^3/n^{1/2})) + O(e^{-D^2/2}).
 \end{aligned} \tag{49}$$

If we choose  $D = n^{1/8}$  e.g. we see that

$$E_{n,6} = (1 - Bv)^{-1/2} (1 + o(1)). \tag{50}$$

Collecting all the estimates (27), (43), (47), (48), (50) and putting  $d = D_1/n^{1/2}$  we see that we have

$$E_n = (1 - Bv)^{-1/2} e^{n(f(0)+h(0))} (1 + o(1)) \tag{51}$$

if we first choose  $D_1$  large enough and then  $n$  so big that  $r(d)$  in (43) is small enough.

When  $B > 0$  the above proof has to be modified somewhat. Now we use (44) with  $z$  imaginary, and (45) is changed into:

$$E_{n,3} = e^{nf(0)} E(e^{(t+cU)S_n}, |S_n/n| \leq d) \tag{52}$$

with  $c = (B/n)^{1/2} \rightarrow 0$  as  $n \rightarrow \infty$ .

Let  $J = [u_-, u_+]$  be the interval such that  $t + cJ$  corresponds to the interval  $[-d, d] \subset D_h$  in the mapping  $t \rightarrow x = g'(t)$ , i.e.  $u_{\pm}$  are determined by

$$g'(t + cu_{\pm}) = \pm d. \tag{53}$$

$0 \in J$  because  $g'(t) = 0$ . We want to compare the integral in (52) to

$$\begin{aligned} E_{n,7} &= E(e^{(t+cU)S_n}, U \in J) = E(e^{ng(t+cU)}, U \in J) \\ &= e^{nh(0)} E(e^{n(g(t+cU) - g(t))}, U \in J), \end{aligned} \tag{54}$$

and to this end we show that the two integrals

$$E_{n,8} = E(E^{(t+cU)S_n}, |S_n/n| > d, U \in J) \tag{55}$$

and

$$E_{n,9} = E(e^{(t+cU)S_n}, |S_n/n| \leq d, U \notin J) \tag{56}$$

are negligible.

For a fixed value  $U = u$  the integral in (55) e.g. can be bounded using Lemma 1 by

$$\begin{aligned} E(e^{(t+cu)S_n}, |S_n/n| > d) &\leq \exp n \sup_{x > d} (h(x) + (t+cu)x) \\ &\quad + \exp n \sup_{x < -d} (h(x) + (t+cu)x). \end{aligned}$$

Since  $u \in J$  in (55)  $h'(x) + (t+cu) \neq 0$  when  $|x| < d$ , so the sup above are attained for  $x = \pm d$  respectively, and since  $t = -h'(0)$  we have

$$\begin{aligned} h(\pm d) \pm td &\leq h(0) - d^2/2\bar{v} \quad \text{if} \\ \bar{v} &= \sup_{u \in J} g''(t+cu) = \sup_{|x| \leq d} (-1/h''(x)). \end{aligned} \tag{57}$$

From this we see that

$$\begin{aligned} E_{n,8} &\leq e^{nh(0) - nd^2/2\bar{v}} E(e^{ncdU} + e^{-ncdU}) \\ &= e^{nh(0) - nd^2/2\bar{v}} 2e^{(ncd)^2/2} = 2e^{nh(0)} e^{-nd^2(1 - B\bar{v})/2\bar{v}} \end{aligned} \tag{58}$$

because  $(nc)^2 = nB$ .

By the same argument we can bound  $E_{n,9}$  as in (58).

Returning now to (54) we have

$$n(g(t+cU) - g(t)) = nc^2U^2v/2 + O(ncd)^3 = BuU^2/2 + O(d^3/n^{1/2})$$

when  $U \in J$ , so

$$E_{n,7} = e^{nh(0)} E(e^{BvU^2/2}, U \in J) (1 + O(d^3/n^{1/2})).$$

From (53) we see that

$$cu_{\pm} = -h'(\pm d) - t = -h'(\pm d) + h'(0) = O(d),$$

so  $u_{\pm} \rightarrow \pm \infty$  if  $d/c = n^{1/2} d/B^{1/2} \rightarrow \infty$ , and

$$\begin{aligned} E_{n,7} &= e^{nh(0)} (E(e^{BvU^2/2}) + O(d^3/n^{1/2}) + O(e^{-(1-Bv)u_{\pm}^2/2})) \\ &= e^{nh(0)} ((1-Bv)^{-1/2} + O(d^3/n^{1/2}) + O(e^{-(1-Bv)u_{\pm}^2/2})). \end{aligned} \tag{59}$$

Collecting the estimates (27), (43), (52), (54), (58), (59) we see again that (51) is true if we choose  $d = D_1/n^{1/2}$  with  $D_1$  large enough and then choose  $n$  big enough as before. Let us sum up this result:

**Theorem 3.** *If  $f(x) + h(x)$  has a single maximum at  $x = 0 \in D_n$ , and if  $f(x)$  is twice continuously differentiable near 0 with  $-2A = f''(0) + h''(0) < 0$ , so (3) holds with  $a = 2$ , and if the regularity condition (26) holds then the asymptotic formula*

$$E_n = (1 + f''(0)/h''(0))^{-1/2} e^{n(f(0) + h(0))} (1 + o(1)) \tag{60}$$

holds for  $E_n$  defined by (1) as  $n \rightarrow \infty$ .

#### 4. The Central Limit Theorem for $n^{1/a}(S_n/n - \bar{x})$

In this section we show that a simple modification of the proofs of Theorem 1 and 2 shows that the limit distribution of  $Y_n = n^{1/a}(S_n/n - \bar{x})$  described in the introduction is obtained as  $n \rightarrow \infty$ .

The modified distribution of  $S_n$  was defined by  $F_{n,f}(dx) = e^{nf(x/n)} F^{n*}(dx)/E_n$ , so the characteristic function of  $Y_n$  is given by: (we take  $\bar{x} = 0$ )

$$\begin{aligned} E_{n,f}(e^{isY_n}) &= E(e^{nf(S_n/n) + isn^{1/a}(S_n/n)})/E(e^{nf(S_n/n)}) \\ &= E_n(s)/E_n \quad \text{for } s \in R^1. \end{aligned} \tag{61}$$

By the continuity theorem for characteristic functions it is enough to show that  $E_n(s)/E_n$  converges to  $\int e^{isx - A|x|^a} A^{1/a} dx / 2\Gamma(1 + 1/a)$  as  $n \rightarrow \infty$  when  $a > 1$ .

$E_n(s)$  is obtained from  $E_n$  simply by perturbing  $f(x)$  to  $f_n(x) = f(x) + isn^{1/a-1}x$ . If we note that since  $a > 1$   $f(x)$  and hence  $f_n(x)$  uniformly in  $n$  satisfy a Lip. condition,  $|f_n(x) - f_n(y)| \leq A'|x - y|$ , near 0 and that  $|e^{nf_n(x)}| = e^{nf(x)}$  we can repeat all the steps of the proofs of Theorem 1 and 2 almost as before and get a corresponding asymptotic formula for  $E_n(s)$ : (27) still holds for  $|E_{n,2}(s)|$ . In  $E_{n,5}$  we have  $d = n^{-1/2a}$ , and  $n(f_n(x/n^{1/a}) + h(x/n^{1/a})) \rightarrow -A|x|^a + isx$ , so by bounded convergence we have

$$E_{n,5}(s) = (n/2\pi)^{1/2} (-h''(0))^{1/2} e^{n(f(0) + h(0))} n^{-1/a} \int e^{isx - A|x|^a} dx (1 + o(1)). \tag{62}$$

(33) and (34) hold just as before, so from (35) we see that

$$E_n(s)/E_n = \int e^{isx - A|x|^a} A^{1/a} dx / 2\Gamma(1 + 1/a) (1 + o(1)). \tag{63}$$

In the proof of Theorem 2 we have to add a term  $isn^{-1/2}S_n$  in the exponent in (41) and (42). Since  $c = (-B/n)^{1/2}$  the exponent in (45) changes to  $(t + ic(U + s/(-B)^{1/2}))S_n$ . The only change in the following is hence that  $U$  is changed into  $U + s/(-B)^{1/2}$  (with  $(-B)^{1/2}$  imaginary if  $B > 0$ ). The rest of the proof then goes as before, and the main term in (49) and (59) is changed into

$$E(e^{Bv(U + s/(-B)^{1/2})^2/2}) = e^{-s^2v/2(1-Bv)/(1-Bv)^{1/2}},$$

and we have

$$E_n(s)/E_n = e^{-s^2v/2(1-Bv)}(1 + o(1)), \quad (64)$$

so the limit distribution is Gaussian with variance  $v/(1-Bv) = -1/(f''(0) + h''(0))$ . Let us sum up this result:

**Theorem 4.** *Under the assumptions of Theorem 2 and 3 respectively, when  $S_n$  is given the distribution  $F_{n,f}(dx) = e^{nJ(x/n)} F^{n*}(dx)/E_n$  then when  $a > 1$  the distribution of  $Y_n = n^{1/a}(S_n/n - \bar{x})$  converges weakly to the one defined by the density  $e^{-A|x|^a} A^{1/a} dx/2\Gamma(1+1/a)$  as  $n \rightarrow \infty$ . When  $a = 2$ ,  $A > 0$  this is a Gaussian with variance*

$$1/2A = -1/(f''(0) + h''(0)).$$

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