

# A Probabilistic and Combinatorial Analysis of a Digit–Sum Power Game

Fix integers  $k \geq 2$  and  $p \geq 1$ . The game proceeds as follows. There are  $N$  tables, each with exactly  $m$  students, so that the total number of participants is  $M = Nm$ . Each student independently selects a  $k$ -digit base–10 integer  $n$  uniformly at random from the set  $\{10^{k-1}, 10^{k-1} + 1, \dots, 10^k - 1\}$ . Equivalently, the leading digit is uniformly distributed on  $\{1, \dots, 9\}$  and each of the remaining  $k - 1$  digits is uniformly distributed on  $\{0, \dots, 9\}$ , all digits being independent. Writing

$$n = \sum_{i=1}^k d_i 10^{k-i},$$

with  $d_1 \in \{1, \dots, 9\}$  and  $d_2, \dots, d_k \in \{0, \dots, 9\}$ , define the digit sum

$$T = \sum_{i=1}^k d_i.$$

Each student computes the score

$$S = n - cT^p,$$

where the scalar  $c \in \mathbb{R}$  is fixed in advance and announced before play begins. Each table computes the average of its students' scores, and the class outcome is defined as the average of the  $N$  table averages. Since all table sizes are equal, the class outcome coincides exactly with the grand mean

$$\bar{S}_M = \frac{1}{M} \sum_{j=1}^M S_j$$

of the  $M$  independent scores.

Under the uniform distribution on  $k$ -digit integers, the expectation of  $n$  is

$$\mathbb{E}[n] = \frac{10^{k-1} + (10^k - 1)}{2} = \frac{11 \cdot 10^{k-1} - 1}{2}.$$

The digit sum satisfies

$$\mathbb{E}[T] = 5 + 4.5(k - 1) = 4.5k + 0.5, \quad \text{Var}(T) = \frac{20}{3} + (k - 1)\frac{33}{4}.$$

The exact distribution of  $T$  is given combinatorially by the generating polynomial

$$P_k(x) = (x + x^2 + \dots + x^9)(1 + x + \dots + x^9)^{k-1},$$

so that

$$\mathbb{P}(T = t) = \frac{[x^t]P_k(x)}{9 \cdot 10^{k-1}},$$

and consequently

$$\mathbb{E}[T^p] = \frac{1}{9 \cdot 10^{k-1}} \sum_{t=1}^{9k} t^p [x^t] P_k(x).$$

For analytic purposes, a second-order Taylor expansion of  $x^p$  about  $\mu_T = \mathbb{E}[T]$  yields the quadratic approximation

$$\mathbb{E}[T^p] \approx \mu_T^p + \frac{1}{2}p(p-1)\mu_T^{p-2}\sigma_T^2,$$

where  $\sigma_T^2 = \text{Var}(T)$ .

To enforce unbiasedness, define

$$c = c^* = \frac{\mathbb{E}[n]}{\mathbb{E}[T^p]},$$

so that  $\mathbb{E}[S] = 0$  and hence  $\mathbb{E}[\bar{S}_M] = 0$ . Substituting the quadratic approximation gives

$$c^* \approx \frac{\mathbb{E}[n]}{\mu_T^p + \frac{1}{2}p(p-1)\mu_T^{p-2}\sigma_T^2}.$$

The random variable  $S$  is bounded almost surely, since

$$10^{k-1} - c(9k)^p \leq S \leq 10^k - 1 - c.$$

In particular, all moments of  $S$  exist. Writing  $\sigma_S^2 = \text{Var}(S)$ , one has the exact identity

$$\sigma_S^2 = \text{Var}(n) + c^2 \text{Var}(T^p) - 2c \text{Cov}(n, T^p).$$

A quadratic approximation for  $\sigma_S^2$  is obtained by expanding  $T^p$  to first order around  $\mu_T$ , which gives

$$\text{Var}(T^p) \approx p^2 \mu_T^{2p-2} \sigma_T^2, \quad \text{Cov}(n, T^p) \approx p \mu_T^{p-1} \text{Cov}(n, T).$$

Since

$$\text{Cov}(n, T) = \sum_{i=1}^k 10^{k-i} \text{Var}(d_i) = 10^{k-1} \frac{20}{3} + \sum_{j=0}^{k-2} 10^j \frac{33}{4},$$

and

$$\text{Var}(n) = \frac{10^{2k} - 10^{2k-2}}{12},$$

one obtains the quadratic variance approximation

$$\sigma_S^2 \approx \text{Var}(n) + c^2 p^2 \mu_T^{2p-2} \sigma_T^2 - 2cp \mu_T^{p-1} \text{Cov}(n, T),$$

with  $c = c^*$  or its quadratic approximation.

Because  $S$  is bounded, Hoeffding's inequality applies. For all  $\varepsilon > 0$ ,

$$\mathbb{P}(|\bar{S}_M| \geq \varepsilon) \leq 2 \exp\left(-\frac{2M\varepsilon^2}{(b-a)^2}\right),$$

where  $a = 10^{k-1} - c(9k)^p$  and  $b = 10^k - 1 - c$ .

Moreover, since  $\mathbb{E}[S] = 0$  and  $0 < \sigma_S^2 < \infty$ , the Berry–Esseen theorem implies that

$$\sup_x \left| \mathbb{P}\left(\frac{\sqrt{M} \bar{S}_M}{\sigma_S} \leq x\right) - \Phi(x) \right| \leq \frac{C \mathbb{E}|S|^3}{\sigma_S^3 \sqrt{M}},$$

where  $\Phi$  denotes the standard normal distribution function and  $C$  is an absolute constant.

Consequently, for fixed  $\alpha \in (0, 1)$  and letting  $z_{1-\alpha/2}$  denote the  $(1 - \alpha/2)$ -quantile of the standard normal law, the central limit theorem yields the asymptotic design condition

$$\mathbb{P}(|\bar{S}_M| < \gamma) \approx 1 - \alpha \quad \text{when} \quad M \approx \left( \frac{z_{1-\alpha/2} \sigma_S}{\gamma} \right)^2.$$