## An Ordered Sample Mean that's a bit like Simpson's Rule

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Abstract: A "Simpson's Rule"-like ordered sample mean is compared with the standard version

Consider the following :

1. Take a sample $X_{1}, X_{2}, X_{3}, \ldots, X_{3 n}$ of size $3 n$.
2. Order the sample by swapping terms such that $X_{1} \leq X_{2} \leq X_{3} \leq \ldots \leq X_{3 n}$
3. Calculate the "Simpson's" Ordered Sample Mean:

$$
\bar{X}_{a l t(3 n)}=\frac{1}{8 n}\left(3\left(X_{1}+X_{3}+X_{4}+X_{6}+\ldots+X_{3 n}\right)+2\left(X_{2}+X_{5}+\ldots+X_{3 n-1}\right)\right)
$$

For example:

1. Take a sample of size 6 from $\exp ($ ran\# $)$ where ran\# is a random number between 0 and 1 .
2. Order them (line 2 in table below).
3. Calculate ordered sample mean and compare with standard sample mean (lines 3 and 4 below)

| random numbers ordered | 0.032316372 | 0.1709593670 .5464619095 | 0.6212804590 .71926393550 .8565404457 |
| :---: | :---: | :---: | :---: |
| exp(ran\#) | 1.0328442166 | 1.18644253931 .7271314476 | 1.86130984752 .05292157162 .3549993364 |
| sample mean | 1.7026081599 |  |  |
| ordered sample mean | 1.7129739229 |  |  |

Notice the ordered mean is closer to the population mean $(1.71828 \ldots=(\mathrm{e}-1)=$ integral of $\exp (\mathrm{x})$ from 0 to1) than the standard mean.

Is this more often the case than not? Heuristic arguments suggest it might be.
The Ordered mean above is constructed by using the expression

$$
\bar{X}_{a l t(3)}=\frac{1}{8}(3(\min +\max )+2(\text { median }))
$$

(where $\min =$ minimum, $\max =$ maximum, etc on a sub-sample of size 3 )
n times on n sub-intervals of the interval $(0,1)$.
See "A Better Type of Sample Mean?" and "An Alternative Model of Probability Theory" at
vixra.org/author/d_williams for more details.
An integral approximation (for non-decreasing $f$ on $(0,1)$ )

$$
\begin{aligned}
& \quad \int_{0}^{1} f(x) d x \approx \frac{1}{8 n}\left(3\left(f\left(\frac{1}{2 n}\right)+f\left(\frac{5}{2 n}\right)+f\left(\frac{7}{2 n}\right)+f\left(\frac{11}{2 n}\right)+\ldots f\left(\frac{2 n-1}{2 n}\right)\right)+\right. \\
& \left.2\left(f\left(\frac{3}{2 n}\right)+f\left(\frac{9}{2 n}\right)+\ldots\right)\right)
\end{aligned}
$$

is related to the first expression. This can be generalised to other finite intervals.
Testing this expression on 6 "random" functions with $\mathrm{n}=6$ gave the following results:

|  | Integral value | Mid Point Rule | Simpson-like Rule |
| :--- | ---: | ---: | ---: |
| $2 x^{\wedge} 3+7 x^{2}+3$ | 7 | 6.9930555556 | 7 |
| $x^{\wedge}(1 / 3)$ | 0.75 | 0.7548623802 | 0.7530626264 |
| $10 \sin \left(\right.$ pi $\left.^{*} \times / 2\right)$ | 6.3661977237 | 6.3844146463 | 6.365864624 |
| $10 \sin \left(\mathrm{pi}^{*} \times / 4\right)$ | 4.4127120031 | 4.40368436 | 4.4122130267 |
| $\exp (\mathrm{x})$ | 1.7182818285 | 1.7162946864 | 1.7182674326 |
| $\ln (1+\mathrm{x})$ | 0.3862943611 | 0.3868714395 | 0.3863079107 |

Notice the "Simpson-like Rule" is better than the Mid Point Rule in each case.
Other types of sample means based on other types of integral approximation are possible and should be worth exploring.

It may be the case that a particular ordered sample mean better approximates the population mean more often than another if the integral approximation derived from its $f(\operatorname{ran} \#)$ function for that particular $n$ is closer to the population mean. A proof or refutation of this would be nice.

