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## Slogan

A relatively small number of, *dimensionality independent* evaluations of a target function suffices to give a good estimate of the integral of the function ... but the requisite sample size becomes untenable if very high precision is necessary.