

# Hyperdimensional Sparse Space

## Some Mathematical Properties

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From: *Pentti Kanerva - Sparse Distributed Memory, MIT Press 1988, ISBN 978-0-262-51469-9*

For a rigorous treatment see:

*Fredrik Sandin and Blerim Emruli and Magnus Sahlgren, Incremental dimension reduction of tensors with random index, <http://arxiv.org/abs/1103.3585>*

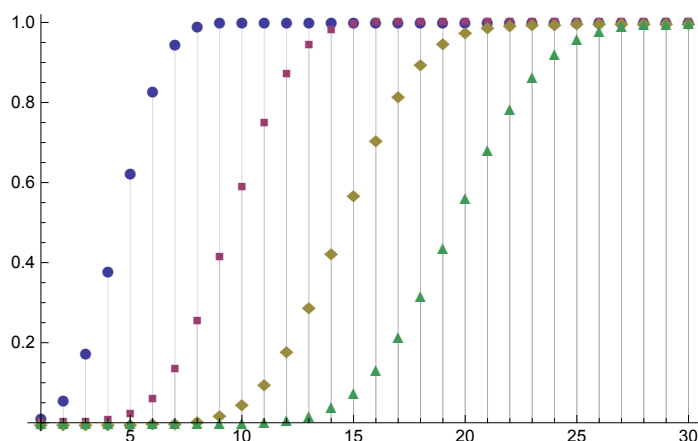
In the binary space  $\{0, 1\}^n$  of vectors of length  $n$  with equal probability of the states 0 and 1 with a distance metric defined as the *Hamming-distance* or the number of non-zero bits in the *xor* of vectors, (equivalent to the square of the Euclidean distance), then the number of vectors in the space that are at a distance  $d$  from a specific vector is given by the binomial coefficient:

```
NumberVectorsAtDistance[n_, d_] := Binomial[n, d]
```

The number of vectors at a certain distance from a reference vector therefore is given by the binomial distribution with probability  $p = \frac{1}{2}$ , which has mean  $\frac{n}{2}$  and variance  $\frac{n}{4}$

```
DistanceFromGivenVector[n_, p_] := BinomialDistribution[n, p]
```

```
DiscretePlot[  
  Evaluate@Table[CDF[DistanceFromGivenVector[n, 0.5], k], {n, {10, 20, 30, 40}}],  
  {k, 30}, PlotRange -> All, PlotMarkers -> Automatic]
```

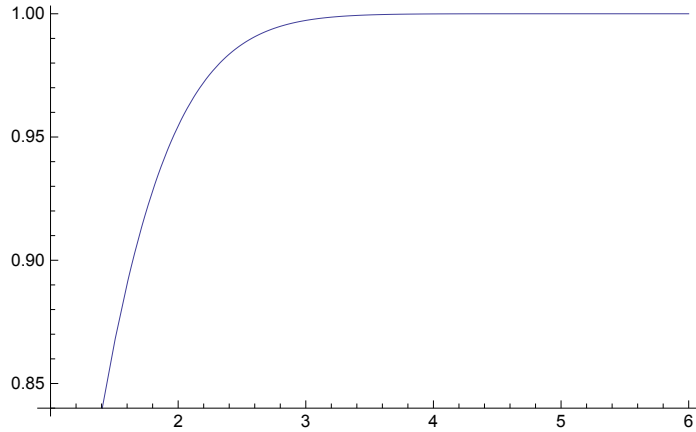


For large values of  $n$  the binomial distribution can be approximated with a normal distribution; therefore the proportion of vectors within  $z$  standard deviations of the mean is  $\text{erf}\left(\frac{z}{\sqrt{2}}\right)$ . The distance distribution is highly concentrated around the mean as the error function quickly approaches

unity for increasing  $z$ :

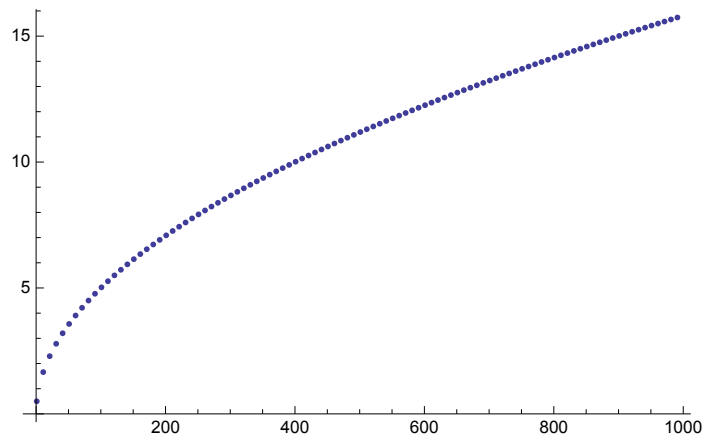
```
ProportionOfVectorsWithinStdDeviation[z_] := Erf[z /  $\sqrt{2}$ ]
```

```
Plot[ProportionOfVectorsWithinStdDeviation[z], {z, 1, 6}]
```

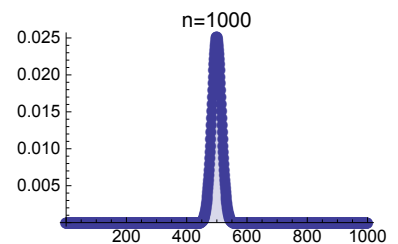
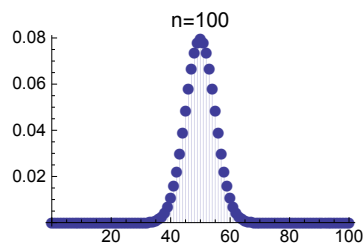
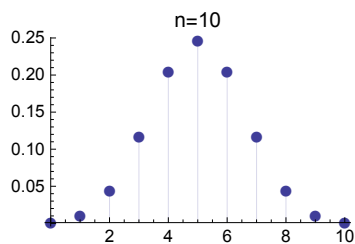


The mean distance is  $n/2$  and the standard deviation of distance is  $\frac{\sqrt{n}}{2}$ . This implies that the mean distance is  $n/2$  standard deviations, thus in a high-dimensional binary space all vectors are located at a distance  $\sim n/2$  from any other vector.

```
ListPlot[Table[{n,  $\sqrt{n} / 2 // N$ }, {n, 1, 1000, 10}]]
```



```
GraphicsRow[Table[DiscretePlot[PDF[BinomialDistribution[n, 0.5], k], {k, 0, n}, PlotRange -> All, PlotMarkers -> Automatic, PlotLabel -> StringForm["n=``", n]], {n, {10, 100, 1000}}]]
```



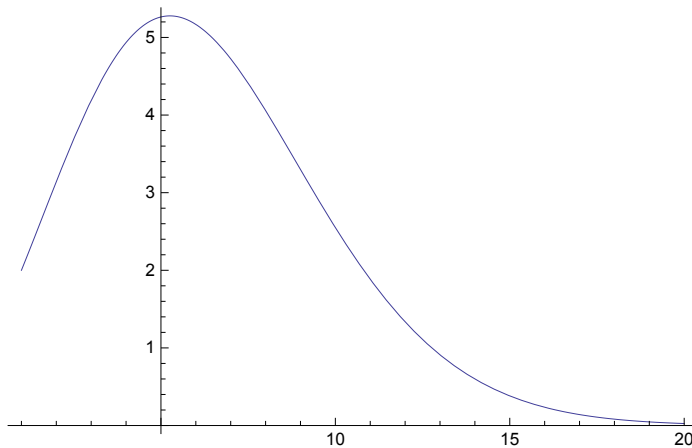
## The volume (content) of the n-cube and n-sphere...

`VolumeOfUnitNCube[n_] := 1n`

`VolumeOfUnitNSphere[n_] :=  $\frac{\pi^{n/2}}{\Gamma[\frac{n}{2} + 1]}$`

The volume of a hypersphere is interesting itself ... and actually  $\rightarrow 0$  as  $n \rightarrow \infty$  but is greatest at just over five dimensions!

`Plot[VolumeOfUnitNSphere[n], {n, 1, 20}]`



*All the space is in the corners? Well there's no space in the middle anyway...*

`ListPlot[Table[Log[VolumeOfUnitNCube[n] / VolumeOfUnitNSphere[n]], {n, 1, 80, 1}]]`

