Hypothesis: The probability of "success" is directly proportional to the distance between the positive and negative data, inversely proportional to the input dimension, and directly proportional to the number of nodes in the hidden layer.

Data : ball-shell dataset

Number of samples : 500

rb : radius of ball (positive data(ball) is from 0 to rb.)

 $\ensuremath{\mathsf{rs}}$: radius of shell (thickness of shell is fixed to 1 so negative data(shell) is from $\ensuremath{\mathsf{rs}}\text{-}0.5$ to

rs+0.5.)

d: distance between the positive and negative data ((rs-0.5)-rb)

n: list of number of nodes in the hidden layer

For training stopping conditions are:

- 1. Minimum 50 epochs
- 2. Maximum 400 epochs
- 3. Best_accuracy accuracy >= 5
- 4. Accuracy is not changing for last 20 epochs

initialisation_method = 'xavier' optimisation_method = 'sgdwm' learning_rate = 0.05, momentum = 0.9

Experiment 3

points_dim = [2, 4, 8, 16, 32]
radius_of_sphere = 1
radius_of_shell = [1.2, 1.4, 1.6, 1.8, 2]
thickness_of_shell = 1
Random_seed = [100, 200, 300, 400, 500, 600, 700, 800, 900, 1000]
N_nodes = [2, 3, 4, 6, 8, 12, 16, 24, 32, 48, 64]

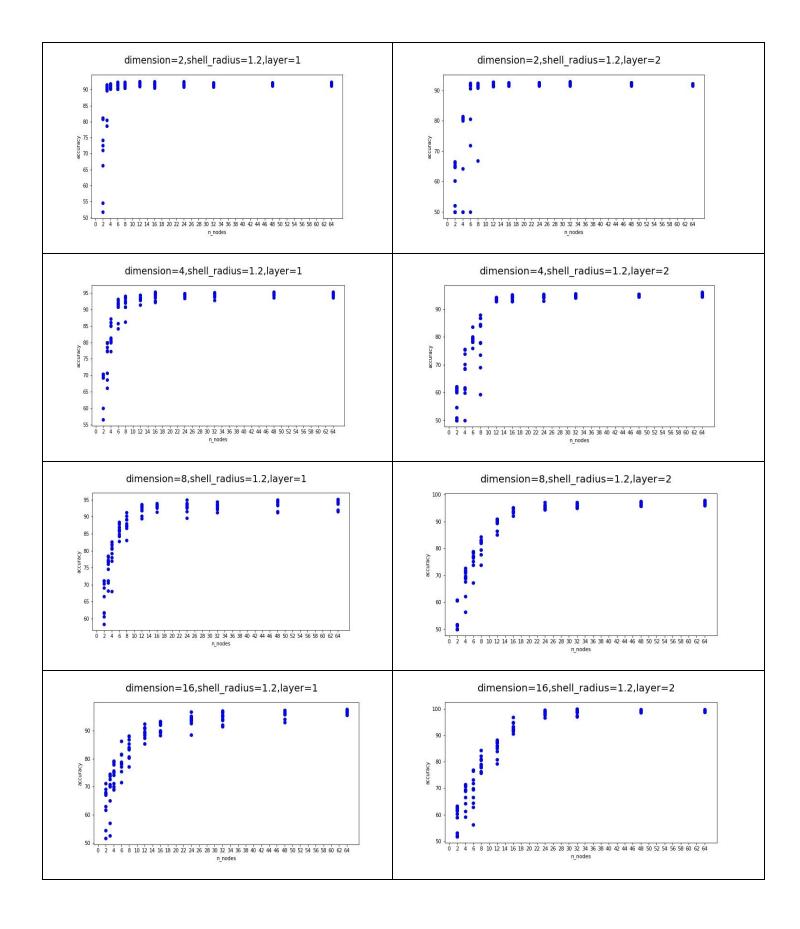
Hidden layer: 1, 2

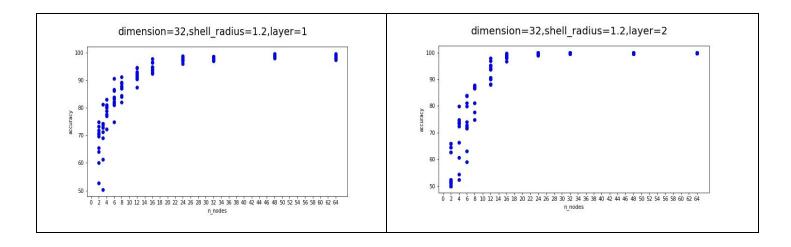
Base Risk:

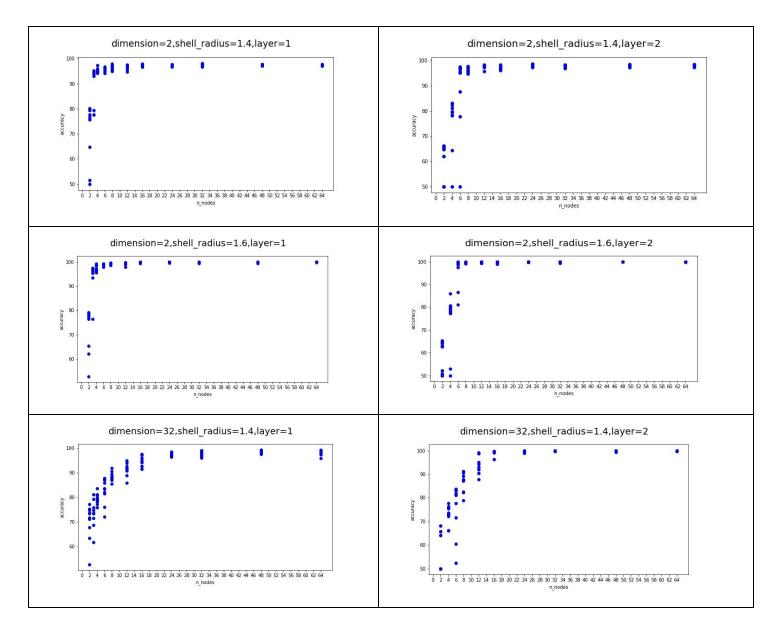
shell_radius	dimension				
	2	4	8	16	32
1.2	0.10625	0.04684	0.00676	0.0001024	2.1115e-08
1.4	0.03393	0.01389	0.00168	1.4123e-05	5.8034e-10

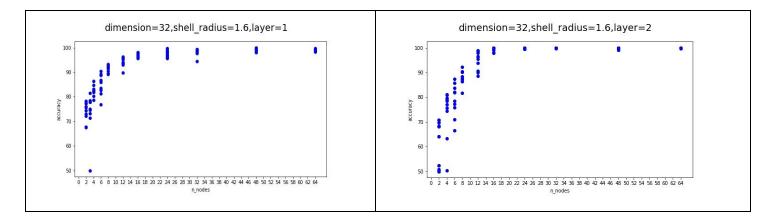
Comparing number of layers with different number of nodes:

Accuracy is less for 2 layers(<16 nodes) than the same number of nodes in 1 layer (<16nodes).

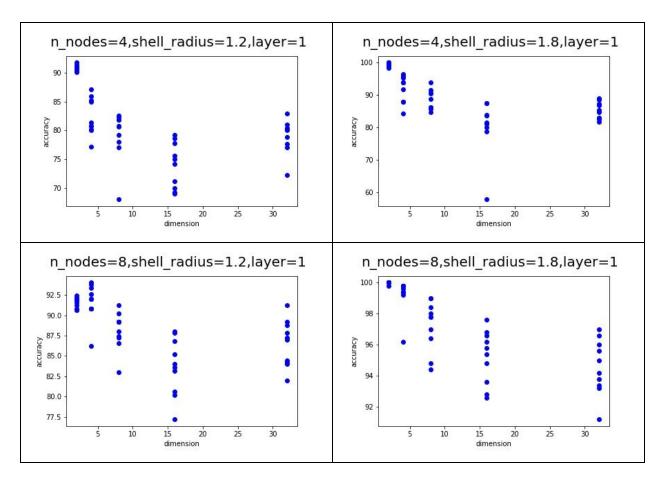








Comparing dimension: it achieves better accuracy on 32 dimensions data compared to 16 dimensions data.



Comparing shell_radius:
Accuracy is increasing as distance between data increases.

