







Kurumbapalayam(Po), Coimbatore - 641 107 Accredited by NAAC-UGC with 'A' Grade Approved by AICTE, Recognized by UGC & Affiliated to Anna University, Chennai

### Department of AI &DS

Course Name - 19AD602 DEEP LEARNING

III Year / VI Semester

**Unit 2-DEEP NETWORKS Topic:VC DIMENSIONS** 



GULSHAN BANU.A/AP/AI AND DS /VC DIMENSION /SNSCE





### Vapnik – Chervonenkis (VC) Dimension

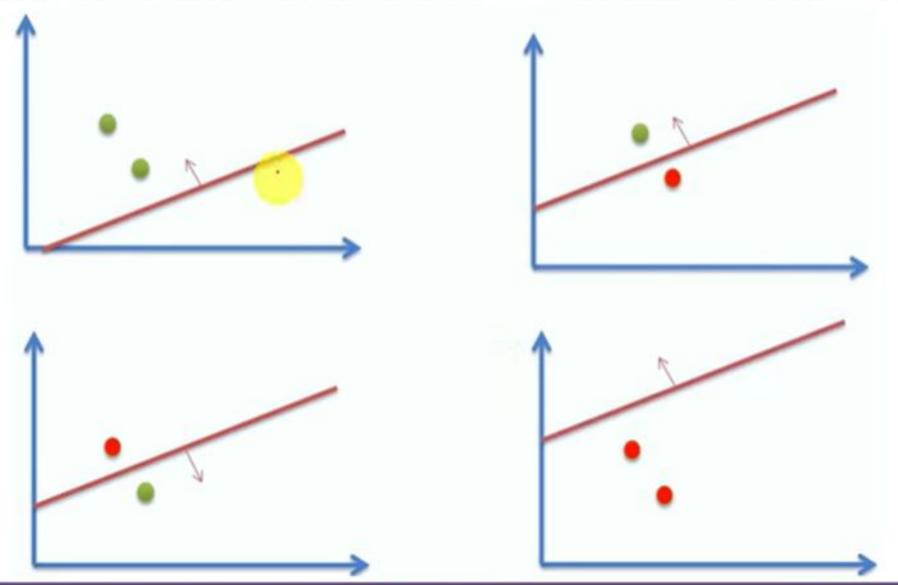
- It was introduced by Vladimir Vapnik and Alexey Chervonenkis in the 1970s and has become a fundamental concept in statistical learning theory.
- The VC dimension of a hypothesis set H is the largest number of points that can be shattered by H.
- A hypothesis set H shatters a set of points S if, for every possible labeling of the
  points in S, there exists a hypothesis in H that correctly classifies the points.





# Vapnik - Chervonenkis (VC) Dimension

• If there are 2 data point, then there are  $2^2=4$  different classifiers exists.







## Vapnik - Chervonenkis (VC) Dimension



- If there are 4 data point, then there are  $2^4 = 16$  different classifiers exists.
- The Maximum Number of data points that can be shattered by a straight line in
   2-Dimissional space is 3.

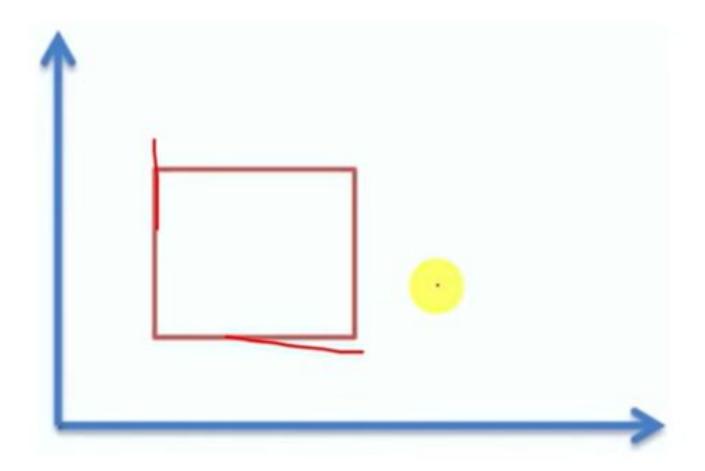






# Vapnik – Chervonenkis (VC) Dimension

Axis aligned rectangle

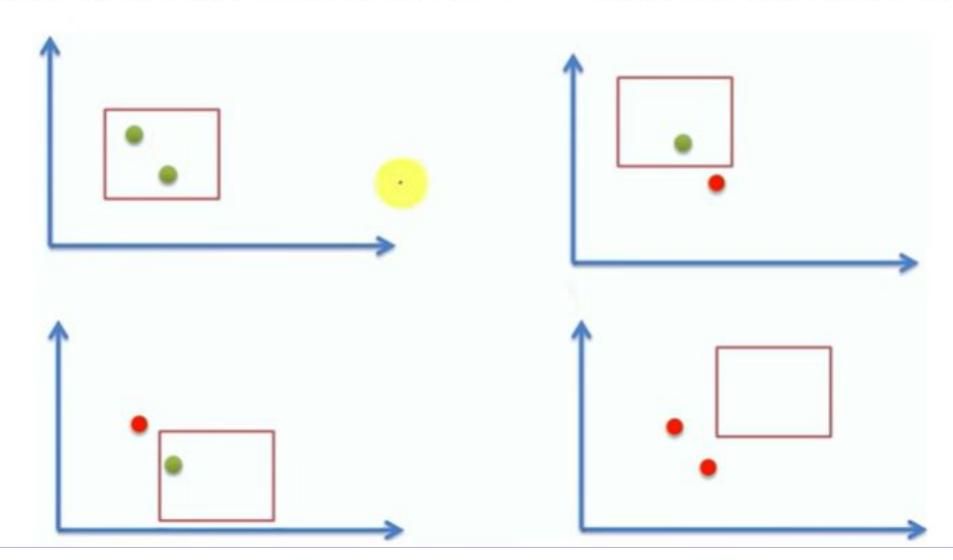






## Vapnik - Chervonenkis (VC) Dimension

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### **Applications of VC Dimension**

- The VC dimension has a wide range of applications in machine learning and statistics.
- For example, it is used to analyze the complexity of neural networks, support vector machines, and decision trees.
- The VC dimension can also be used to design new learning algorithms that are robust to noise and can generalize well to unseen data.
- The VC dimension can be extended to more complex learning scenarios, such as multiclass classification and regression.





### THANK YOU