



# Probabilistic hierarchical Model

Presented By Jainil Jain

# Things we will go through

Presentation Overview	1	Hierarchical Models	2
Types of Hierarchical Models	3	Why use Hierarchical Models	4
Probabilistic Models	5	GMM probabilistic model	6
what Lacked?	7	probabilistic Hierarchical Models	8

# Presentation Overview

Probabilistic Hierarchical Model: The ultimate combination of two of the most popular classification techniques in ML but first the basics.



## Hierarchical Classifiers

Flexibility of the Agglomerative and divisive hierarchical classifiers



## Probabilistic Classifiers

Data distribution understanding of the probabilistic GMM classifiers.

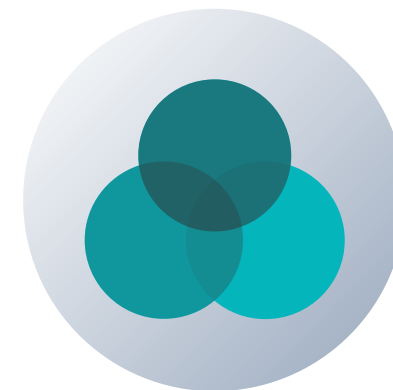
# Hierarchical Models

It is a powerful unsupervised machine learning algorithm used to group data points into a hierarchy of clusters.



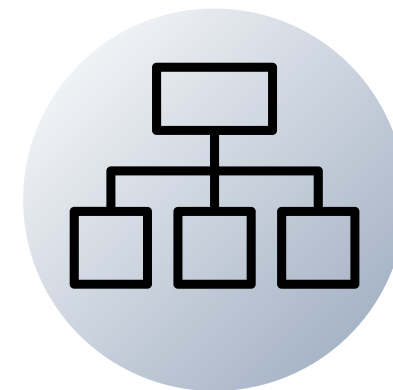
## No predefined Clusters

Don't know how many clusters you need? we got that covered



## Dendrogram diagrams

Want quick visualization of the cluster distribution? we can do that too

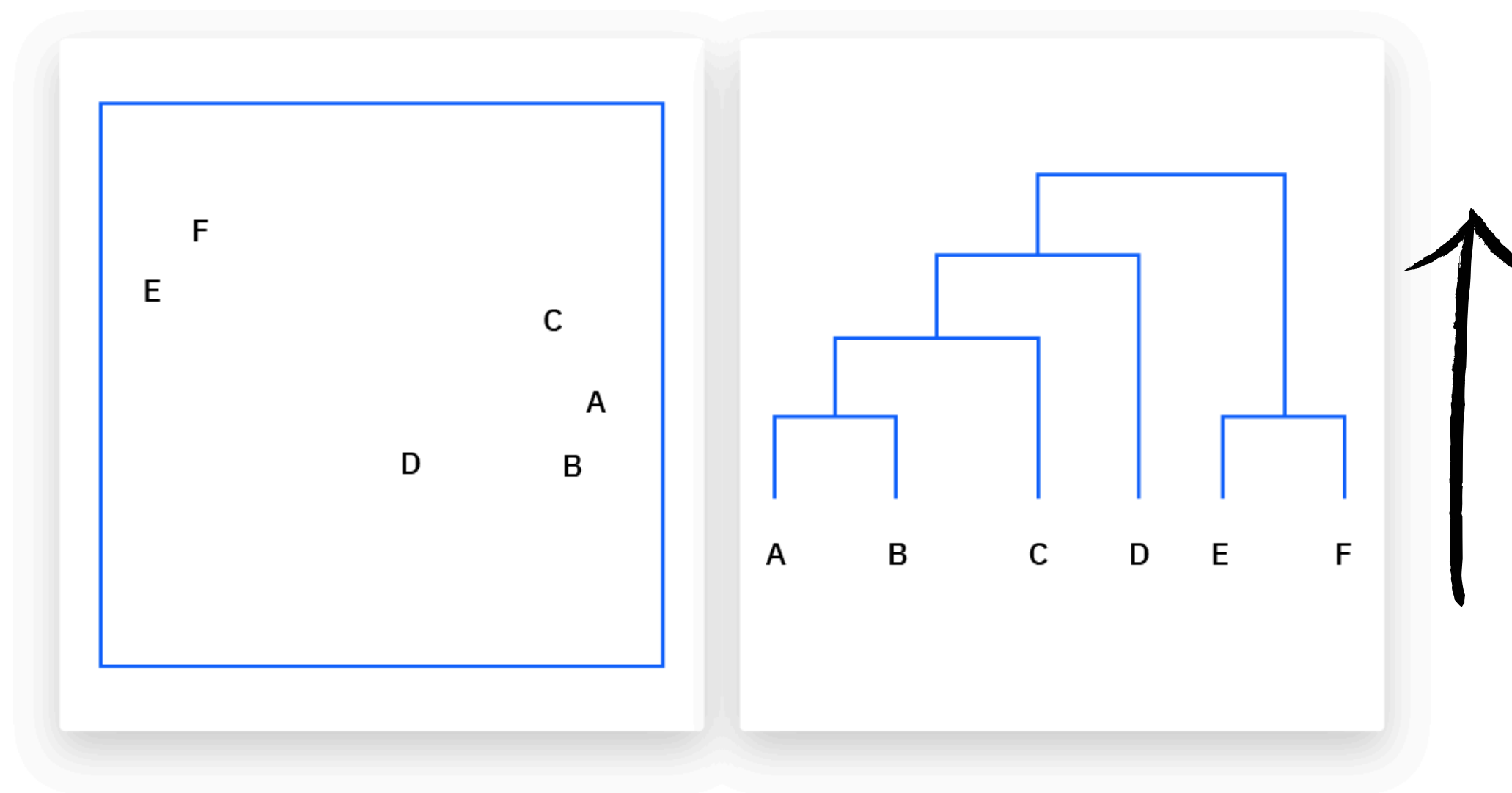


## Cluster Hierarchy

Want to know the hierarchical structure? don't worry about it

# Agglomerative approach

Bottom up approach with each data point a indiv. cluster



what is it?

ITS A BOTTOM-UP APPROACH FOR DEALING WITH CLASSIFICATION PROBLEMS USING HIERARCHICAL STRUCTURE

methodology

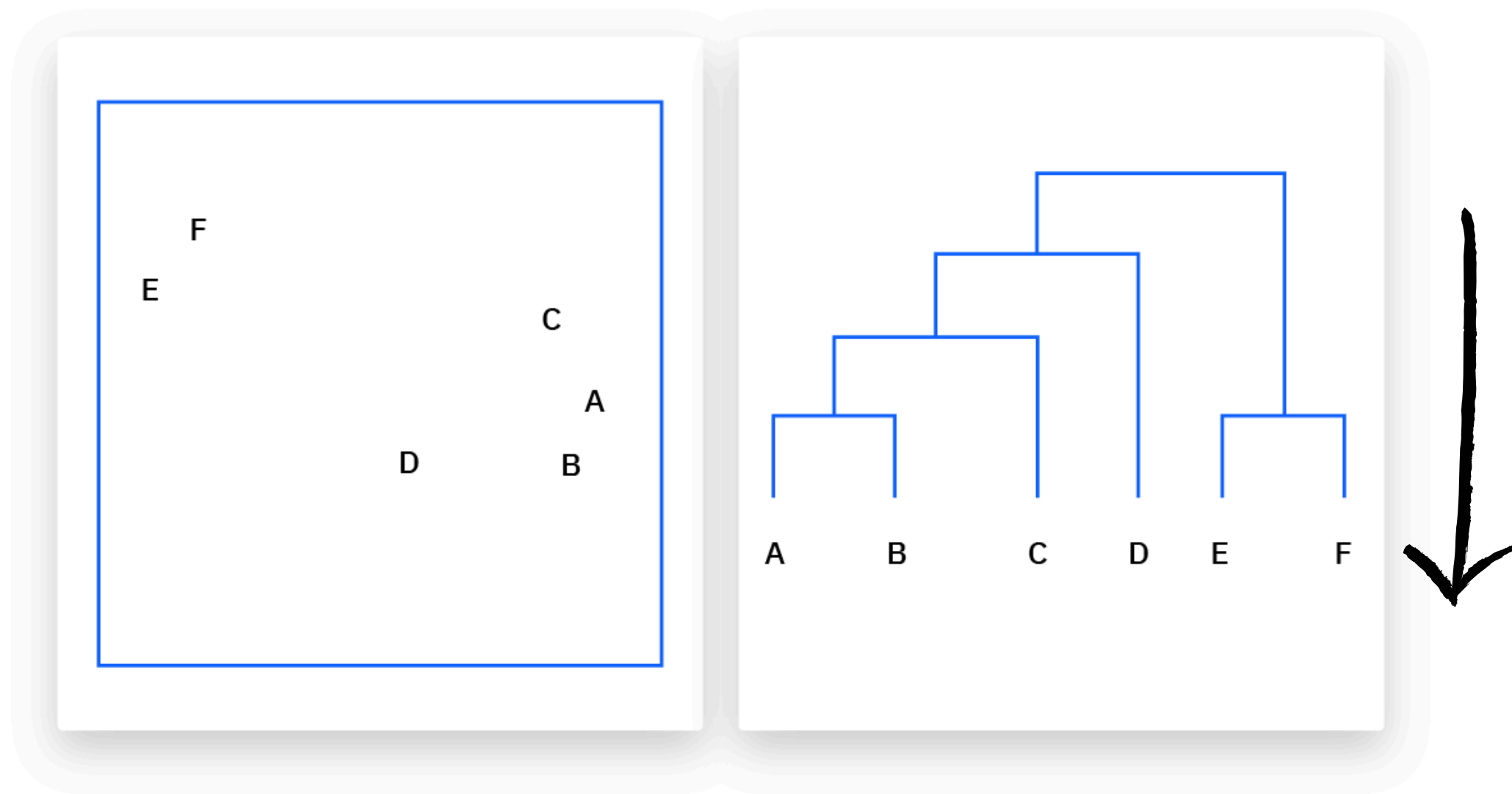
USES DATA POINT LINKAGE TO DEFINE RELATIONS AND MERGE

benefits

EASY TO IMPLEMENT AND COMPREHEND

# Divisive approach

Top-down approach with all the data points represented as a single cluster



what is it?

TOP-DOWN APPROACH WITH ALL THE DATA POINTS IN A SINGLE CLUSTER ITERATIVELY SEPARATED

methodology

USES CUT COST AND INTRA COST TO JUSTIFY SPLIT.

benefits

BENEFICIAL WHEN EACH ITERATION IS TO BE CHECKED FOR PATH

# Why Bother about Hierarchical Models?

what advantages does this guy offer?

## FLEXIBILITY

No need to specify number of clusters

## INTUITIVE AND EASY TO GET

Dendrograms provide easy visualization

## DETERMINISTIC

Same clusters every time any time

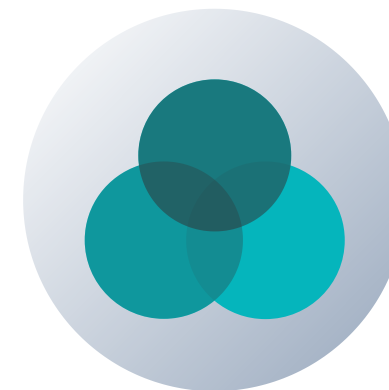
# Probabilistic Models

Models that incorporate the concepts of probability to define and handle uncertainty in the data and make predictions



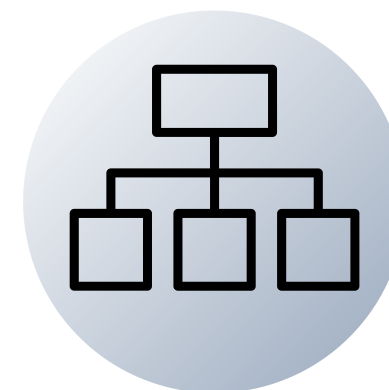
## Uncertainty Quantification

Can provide a measure of uncertainty for those risking problems.



## Flexibility in Complexity

Can handle complex and high-dimensional data like a breeze.



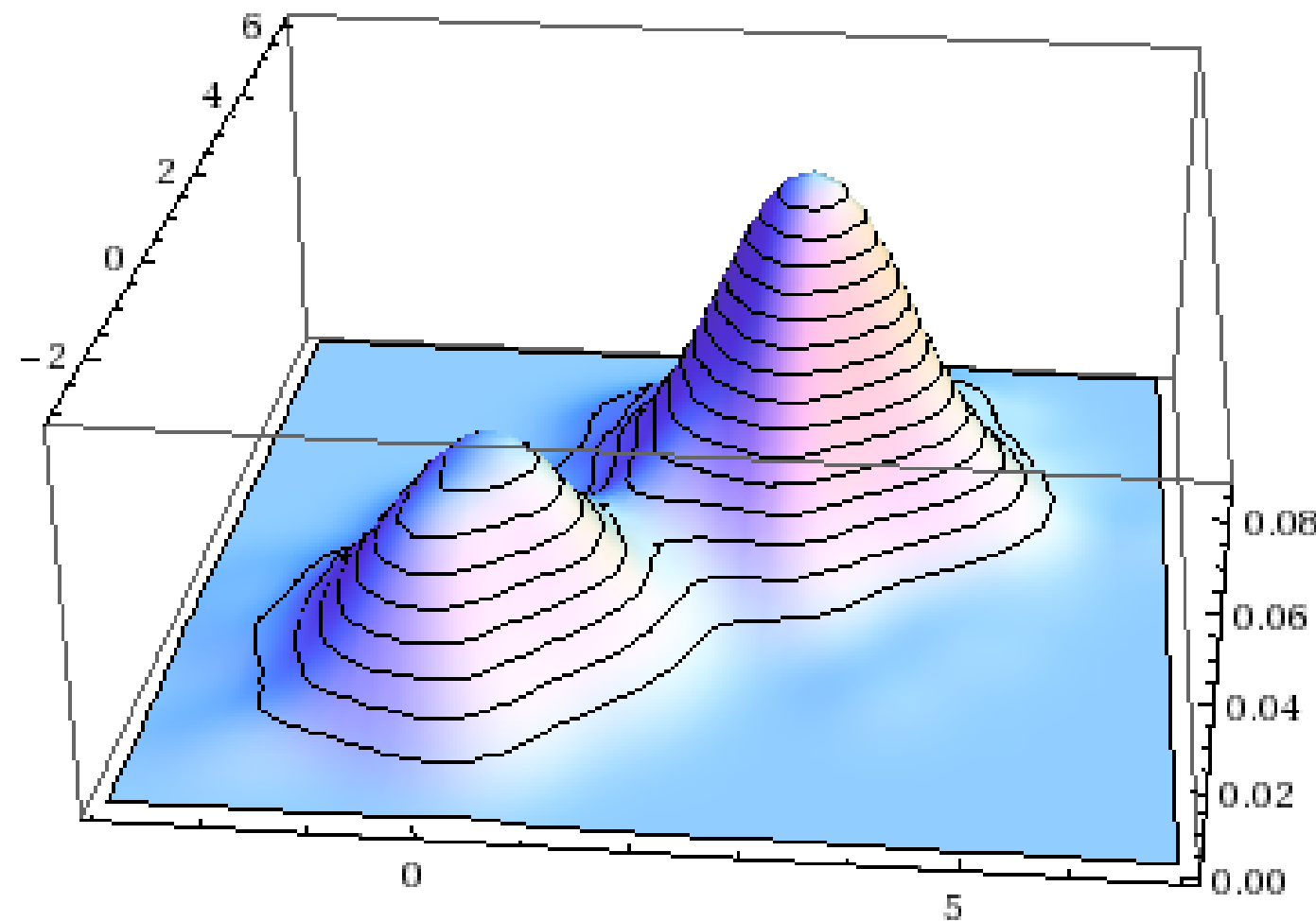
## Probabilistic inference

Dynamic hypothesis updates to make predictions easy and life of ML engineers livable.



# Gaussian Mixture Model (GMM)

Top of the line probabilistic model for classification and clustering



Mixture of Gaussian's

A MODEL FORMED BY COMBINING THE GAUSSIAN CLASSIFICATION OF VARIOUS CLUSTERS TOGETHER

Soft Clustering

PROVIDES PROBABILITY OF A DATA POINT EXISTING IN MULTIPLE CLUSTERS ALLOWING FLEXIBILITY

Robustness to outliers

PROBABILITIES REDUCE THE RISK ASSOCIATED WITH HARD LABEL CLASSIFIERS

# What Lacked?

Until now we have been repeating that these models are the best in their class but where are they lacking?



high complexity when  
dealing with large  
datasets



limited flexibility in  
complex datasets



limited update potential  
and can suffer from bias  
caused by linkage

# Probabilistic Hierarchical Models

builds on top of the Hierarchical model with a little squeeze of probabilistic freshness

## **likelihood based classification**

no more distance based uncertainty linkage

## **complex data? no problem**

Using the internal probabilistic likelihood can better understand complex data nature

## **uncertainty causing issues?**

outliers dont have much effect now thanks to probabilistic nature

## **built in Flexibility**

can handle more complex distributions and true to the data