Machine dearning Machine dearning: Learnor train model/logic from data classifal dearning: Define algo/logic to Compute output Classical dearning Data (input) s out put Program Algo logic Machine dearning.

Deta (input) / Training. > model / logic / Program ontput machine decening Basic Advanced - Concept Transfer Unsupervised Reinforcement Deepharmy deem Supervised Classification Reg clusting DR model model band free K-mean Le PeA 1 Heraelmal LLDA Dynamic A-denny KNIN LR Logit Naive

## Con upt dearning / Concept dearning as Secret

Conapt dearning is used to recognize patterns or Concept within data set. The general idea behind Concept learning is to develop a hypothesis erset of band on their Athributus (Features)

Basically et is entension of Supervised Learning

\*> Span mail identication.

Alzheimus Disease Staging using Compt deceming

- Carly, mild, moderate, Severe

Correlation blw features -> Pattern of features.

( for training)

finding Conopt within bedast forenp: - Hippocampus

Volume Redbuck L. Serve.

Cencept dearing as Search!

Secureling a Soldition through a hypothesis is pare. For best Sits 'Seach strategy or solution to a problem

for best sits 'Seach strategy or solution to a problem									
Hupothesis Space.									
Conapt Space									
Example !	budan	Conept							
Constomer ID	Age	Gender	Income	Amount	Member togh				
	35	M	TOK		y 5 FF Y				
2	45	F	75 K		H8 K				
3	28	M	40K	150	3 Yw) N				
4	50	·F	1000 K 100000	500	10 xes y				
	40	M	60 K	250	6 NO N				

IH: - if a customer Spends more than \$300, they are a high-value Customer

Evaluation: - Incorretty Classifies Some high Value Constomers who Spend less than \$300

RH! if a customer Spends most than \$300 and is a member of the loyalty Proport, they are a high-value Customer.

But mis classify non non high Value Customs who also meet same Ending.

fR! - if a astomer Spards more than \$300 cr , is a number of the logalty program, , and has an in come greater than Gok thy are high valued continuer.

Arimal				Label
Ani mal	fur	Loy Eggs	wings M.	ma mual
Dog	Yel	Н	7	Bird
eagle	7.	H	н .	Reptile
Snoke	H	H	*	mammed
Bat	4	V.	17	Reptilu.
Perguir	, N	4	М.	
Crocode	W 19			

if or animal has fur, it is mammal.

is not us. hay and constant

Concept dearning-find(s) manumom specific Hypothesis								
	1		Humidity		water	forecast		
1	Sunny	warm	Normal	Storng	warm	Same	Yes	
2	Sonny	warm	High	Strong	11	Same	Xes	
3	Rainy	cord	Heigh		17	Charge		
4		Warm	Hligh	) (	Cool	,,	Yes	
							1	
			Attribut	e			Target Value	
1.	1. initilize to the most specific hypothesis in H							
		~ ·			4 >			
		h(0) = 4	(0,0,0)	0,91	1/			
	$h(0) = \langle Q, Q, Q, Q, Q, Q \rangle$ $ Q_1   Q_2   Q_3   Q_3   Q_3 $							
	,	X1 = 1 C. 1.	/ / /		1 chan	a Janu	(any)	
7 +7	re bon-		my, warm			g, varm,		
	→ -P	= { com	14 1.26am	Kl man ii	1 Char	10 Warm	(ame)	
	ni Song warm of Normal, Strong, Same							
This (Sumy warms, Hornal, Strong, warm, Same)  Itorelev-2  Sunny, warn, high, story, warn, Same)								
-> 1/2 - ( sorry, wam, night, stry, wam, same)								
-> thz = < Sunny, warm, ?, strong, warm, Same								
Iterahar 3 - Ignose -> 23								
- th3 = < Sunmy, wam, ?, Strog, wam, Same)								

iteration-4
X4 = {Sunny, warm, Frigh, string cool, chape}

that = {Sunny, warm, 8, String, 8 ?}

if so Sty is sunny, Arrtem is warm, ook,

if Sty = Sunny, Airtem = warm, crind = String

then only enjoy spects.

						hur
escomple	atabay	Size	irub	price	edting	9-1
				afferbb4	many	ho
	Some	Small	NO	9 (15, 15)	/	
2	may.	big	no	Expersive	one	Xrey
0			0		few	No
3	Some	big	always	4,		
4	mary	1	no	1)	many	1/mg
	mary	medium	10-	011 006	in many	Yes
S	mary	1 Small	No	afferdab	1	-
	1					

the catalians many, inclubrage no)=buy (Yes)

How may concept are lossible: 
2×3×2×2×3 =72

How many typothesis are possible: -

Distinct Hypothesis: - (3×4×3×3×4)+1=433

Consistant typothesis Candidate Elimination Algorithm: wind water forecast Emaystr Sky flumidity Same Ves AirTemp strong worm Sunny Warm Hormal Same Yes Strong Vanm Sunny High Warm Change HO Roany Storng Warm Cold High Sunny charge Yes 1-ligh warm Storng Corol So! most spenfic <p, 0, 0, 0, 0, 0) ( Sunny, Jam, Momal, String, warm, Same) Sumy, wam, be ?, Strg, wam, Sare) S2: S3 S4! ( Sunny, Jam, ? Stry, ??)

G3: (Sunny, ?,???) (? wam, Hand (Cont) (Sane)

G2: (?,?!??)

G1: (?,?!??) Mo most Go: Generic bourdaries <?,?,?,????> 91. 9 S (Sunny, war ? Shing ? ?) Ksuny, ?? sty,?? ?> (Suny, vcm, 1273) <? vam, ? strg,??>

G (Sonry,?)?) (? wm,?????) Potal = 6

Version Space: Subset of Hypothesis H Consistant with the training exampleD VSH,D = ShEH | Consistant (t,D) } (this consisted with D) Where H=hypothesis
D = Training examples. Consistent? = h(x) = C(x)4) Tourget function ty pothesis how=cery

$$\frac{\text{Exp'}}{\text{(h) (h)}} \xrightarrow{\text{(h) (h)}} \frac{\text{(h) (h)}}{\text{(h) (h)}} = \frac{\text{(h) (h)}}{\text{(h)$$

Exp!

IMIP.

< Sunny, warm, ? Strong,?,?>

(Sunny, ?? String, ??) (Sunny, warm, ?????)

< sunny,?,?,?,???> <?, warm,?????>

1 may rethang								
	Sky	Airtemp	Humidity	wind	water	faceast	Enjoys burb	
A	sunny	warm	Hormal	Strong	Cool	change	7	
B	Rainy	Cold	Normal	light	warm		,	
	Sunny	Warm	Hormal.	light	warm			
		Cold	Homal.	Stong	warm	Same	?	
						1		

Countain not / with the excupli Lib Arl Price edution Buy Exp. Citations Size · < lok one Mo No 1 Some Small Yes many. >10K 40 2 may Big thi= (?,?, No,?, Mary) - No y Const xes h2: (?,?, No,?,?) -> not consist +nt