



MACHINE INTELLIGENCE

Concept Learning

K.S.Srinivas

Department of Computer Science and Engineering

MACHINE INTELLIGENCE

Concept Learning

Srinivas K S.

Associate Professor, Department of Computer Science

MACHINE INTELLIGENCE

Concept Learning

- what is a **concept**?
- we will try to understand it practically

Let's make labels for images with labels of chairs. A machine learning algorithm might learn to label images with labels if it is given a set of images and labels. It might learn to label images with labels if it is given a set of images and labels. It might learn to label images with labels if it is given a set of images and labels. It might learn to label images with labels if it is given a set of images and labels.

Concept

Data object

C

x

label

x- belongs to concept c

1

or

not belongs to concept c

0



CONCEPT?

YES

MACHINE INTELLIGENCE

Concept Learning

- what is a **concept**?
- we will try to understand it practically

Concept

Data object

C

x

label

$X = \left\{ \begin{array}{l} \text{x- belongs to concept c} \\ \text{(oval, large, dark), (circular, small, light)} \end{array} \right\}$

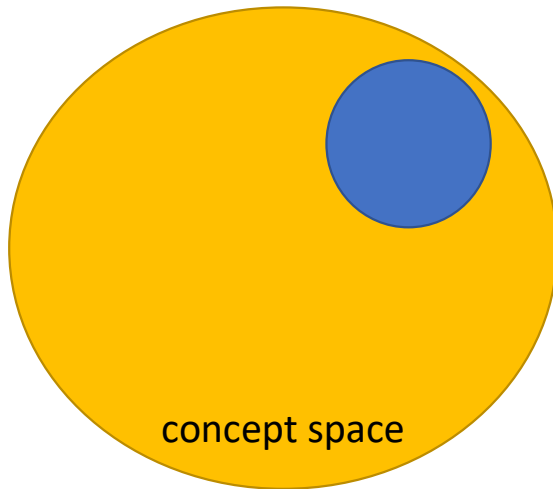
not belongs to concept c 0

each object x is defined by specific features
over which we make our decision. This is done down
through which we can decide if it belongs to the
concept or not. (8 attributes (shape, size, colour)
and once we know it we can label any new data
objects ...now!!!)
accordingly.....this the concept learning task

shape-oval, circular
size-large, small
colour-dark, light

MACHINE INTELLIGENCE

Concept Learning



How can I make a program that can learn to identify any 2⁸ concepts how???

Suppose our data is defined using d binary attributes then there are 2^d let us say it as b, then we have 2^b concepts

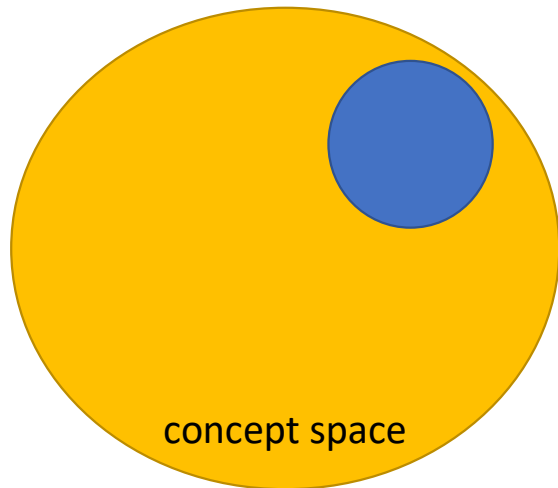
the learning task we need for a B attribute object space we have space with 256 going to look at

suppose our data is defined using d binary attributes then there are 2^d let us say it as b, then we have 2^b concepts

MACHINE INTELLIGENCE

Inductive bias

considerable number of objects belong to a particular kind or concept we can view as we will see that function which take conjunction (logical AND) of few of these basically a subset of all possible feature



$X_1 \rightarrow (1,0,1)$
 $X_2 \rightarrow (0,0,1)$

circular \wedge dark

the reason we call this concept is only if both of this is true for a data object then the object belongs to the concept

0	1
oval	circular
small	large
light	dark

MACHINE INTELLIGENCE

Inductive bias

___ ^ ___ ^ ___

circular ^ dark can be represented like this

circular ^ _?_ ^ dark

? means anything, that is any value there is true

? ^ ? ^ ? is a concept that accepts everything, we call it accept all

∅ is a concept that rejects everything, we call it reject all

since we have three attributes here we can mention our conjunctive concept something like this

MACHINE INTELLIGENCE

Inductive bias



for every position we can have three value that is two among the binary value and one ?

also we have one concept of \emptyset

so we have $(3 \times 3 \times 3) + 1 = 28$ conjunctive concepts

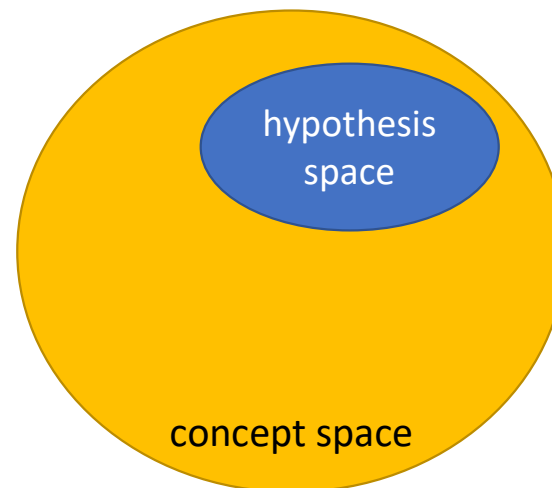
if u carefully observe we had total of 256

concepts in our concept space but now we have only 28 concepts in our conjunctive concept space

i.e we have shrunk our search space

this shrunk space is what we call as hypothesis space

Next operation we will take is how many then we will have 3 conjunctive concepts for a object with 3 binary attributes



MACHINE INTELLIGENCE

Find S algorithm

Concept- Days on which person enjoys sports

Attributes-

Sky-sunny,rainy

Temp-warm,cold

Humidity-Normal,High

Wind- strong,weak

Water-warm,cool

Forecast-same,change

Let's call this problem as finding a concept
first algorithm that is Find S algorithm

MACHINE INTELLIGENCE

Find S algorithm

1. start with $h=\emptyset$
2. use next input $\{x, c(x)\}$
3. if $c(x)=0$,go to step 2
4. $h \leftarrow h \wedge x$ (pairwise-and)
5. if more examples .Go to step 2
6. stop

Let us look at the pseudo code of this algorithm



Pairwise -and rules

$$a_h \wedge a_x = \begin{cases} a_x & : \text{if } a_h = \emptyset \\ a_x & : \text{if } a_h = a_x \\ ? & : \text{if } a_h \neq a_x \\ ? & : \text{if } a_h = ? \end{cases}$$

MACHINE INTELLIGENCE

Find S algorithm-Problem

step1: start with $h=\emptyset$

$h=\{\emptyset,\emptyset,\emptyset,\emptyset,\emptyset,\emptyset\}$

step2: use next input $\{x,c(x)\}$

$x=\{\text{sunny,warm,normal,strong,warm,same}\}$

$c(x)=\text{yes}==1$

step3: if $c(x)=0$, go to step 2

step4: $h \leftarrow h \wedge x$ (pairwise-and)

$h_0=h_0 \wedge x$

$h_0=\{\text{sunny,warm,normal,strong,warm,same}\}$

step5: if more examples, go to step 2

$h(0)=\text{sunny}$

$h(1)=\text{warm}$

$h(2)=\text{normal}$

$h(3)=\text{strong}$

$h(4)=\text{warm}$

$h(5)=\text{same}$

We will learn the concept of the step by step and rule we need to learn this concept with the training data below

sky	temp	humidity	wind	water	forecast	enjoy
sunny	warm	normal	strong	warm	same	yes
sunny	warm	high	strong	warm	same	yes
rainy	cold	high	strong	warm	change	no
sunny	warm	high	strong	cool	same	yes

$$a_h \wedge a_x = \begin{cases} a_x & : \text{if } a_h = \emptyset \\ a_x & : \text{if } a_h = a_x \\ ? & : \text{if } a_h \neq a_x \\ ? & : \text{if } a_h = ? \end{cases}$$

MACHINE INTELLIGENCE

Find S algorithm-Problem

step2: next input $\{x, c(x)\}$

$x = \{\text{sunny, warm, high, strong, warm, same}\}$
 $c(x) = \text{yes} = 1$

step3: if $c(x) = 0$, go to step 2

step4: $h \leftarrow h \wedge x$ (pairwise-and)

$h_0 = h_0 \wedge x$

$h_0 = \{\text{sunny, warm, ?, strong, warm, same}\}$

step5: if more examples

Go to step 2
 $h(0) = \text{sunny}$
 $h(1) = \text{warm}$
 $h(2) = ?$
 $h(3) = \text{strong}$
 $h(4) = \text{warm}$
 $h(5) = \text{same}$

sky	temp	humidity	wind	water	forecast	enjoy
sunny	warm	normal	strong	warm	same	yes
sunny	warm	high	strong	warm	same	yes
rainy	cold	high	strong	warm	change	no
sunny	warm	high	strong	cool	same	yes

$$a_h \wedge a_x = \begin{cases} a_x & \text{:if } a_h = \emptyset \\ a_x & \text{:if } a_h = a_x \\ ? & \text{:if } a_h \neq a_x \\ ? & \text{:if } a_h = ? \end{cases}$$

MACHINE INTELLIGENCE

Find S algorithm-Problem

step2: next input $\{x, c(x)\}$

$x = \{\text{rainy, cold, high, strong, warm, change}\}$
 $c(x) = \text{yes} == 0$

step3: if $c(x) = 0$, go to step 2

sky	temp	humidity	wind	water	forecast	enjoy
sunny	warm	normal	strong	warm	same	yes
sunny	warm	high	strong	warm	same	yes
rainy	cold	high	strong	warm	change	no
sunny	warm	high	strong	cool	same	yes

$$a_h \wedge a_x = \begin{cases} a_x & : \text{if } a_h = \emptyset \\ a_x & : \text{if } a_h = a_x \\ ? & : \text{if } a_h \neq a_x \\ ? & : \text{if } a_h = ? \end{cases}$$

MACHINE INTELLIGENCE

Find S algorithm-Problem

step2: next input $\{x, c(x)\}$

$x = \{\text{sunny, warm, high, strong, cool, same}\}$
 $c(x) = \text{yes} == 1$

step3: if $c(x) = 0$, go to step 2

step4: $h \leftarrow h \wedge x$ (pairwise-and)

$h_0 = h_0 \wedge x$

$h_0 = \{\text{sunny, warm, ?, strong, ?, same}\}$

step5: if more examples

step6: stop

Go to step 2
 $h(0) = \text{sunny}$
 $h(1) = \text{warm}$
 $h(2) = ?$
 $h(3) = \text{strong}$
 $h(4) = ?$
 $h(5) = \text{same}$

sky	temp	humidity	wind	water	forecast	enjoy
sunny	warm	normal	strong	warm	same	yes
sunny	warm	high	strong	warm	same	yes
rainy	cold	high	strong	warm	change	no
sunny	warm	high	strong	cool	same	yes

$$a_h \wedge a_x = \begin{cases} a_x & : \text{if } a_h = \emptyset \\ a_x & : \text{if } a_h = a_x \\ ? & : \text{if } a_h \neq a_x \\ ? & : \text{if } a_h = ? \end{cases}$$

MACHINE INTELLIGENCE

Find S algorithm-prediction

$C = \{\text{sunny, warm, ?, strong, ?, same}\}$

$x = \{\text{sunny, warm, high, strong, warm, same}\}$

$c(x) = 1$

suppose now we are given a new data set, we can pass the data through our concept and give our final hypothesis as our concept that predict the label. we use the concept learned to predict the labels that is if the person will play a sport or no

MACHINE INTELLIGENCE

Version space

a hypothesis is said to be consistent wrt to training data set if it classifies all the object of training data set to their corresponding

classes

<u>X</u>	<u>C(X)</u>
----------	-------------

x_1	$c(x_1)$
-------	----------

x_2	$c(x_2)$
-------	----------

so on.....

a hypothesis is said to be consistent if $h(x_i)=C(x_i)$

at any given point let H be our hypothesis space

so a version space is a subset of H

but all concepts in version space VS are

consistent wrt to training set

i.e $VS=\{h : h \in H \text{ and } h \text{ is consistent with } D_{\text{training}}\}$

this ensures our algorithm learns only the best hypothesis

to be consistent wrt to training data set, a hypothesis must correctly classify all the objects in the training data set. If a hypothesis fails to do so, it is not consistent. The version space is the set of all hypotheses that are consistent with the training data. It is a subset of the hypothesis space. The version space may be empty if the training data is not consistent. If the version space is not empty, it contains all the hypotheses that are consistent with the training data. The version space is a useful concept in machine learning because it allows us to focus on the hypotheses that are most likely to be the correct one.



THANK YOU

K.S.Srinivas
srinivasks@pes.edu
+91 80 2672 1983 Extn 701