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A computer program is said to learn from experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, unproves with Experience E. 3 examples are: -> Detecting tumor in scans: This is semantic segmentation, where each pixel in the unage is classified using CNN's. → Summarizing long documents: Its a branch of NUP that summarizes long text. → Olyect Detection: This is done using Image Segmentation using Deep Learning. Question2 The main challenges in Machine Learning is as follows: → Nonrepresentative training Datas It is necessary that the training data contains such samples as are needed for it to generalise. Otherwise it does not generalise well. If the sample is too small, we have sampling noise and large samples can be flamed. This is also called sampling -> Insufficient Quality of Training Data: It usually takes lots of data for Machine learning algorithms to work properly · Even for simple things a lot of examples are needed. Hence it is necessar to provide the HL algo with sufficient data. - Poor-Quality Data: Data that contains errors, outliers and noise are considered to be poor Quality data. Such data makes it difficult for ML algorithms to identify underlying patterns. Therefore it is necessary to clean up data before using for Machine hearning.

	elevant Features:
feat	IL algorithm performs well only if it gets goo
feat	essary to come up with good set of training ares. This is called feature ensineeding to the
eng	ineering involves:
Feat	Ture selection deature out to
neu	features. feature extraction and creating
Wes	litting at a t
wws	is when the model
data	hut does not generalize the testing data well happens if the model is too comblex: It as
ww	happens if the made is a
lie	happens if the model is too complex. It can reduced by simplifying the model
trail	sing data.
······································	terfitting of Data:
- VVV	is when the model is used to a
und	erlying pattern. This happens when the model is
too	simple to predict . It can be reduced by select
an	sore powerful model and feed better features.
Que	stion3
> 1/0 F	
e vu	ing Classifier:
su	before there are a few classifiers and each one vienes different accuracy. Subhand
";	
Rea	ression, SVM, Decision Tree and so on A better
Rea	get the results with more accuracy is to agare
Rea	get the results with more accuracy is to aggree results of all classifiers and bredict the
to the	get the results with more accuracy is to aggre results of all classifiers and phedict the class it get the most results . This is the band out
to the	get the results with more accuracy is to aggre results of all classifiers and phedict the class it get the most results. This is the hard-votes sifier. This methods gives a better accuracy.
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the the class site can	get the results with more accuracy is to aggre results of all classifiers and phedict the classifiers the most results. This is the hard-votes sifier. This methods gives a better accuracy agging:

he	sampled several times for same predictor ofter
this	sampled several times for same predictor After aggregation can be applied to get the results:
> Out	of Bag Evaluation:
Sov	ne vistances are sampled several times while
son	re instances may not be sampled at all. A
Bag	ging classifier by default samples only 'm'
atro	ining instances with replacement. Since the
res	t are never less to the prediction
1120	d for validation. This is called out-of-lox
81/0	luation.
Ques	tion4
Yh e	3 major types of ML algorithms are:
-> l	sherrised:
	He to inia data is labelled and the input
н.	1 a l a l a l a l a l l a l l a l l a l l a l l a
ine	alling it to make predictions on new data.
	where it is a second of the se
-> UW	superised:
	to the model learns patterns without guidance
aa	re the training algorithm learns with unlabelle ta. The model learns patterns without guidance nforcement:
	Deane to viteract with the
He	ironment by taking actions and learning on
em	dbacks by rewards or penalties goal is to
- fee	abacks any
ma	nimize rewards.
Que	stion5
	le la
_dn_	example of superised learning is classificated
An	example of unsuperment
cust	enal as span enal enal enal is using example of unsupervised learning is using oner purchase history data to group them
into	segments.
	The state of the s