Assignment-1 Semi- supereised learning

1) write a short note on seni-supervised

Ans) Seni supervised leavening is a markine leavening
capproach that his between supervised & unsupervised
leavening it leverages both labeled and unlabeled
data during training, making it porticulably useful
when acquiring labeled data is expressive on time—
courseming, but inhabeled data is abundant.

In semi supervised leavening, the model leavens from a
small amount of labeled data to improve its performance
The Key assemption is that the streeture of the data
distribution (seron unlabeled data) alice a well with

The Key assemption is that the stewarture of the data distribution (from unlabeled data) aligns well with the class labels (from labeled data).

Types of Seni superwised leavening Algorithms:

DSELF Teraining The model is initially trained on labelled

data then predicts could be for contabled data, which

are added to the training Let iteratively.

2) co-training: Two models are trained on disperent views of the data lace andel tabels the unlabeled data for the other, improving performal iteratively 3) brages based methods: These algorithms build a graph viewe nodes represent data points and edges respected similarities. The labels are peropagated

Through graph based on truse Connections.

Through graph based on truse models tower true joint

Downerative models: Thuse models tower true joint

peoplability distrebution of the data and true labels

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allowing them to infer labels for the unlabeled

data.

Advantages: «Pedius the cost of Obtaining Cabelled date
a con impriore model performance by utilizing unlabeled
data:

Challenges - Requires careful balancing of cabelled
and unlabeled obta to avoid propagating immorant
labels, «Assumptions about data distribution may

Application: text classification, Image knognition and speech pubulsing.

and speech pubulsing.

D'ine examples of Semi supervised Covening describing and i) self training: Text classification where a

not always hold.

small set of labeled documents is supplemented with pseudo labele generated from a large pool of unlabeled documents.

The co-teraining: Tmage classification where one nodel was color platimes and another ages texture features, each model helps labels the unlabeled images for the other.

The proph based method: Social network analysis

oure nodes represent users, and labels (guserinstouts)
are prograted therough the network
iv) (abel progration: Handwriting leelognition

Expoundion to stury samples with known labels prepagate information to stury samples.

Downerative models: bronzective adellaracial retwork (mans) where the discuminator can be used to label

unlabelled dota based on its leavered distribution.

vi) Consistency Regularization: Image classification

where augmentations of images leg sustations, translation

one used to ensure that predictions remain/ousistent.

vii) Pseudo leaving: speech recognition where a

model terained on a small set of transcribed aedio

Con generate pseudo corbels poro larger set of untrans-

- beided audio data.

3) weite a few applications of semi-supervised towning algorithms.

And Semi-supervised towning algorithms are applied in various domains where cabelled data is timeted but unlabelled data is abundant,

* Test classification: sentiment analysis topic sategoring atean, and span detection.

* Image classification: object subgrition, paid subgrition, and medical imaging.

* Speech Regionition: Transcription of speech language & voile Commonder.

nurb content classification: Categoriezing are pages on Jiltering Curb Contens: friend recommendation as orial Network Analysis: friend recommendation formunity detection, and inflience perogration and Anomaly detection: fraud detection, network received. * Video Analysis: Action relognation, Object texaction - Natural Canquage Professing (NLP): Named entite herognition, mainine teranslation, and question arrivering Assignment - 2 Decision tree. Develop the decision true for the following AXORB