Multilabel Classification Techniques

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- Introduction
 - What is Multilabel Classification?
- 2 Literature Survey
 - Available Algorithms
- Frame Work
 - Algorithms
 - Performance Measures

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What is Multilabel Data?

Introduction



Single label classification: Is this a picture of a beach? $\epsilon\{yes, no\}$

Multi label classification¹: Which labels are relevant to this picture?

 $\subseteq \{beach, foilage, field, mountain\}$

i.e. each example could belongs to more than one label So, for a given image we could express above problem into below format:

X	Beach	Foilage	Field	Mountain
Ex1	1	0	0	1

¹Multi-label classification by Jasse Read

Multilabel data

Types of Method

- 1. Problem Transformation Method²
- 2. Algorithm Adaptation Method

Project Focus Area

Problem Transformation Method

- Problem Transformation Method.
- Project work involves learning and implementing various algorithm to tackle given data and produce result to validate the algorithm.

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Literature Survey

- 1. Major Algorithm³:
 - Binary Relevance
 - a) Chain Classifier
 - b) Two Level classifier
 - RAkEL
 - a) Disjoint
 - b) Overlap
 - Label Power-set

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Algorithms - Examples



Ex1: Beach



Ex3: Beach + Mountain



Ex2: Foliage



Ex4: Field + Mountain

Algorithms - Binary Relevance

Creates L separate binary problems (L = no. of labels) Generate 4 models for our 4 label problem (considering a single label at a time) Beach Foliage Field Mountain Ex1 1 0 0 0 Fx2 0 1 0 0 1 Ex3 0 0 1 Ex4 0 0 1 1 Beach **Foliage** Field Mountain Ex1 1 Ex1 0 Ex1 0 Ex1 0

4 🗇 🕨	4 A	4 = 6	4 = 5	-	90 Q

Ex2

Ex3

Ex4

0

1

0

Ex2

Ex3

Ex4

0

0

Ex2

Ex3

Ex4

0

0

1

Ex2

Ex3

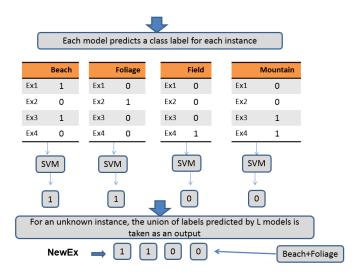
Ex4

0

1

1

Algorithms - Binary Relevance



Algorithms - Label Powerset

Here, every distinct labelset in the original multi-label data is considered as a new class



The multi-label problem is converted into a multiclass (single label) problem with 2^L possible class values (L = no. of labels)

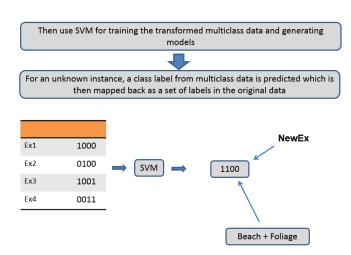
Foliage Field

Beach

Ex1	1	0	0	0
Ex2	0	1	0	0
Ex3	1	0	0	1
Ex4	0	0	1	1
	Ex1	10	00 🚄	_
	Ex2	01	00 👉	
	Ex3	10	01 ←	
	Ex4	00	11 ←	

Mountain

Algorithms - Label Powerset



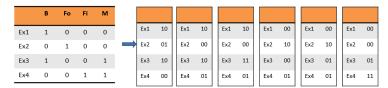
Algorithms - RAkEL

Randomly select a value of k (size of a label subset) and build m Label-Powerset classifiers



 2^k problems are considered at a time rather than 2^L problems like in LP (L = no. of labels)

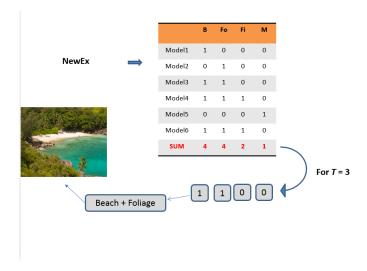
For k = 2 and L = 4 \implies m = 6 models



Algorithms - RAkEL

Learn and train LP classifiers with varying values of k and m at each iteration For an unknown instance, LP classifier predicts a label from the corresponding to k-labelset A total (or average) is calculated for each label in L Final label is assigned based on a certain Threshold value If the total (average) is greater than the threshold (T), then label is positive

Algorithms - RAkEL



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Performance Measures

Multilabel Algorithm performance is evaluated with below mentioned measures:

- 1. Accuracy
 - 1.1 Denotes proportion of Correctly predicted class to total number of class
- 2. Precision
 - 2.1 Denotes proportion of predicted correct labels to total number of Actual labels, averaged over all instances.
- 3. Recall
 - 3.1 Denotes proportion of predicted correct labels to total number of Predicted labels
- 4. F1-measure
 - 4.1 Harmonic mean of Precision and Recall.
- 5. Hamming Loss
 - 5.1 It is a loss function which calculates the proportion of misclassified labels to the total number of labels

Thank you

