

# Forest Cover Classification

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## I. INTRODUCTION

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TABLE I  
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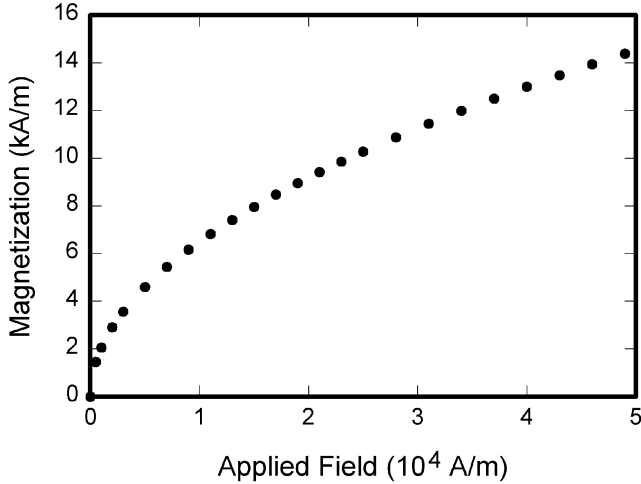


Fig. 1. Example of a figure caption.

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## II. INTRODUCTION

## III. BACKGROUND

### A. *k*-nearest Neighbours (*k*NN)

The nearest neighbour algorithm is an example of instance-based learning where the prediction for an instance is made by comparing the instance to similar training instances. This is a lazy learning approach since we delay processing training data until prediction is needed. Since *k*NN is also similarity-based, it is non-parametric, which means that it does not learn a mapping from the input space to the output space.

The *k*-nearest neighbour algorithm is presented in Algorithm 1 which explains that we calculate the distance from each data point in our training dataset to the distance. From there, we choose the closed *k* observations and perform inference. In a classification context we may perform majority voting to determine the class, in a regression context we use the mean or median of the *k* nearest observations.

In order to use the algorithm we need to define a value of *k* which decides the number of nearest neighbour instances to consider when predicting. Choosing a value a low value for *k* results in a high variance, but low bias leading to an unstable model which tends to overfit. Choosing a large value for *k* results in a smaller variance, but high bias, leading to a more stable model which tending to underfit. Measures such as *k*-fold cross validation assist in the selection of a value of *k*.

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**Algorithm 1** *k*-Nearest Neighbors Algorithm

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```
function KNN(D, x, k)  
  for all x' ∈ D do  
    d = distance(x, x')  
  end for  
  sort(d)  
  S = set of k patterns in D closest to x  
  return class as majority class in S  
end function
```

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**Notation:** *D* denotes the training dataset, *x* is the query instance, *k* is the number of nearest neighbours, and distance(·) is a distance metric (e.g., Euclidean distance).

## B. Classification Trees

Classification trees represent a fundamental approach in supervised machine learning for predictive modelling, where the learned model forms a hierarchical tree structure with non-terminal nodes representing decisions on descriptive features and terminal leaf nodes representing target feature predictions [?]. A classification tree is a type of decision tree where leaf nodes represent different discrete classes.

## C. Expectations With Respect to Data Quality Issues

### IV. METHODOLOGY

### V. EMPIRICAL PROCEDURE

### VI. RESEARCH RESULTS

### VII. CONCLUSION

### REFERENCES

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