

(CS454) Report for:  
*Solving TSP using Stochastic Optimization*

**Junghyun Lee\*** (20170500)

Dept. of Mathematical Sciences & School of Computing, KAIST

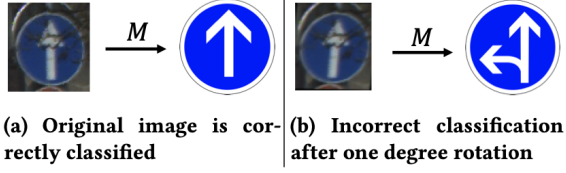
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\*jh\_lee00@kaist.ac.kr



(a) A motivating example from the GTSRB dataset

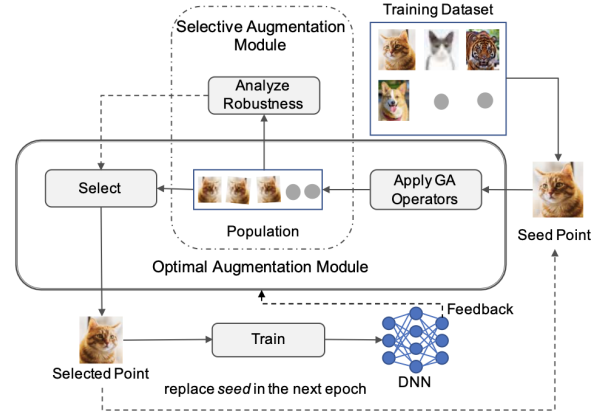


Figure 1: Figures from [1]

## 1 Background

## 2 Algorithm

### 2.1 k-Means Clustering

#### 2.1.1 Conjecture

Note that the number of clusters

### 2.2 Intracluster ACOs

### 2.3 Intercluster ACO

### 2.4 Combine the local solutions into a global solutions

### 2.5 3-Opt Algorithm

## 3 Experiments

To see the effectiveness of this algorithm, I've done several experiments

## 4 Implementation

## 5 Usage

## References

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