

Genome - SD1

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Outline

- Problem & Solution
- Our Progress
- Deep Learning & Genetic Algorithm
- Software Tools
- Implemented Algorithm
- Demo
- Results
- Limitations
- Future Works

Problem Being Addressed

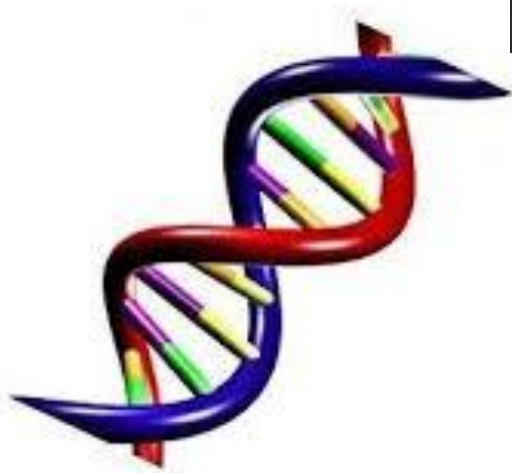
Finding optimal model parameters for Deep learning Convolutional Neural Networks (CNN) require:

- extensive manual testing
- expensive resources
- 40% of companies take >30 days



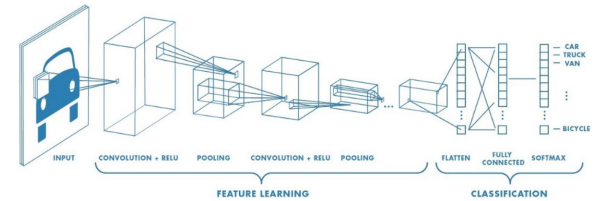
Proposed Solution

Integrate a Genetic Algorithm with Deep Learning to quickly find optimal model parameters for CNNs



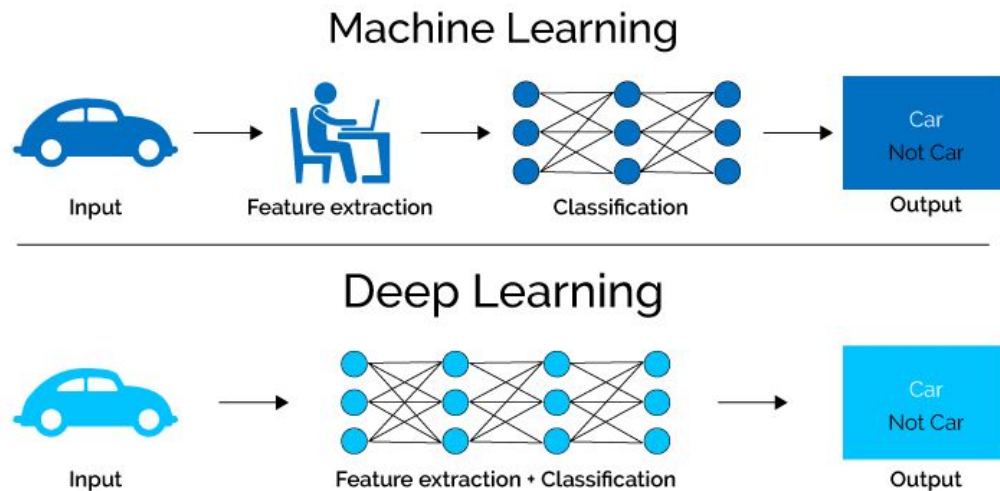
Our Progress

- Week 1 - 2: Genetic Algorithm
- Week 3 - 4: Deep Learning
- Week 5 - 6: Integration of GA with DL
- Week 7 - 9: Increasing complexity of Algorithm
- Week 10 - 11: Testing on Vermeer Data & Improving Algorithm



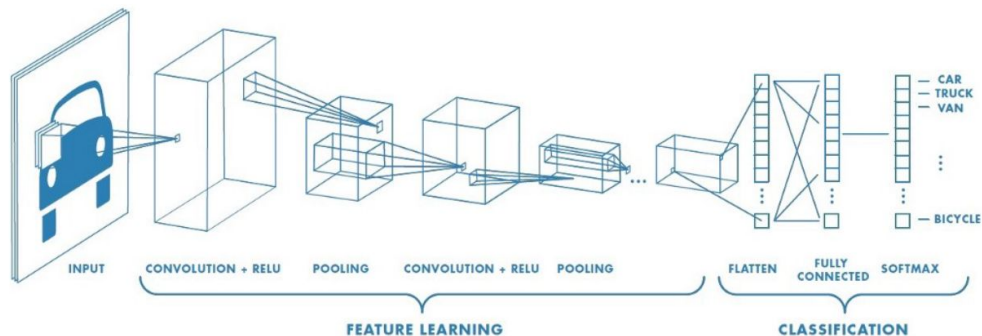
Deep Learning

- Artificial Neural Network
- Levels of neurons
- Performance improves with practice



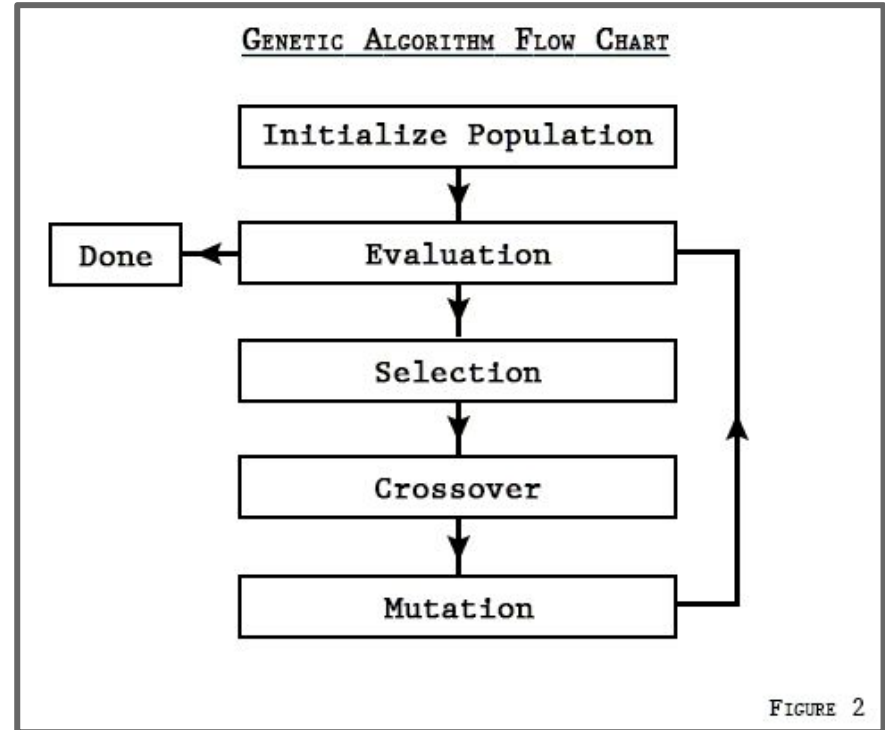
Deep Learning

- CNN
 - Classify Images
 - Built in TensorFlow



Genetic Algorithm

- Parameters
 - Population Size
 - Generation Count
 - Population Retained
- Variables
 - Batch Size
 - Number of Layers, Filters, Nodes
 - Etc.
- Fitness Function



Genetic Algorithm

- Process of Natural selection
- Crossover
- Mutation

Before Mutation

A5

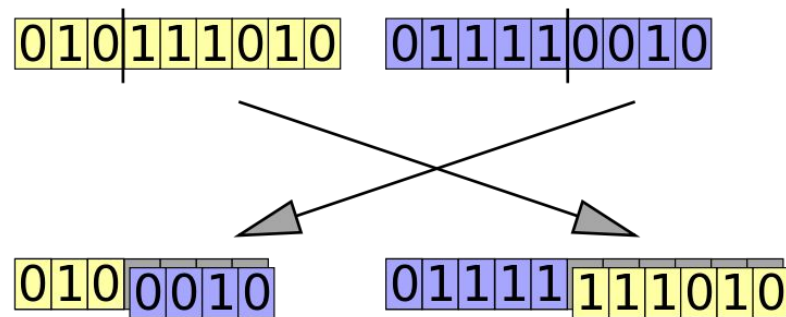
1	1	1	0	0	0
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After Mutation

A5

1	1	0	1	1	0
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Crossover example



Software / Tools

- Genetic Algorithm
- TensorFlow
- Keras
- CNN model architecture
- Python
- Jupyter notebook / IDE
- Colab notebook
- GCP VMs



- Optimization
- Classification (GBML)
- Human Comparable Design



Implemented Algorithm

- Implement a Genetic Algorithm
- Fitness Function
 - Integrate with Genetic Algorithm
 - Create baseline model
- Add variables for Genetic Algorithm to optimize
 - Batch Size, Activation Function, Optimizer
 - Layers - Filter, Node, Pool Size

```
BatchList = [32, 64, 128, 256]  
Optimizers = ['Adam', 'Nadam', 'Adadelata', 'Adamax']  
Activations = ['relu', 'tanh', 'selu', 'elu']
```

Input (fixed)

C: Convolutional Layer
P: Max Pooling

(Pair)

Flatten

Dense1()

Dense2()

Output(fixed)

Baseline model

Implemented Algorithm

- Variables

- Layers -

- Filter size, Number of Nodes, Pooling Size
- Number of Pairs & Number of layers in each Pair

Input (fixed)

C: Convolutional Layer
P: Max Pooling

(Pair)

Flatten

Dense1()

Dense2()

Output(fixed)

Baseline model

Input (fixed)

C: Convolutional Layer
C: Convolutional Layer
C: Convolutional Layer
P: Max Pooling

(Pair)

C: Convolutional Layer
P: Max Pooling

(Pair)

C: Convolutional Layer
C: Convolutional Layer
P: Max Pooling

(Pair)

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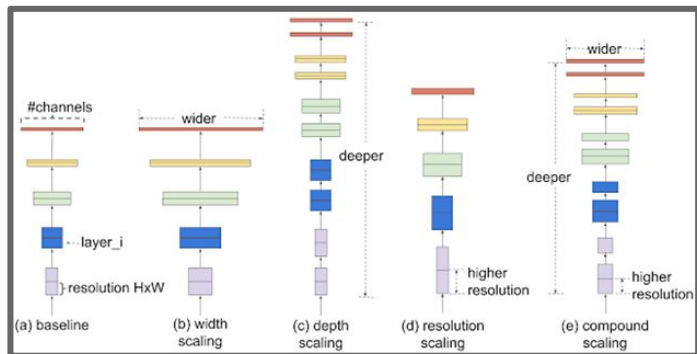
Flatten

Dense1()


Dense2()

Output(fixed)

Generated model

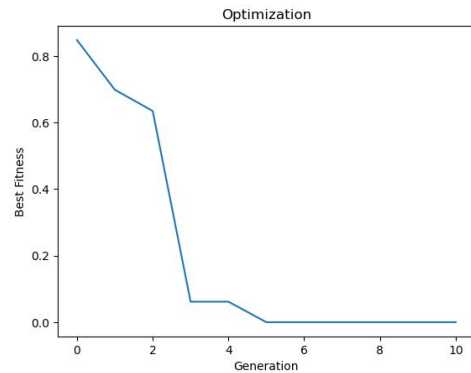
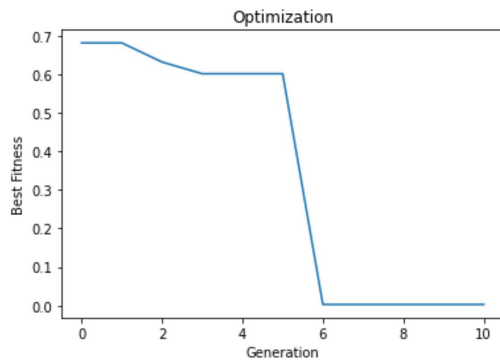


Demonstration

- Vermeer Data
 - Algorithm
 - GCP
 - Results
- 

Results

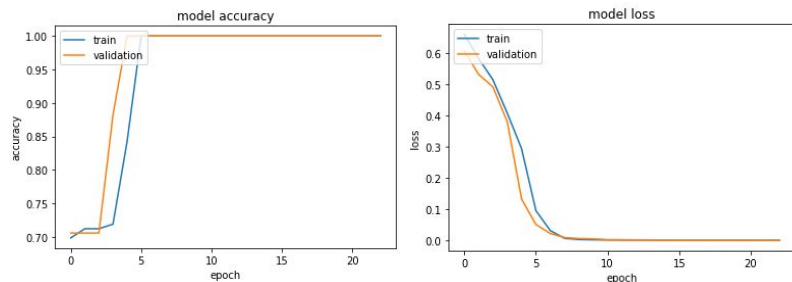
- Genetic Algorithm by nature reaches optimal solution:
 - Validation Loss gradually decreases
- Best Model & Parameters are saved
- Successfully works on Vermeer Data



Limitations

- Uses lots of memory
- Needs lots of compute
- More variables to add
 - Learning Rate, Dropout Layer, Patience for Early Stopping
- Tested algorithm on a few datasets for convergence

ResourceExhaustedError: OOM when allocating tensor with shape



Future Works

- Improvement:
 - Add more hyper parameter variables
 - Improve algorithm for better convergence
- Testing:
 - Test on new Datasets
 - Run with different population size, generations
- Compare results with other Research Algorithms
- Write a paper on the Algorithm

