Multi-objective Bayesian Optimization for Neural Architecture Search

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Motivation

Neural Architecture Search (NAS)

- hot research topic nowadays
- need for automatic design of neural architectures

Bayesian Optimisation

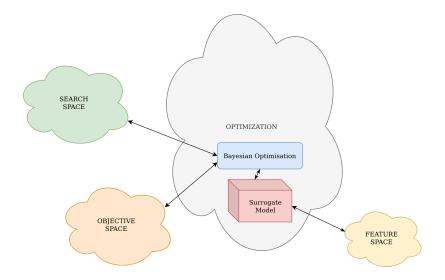
 suitable for problems with computationally expensive objective functions

Multi-objective Optimisation

• optimising not only network performance, but also size, energy consumption, etc.

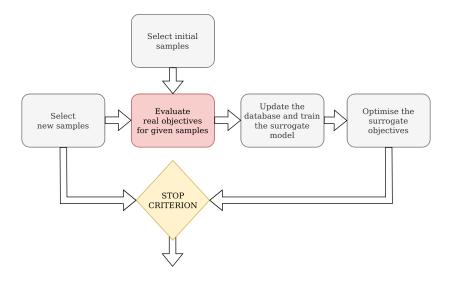
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Multi-objective Bayesian Optimization for NAS



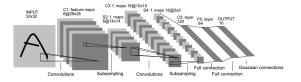
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Bayesian Optimization



Search Space

Feed-forward dense and convolutional networks



Individual encoding

individuals consists of two parts: convolutional and dense

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- each part consists of blocks corresponding to layers
- block is a tuple coding parameters of a layer

Search Space - Individual Encoding

Input Layer Image: 28x28x1 (width x height x channels)

Convolutional Layer 6 filters, kernel 5 x 5, sigmoid

Max Pooling Layer pool 2 x 2

Convolutional Layer 16 filters, kernel 5 x 5, sigmoid

Max Pooling Layer pool 2 x 2

Dense Layer 128 neurons, sigmoid

Dense Layer 64 neurons, sigmoid

Dense Layer 10 neurons, softmax

Code:

[[("Conv", 6, 5, "sigmoid"), ("Pool", 2), ("Conv", 16, 5, "sigmoid" ("Pool", 2)], [[128, "sigmoid",0.0), (64, "sigmoid", 0.0)]]

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Building Blocks

Optimising the surrogate function

- NSGAII algorithm
- using crossover on whole blocks, mutations modify parameters of blocks

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Surrogate model

- Gaussian processes with Matern kernel
- works on feature space

Feature space

- inputs for surrogate model, fixed size vector
- features describing the network

| N _A | The number of network parameters |
|---------------------|--|
| Ν | The number of convolutional layers |
| Ρ | The number of pooling layers |
| k | The mean size of the convolutional filter |
| a ^C M | Relative numbers of individual activations in convolutional part |
| | The number of dense layers |
| a ^D d | Relative numbers of individual activations in dense part |
| ď | The minimal, maximal, and mean dropout values |

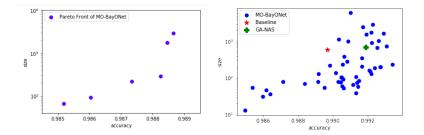
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Experiments

- Datasets: MNIST, fashion-MNIST, CIFAR10
- **Objectives:** cross-validation accuracy, network size (number of parameters)
- Iterations: 100
- Final metric: accuracy on test set (after 20 epochs)
- Compared to: small baseline model, solution found by GA

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Experimental Results



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Thank you! Questions?



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