# Neural Networks Using Genetic Algorithms

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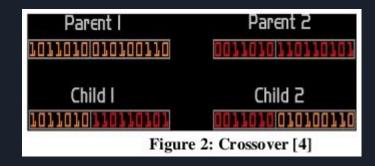
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# Introdução

- Combinar algoritmos genéticos com redes neurais leva às chamadas redes neurais evolucionárias
- Algoritmos genéticos podem ser usado no treinamento ou na escolha da topologia de redes neurais
  - Melhor forma de calcular os pesos
  - Melhor forma de interligar os neurônios da rede

# Algoritmos Genéticos

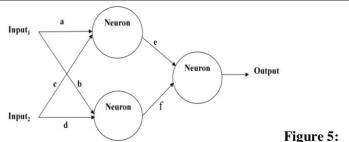
- População de strings binários relacionados e ranqueados pela função de "fitness"
- Métodos de seleção, crossover e mutação





## GA em Redes Neurais

No treinamento:



Simple Neural Network

All the weights in the network are joined to make one string. This string is then used in the GA as a member of the population. Each string represents the weights of a complete network.

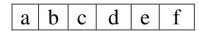


Figure 6: String or Chromosome

Figure 6 depicts the value of chromosome obtained from simple neural network as figure 5.

#### b) How to evaluate Fitness

Fitness is measured by calculating the error (target – output) (i.e. fitness= 1/error) - the lower the error the higher the fitness.

Os pesos podem ser evoluídos pelos métodos genéticos para se adequar melhor ao fitness

### GA em Redes Neurais

Na topologia:

Consider a simple neuron network. If there is a connection of one neuron with other neuron, it will be represented by 1 otherwise 0.

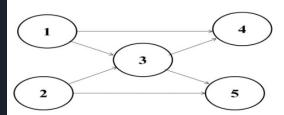


Figure 7: Simple Neural Network

From figure 7, consider the connections from neuron 1. These may be represented by the string shown below:

00110

The first zero represents the fact that neuron 1 is not connected to itself. The second zero means that neuron 1 is not connected to neuron 2. The third digit, which is 1, means that neuron 1 is connected to neuron 3; and so on.

The complete network may be represented by the matrix shown in figure 8.

0 0 1 1 0 Neuron 1 0 0 1 0 1 Neuron 2 0 0 0 1 1 Neuron 3 0 0 0 0 0 Neuron 4 0 0 0 0 0 Neuron 5 A matriz de conexão pode ser convertida em um string, sendo modificado e testando diferentes topologias para maximizar o fitness

## Conclusão

- 1. GA helps to generate better population from good parents, these results close to global optimum.
- 2. Important character of GA, it is robust.
- 3. They works well in various fields as:
  - a) In pattern matching
  - b) Speech recognition, text-to-speech
  - c) Machines that are able learn
  - d) Optical character recognition (OCR)
  - e) Fraudulent credit card detection (VISA)f) Image compression

#### 6.2 Cons:

- 1. It remains a 'black box' which once fed with inputs produces an output. However, their excellent result record might compensate for that deficiency.
- 2. A second drawback is that inputs have to be altered before being fed to the network.
- 3. It is fail to depict followings:
  - a) Which network (architecture) to use?
  - b) How many hidden layers?
  - c) How many neurons?
  - d) What activation functions should I use?
  - e) What cost function is the most appropriate?
  - f) Which training algorithm to apply?