



# Training Neural Networks with GA Hybrid Algorithms

ANNs

Algorithms

Experiments

Conclusions &  
Future Work

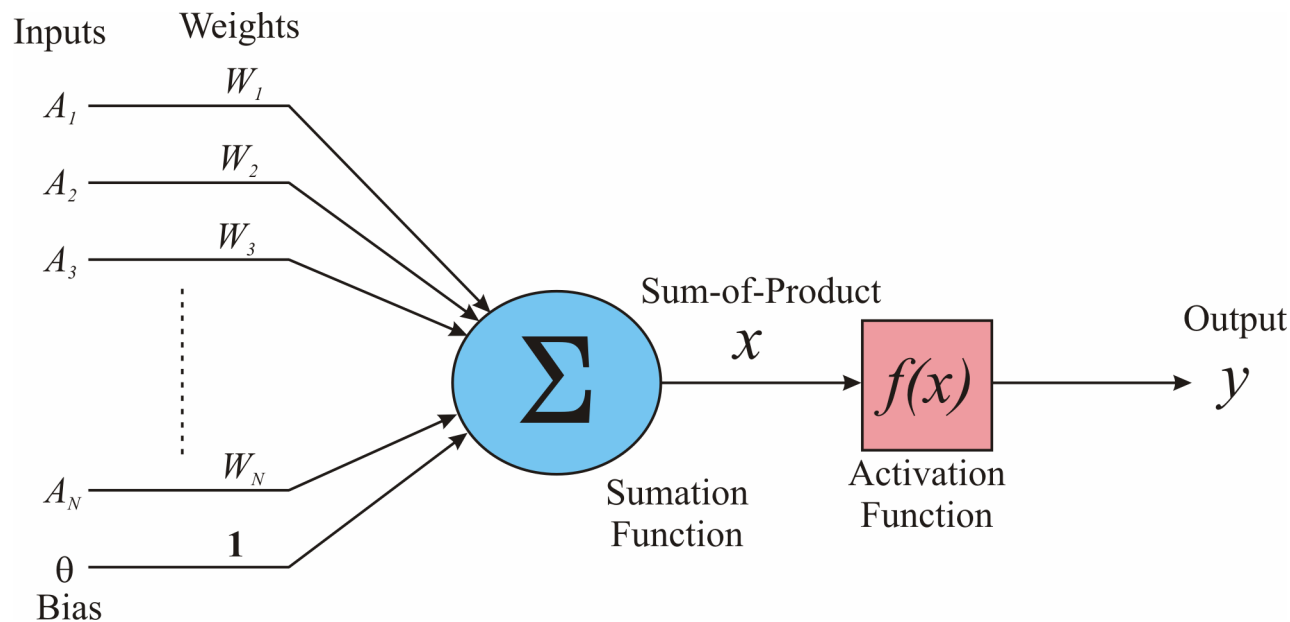


Lenguajes y Ciencias  
de la Computación

Enrique Alba and J. Francisco Chicano

# Artificial Neural Networks

- An ANN is a structured pool of **Artificial Neurons**:



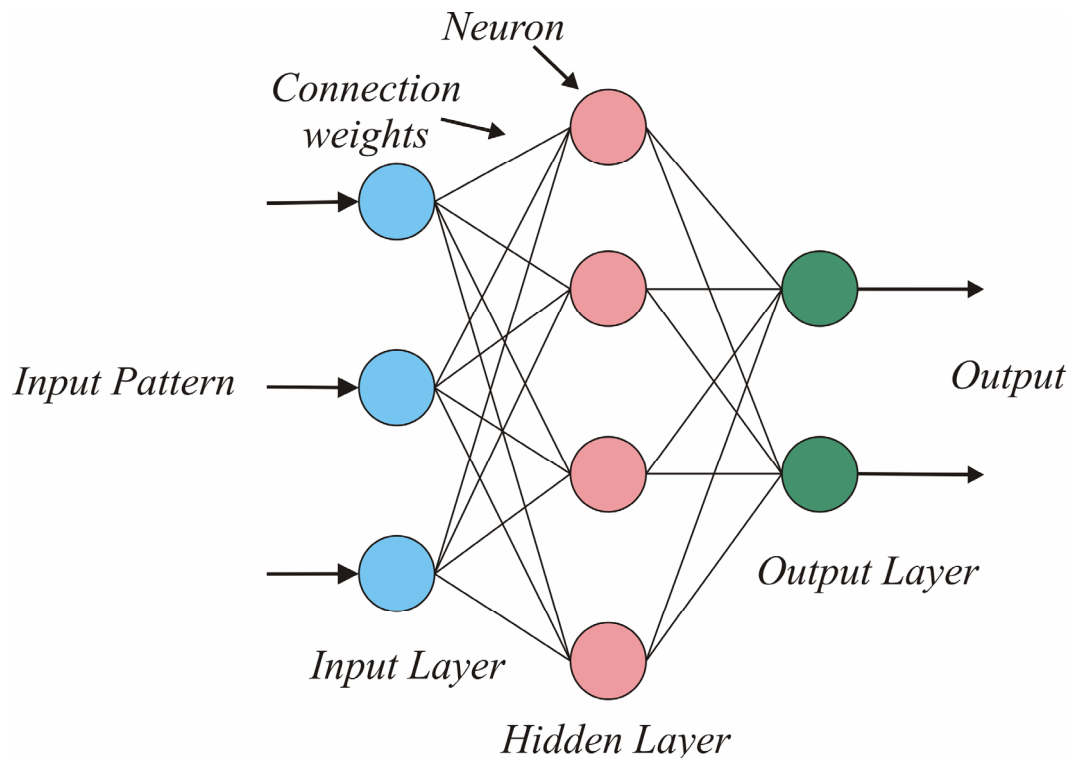
- The architecture determines how the neurons are connected:
  - **Feedforward networks (used here)**
  - **Recurrent networks**





# Multilayer Perceptron

- We use **Multilayer Perceptrons**



- **Parameters: number of layers, neurons per layer, activating functions**

# Training



- The **training process** consists in adjusting the network weights:

- **Supervised learning**
- **Non-supervised learning**

- **Supervised training process** consists in adjusting the weights to get the desired output for the present input patterns

- To measure de quality of the network several options exist:

- **SEP**

$$SEP = 100 \cdot \frac{o_{max} - o_{min}}{P \cdot S} \sum_{p=1}^P \sum_{i=1}^S (t_i^p - o_i^p)^2$$

- **CEP: percentage of incorrectly classified patterns**

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# Backpropagation



- BP is a **gradient-based** method

$$E = \sum_{p=1}^P \sum_{i=1}^S (t_i^p - o_i^p)^2$$

$$\Delta \mathbf{w} = -\underset{\substack{\text{learning rate} \\ \downarrow}}{\eta} \nabla E$$

```

InitializeWeights;
while not StopCriterion do
  for all i,j do
     $w_{ij} := w_{ij} - \eta \frac{\partial E}{\partial w_{ij}};$ 
  endfor;
endwhile;

```

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# Levenberg-Marquardt

- LM is a “quasi” second order method

$$\Delta \mathbf{w} = - \left[ \begin{array}{c} \downarrow \\ \mu I + \sum_{p=1}^P \overset{\text{Jacobian}}{J^p(\mathbf{w})^T} \overset{\text{Jacobian}}{J^p(\mathbf{w})} \end{array} \right]^{-1} \nabla E(\mathbf{w})$$

```

InitializeWeights;
while not StopCriterion do
  Calculates ep(w) for each pattern;
  e1 := ∑p=1P ep(w)T ep(w);
  Calculates Jp(w) for each pattern;
  repeat
    Calculates Δw;
    e2 := ∑p=1P ep(w + Δw)T ep(w + Δw);
    if (e1 <= e2) then
      μ := μ * β;
    endif;
  until (e2 < e1);
  μ := μ / β;
  w := w + Δw;
endwhile;

```

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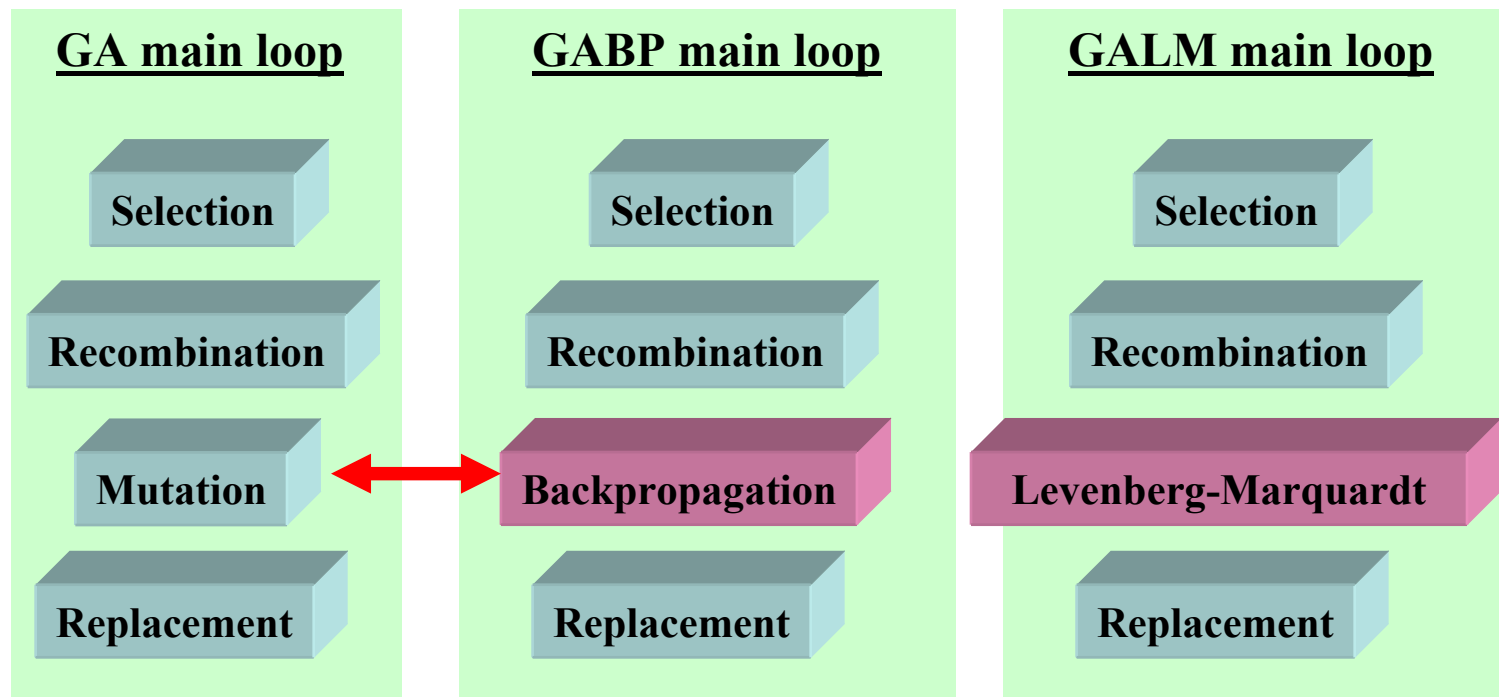
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# Hybrid Algorithms

- **Hybridization:** Inclusion of problem knowledge into the algorithm
- **Two possible classes of hybrid algorithms:**
  - **Strong:** Specific representation and operators
  - **Weak:** Combination of several algorithms (cooperation)



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# Experiments



- Three **classification** problems from **PROBEN1** benchmark

	Inputs	Outputs	Patterns
Cancer	9	2	699
Diabetes	8	2	768
Heart	35	2	920

- **Multilayer perceptrons** with **one output neuron per class**
- **Six neurons** in the hidden layer
- **Sigmoid** activating function
- In GA and hybrids, **fitness** = the inverse of SEP
- Pattern sets: **Training (75%)** and **Test (25%)**

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Results

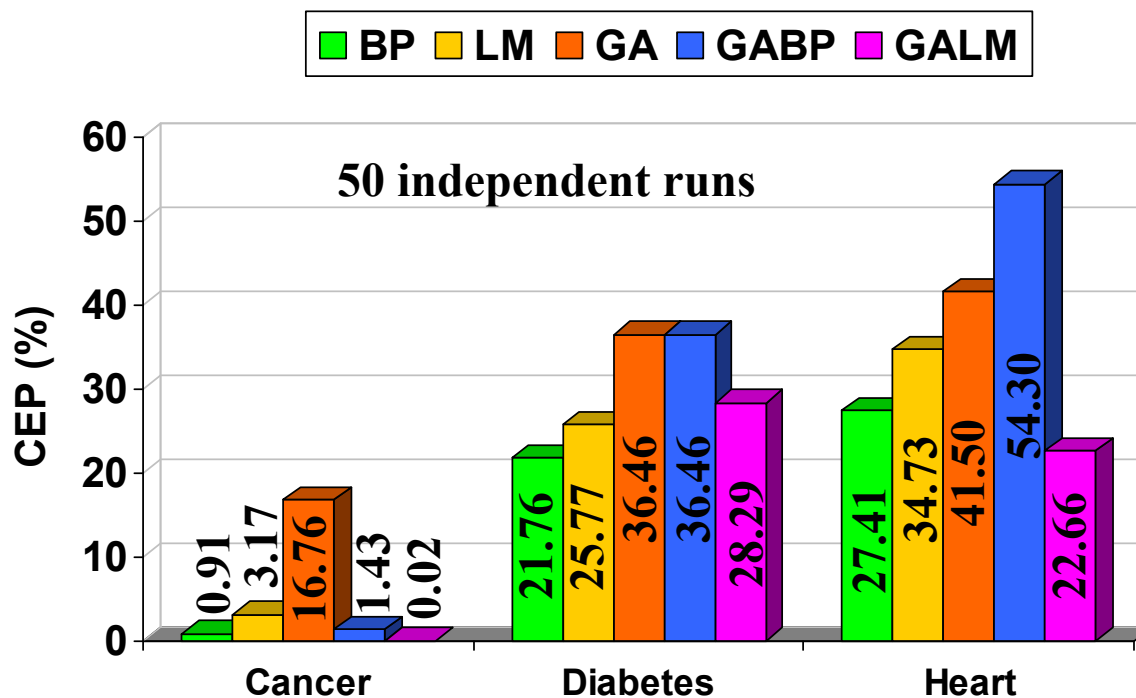
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# Results



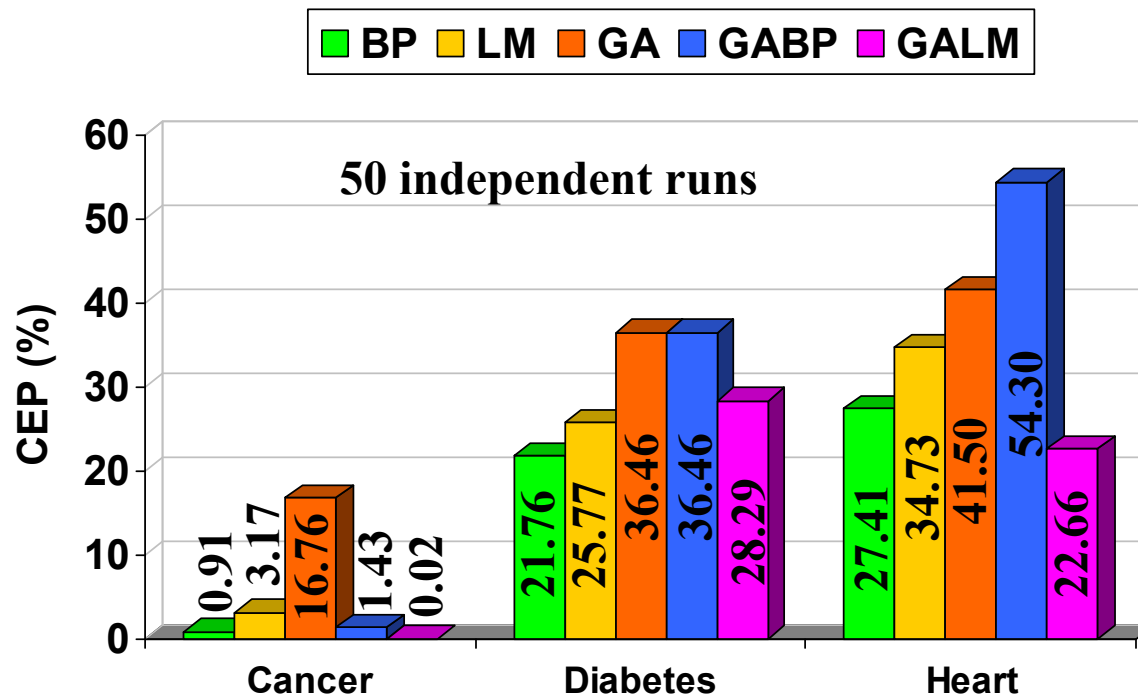
- GA gets higher CEP than BP and LM (**expected result**)
- BP more accurate than LM
- GALM more accurate than GABP



# Results



- GA gets higher CEP than BP and LM (expected result)
  - BP more accurate than LM
  - GALM more accurate than GABP
- unexpected behavior!

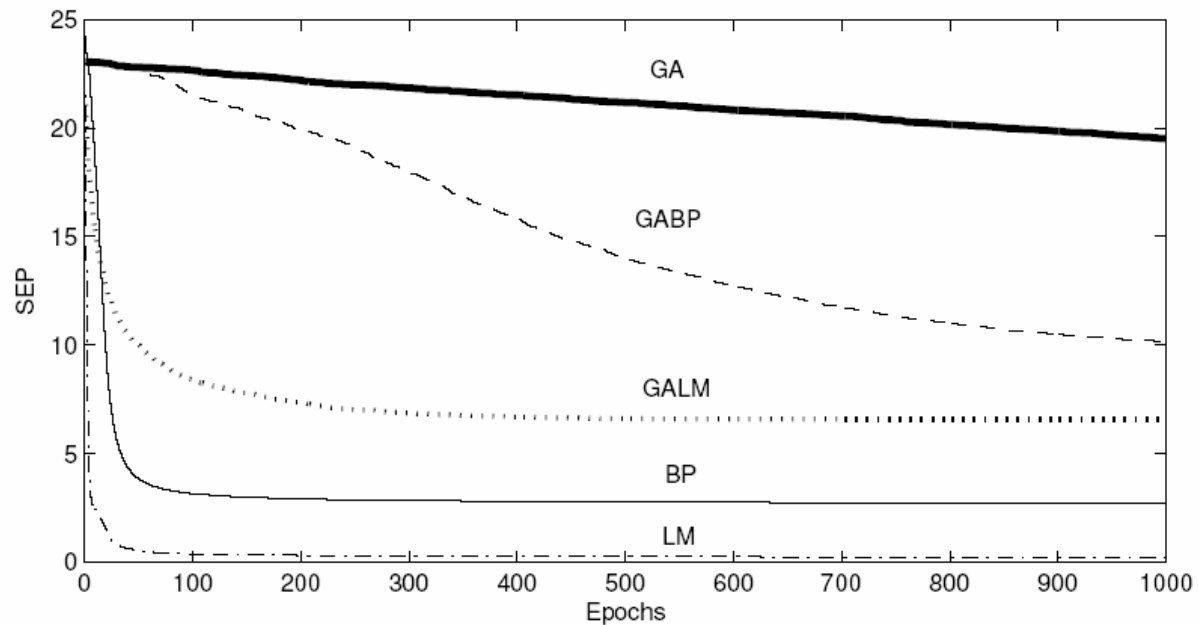


# Results



- LM is **the fastest** algorithm in minimizing the SEP
- GA evolution is **accelerated** with the inclusion of BP and LM
- BP and LM **over-train** the ANN to the used pattern set

50 independent runs





# Conclusions & Future Work

## Conclusions

- **BP** and **LM** outperform GA
- **Hybrid algorithms** outperform pure ones
- **GALM** hybrid found the best CEP (known to our knowledge) for **Cancer** (0.02%) and **Heart** (22.66%)

## Future Work

- Study other **algorithms** for these problems (ES, ESBP, ESLM)
- Solve additional **instances** (Gene, Soybean, Thyroid, ...)

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# THE END



## Thanks for your attention !!!

