

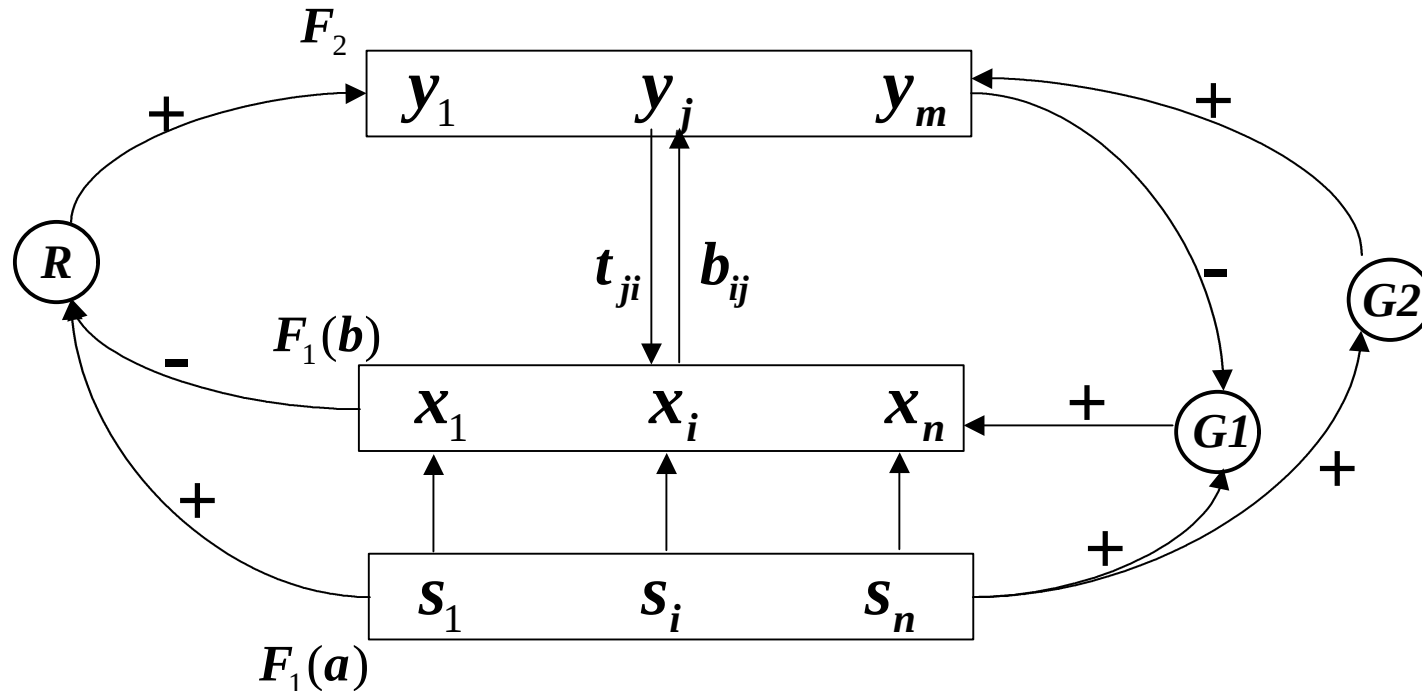
Chapter 5. Adaptive Resonance Theory (ART)

- **ART1**: for binary patterns; **ART2**: for continuous patterns
- Motivations: Previous methods have the following problem:
 1. Training is non-incremental:
 - with a fixed set of samples,
 - adding new samples often requires re-train the network with the enlarged training set until a new stable state is reached.
 2. Number of class nodes is pre-determined and fixed.
 - Under- and over- classification may result from training
 - No way to add a new class node (unless there is a free class node happens to be close to the new input).
 - Any new input x has to be classified into one of an existing classes (causing one to win), no matter how far away x is from the winner. no control of the degree of similarity.

- Ideas of ART model:
 - suppose the input samples have been appropriately classified into k clusters (say by competitive learning).
 - each weight vector $W_{\bullet,j}$ is a representative (average) of all samples in that cluster.
 - when a new input vector x arrives
 1. Find the winner j among all k cluster nodes
 2. Compare $W_{\bullet,j}$ with x
 - if they are sufficiently similar (x resonates with class j),
then update $W_{\bullet,j}$ based on $x - W_{\bullet,j}$
 - else, find/create a free class node and make x as its first member.

- To achieve these, we need:
 - a mechanism for testing and determining similarity.
 - a control for finding/creating new class nodes.
 - need to have all operations implemented by units of local computation.

ART1 Architecture



$F_1(a)$: input units

$F_1(b)$: interface units

F_2 : cluster units

$F_1(a)$ to $F_1(b)$: pair - wise connection
 between F_2 and $F_1(b)$: full connection

b_{ij} : bottom up weights
 from x_i to y_j (real value)

t_{ji} : top down weights
 from y_j to x_i (representing class j
 binary/bipolar)

$R, G1, G2$: control units

- F_2 cluster units: competitive, receive input vector $\mathbf{x}$ through weights $\mathbf{b}$: to determine winner $\mathbf{j}$.
- $F_1(\mathbf{a})$ input units: placeholder or external inputs
- $F_1(\mathbf{b})$ interface units:
 - pass $\mathbf{s}$ to $\mathbf{x}$ as input vector for classification by F_2
 - compare $\mathbf{x}$ and $\mathbf{t}_j$. (projection from winner $\mathbf{y}_j$)
 - controlled by gain control unit $\mathbf{G1}$
- Nodes in both $F_1(\mathbf{b})$ and F_2 obey 2/3 rule (output 1 if two of the three inputs are 1)

G_1

Input to $F_1(\mathbf{b}) : \mathbf{s}_i, \mathbf{G1}, \mathbf{t}_{ji}$ Input to $F_2 : \mathbf{t}_{ji}, \mathbf{G2}, \mathbf{R}$
- Needs to sequence the three phases (by control units $\mathbf{G1}$, $\mathbf{G2}$, and $\mathbf{R}$)

$$\mathbf{G}_1 = \begin{cases} 1 & \text{if } \mathbf{s} \neq 0 \text{ and } \mathbf{y} = 0 \\ 0 & \text{otherwise} \end{cases}$$

$\mathbf{G}_1 = 1$: $\mathbf{F}_1(\mathbf{b})$ open to receive $\mathbf{s} \neq 0$

$\mathbf{G}_1 = 0$: $\mathbf{F}_1(\mathbf{b})$ open for $\mathbf{t}_{J\bullet}$.

$$\mathbf{G}_2 = \begin{cases} 1 & \text{if } \mathbf{s} \neq 0 \\ 0 & \text{otherwise} \end{cases}$$

$\mathbf{G}_2 = 1$ signals the start of a new classification for a new input

$$\mathbf{R} = \begin{cases} 0 & \text{if } \frac{\|\mathbf{x}\|}{\|\mathbf{s}\|} \geq r \\ 1 & \text{otherwise} \end{cases}$$

$0 < r < 1$ vigilance parameter

$\mathbf{R} = 0$: resonance occurs, update $\mathbf{b}_{\bullet J}$ and $\mathbf{t}_{J\bullet}$.

$\mathbf{R} = 1$: fails similarity test, inhibits $\mathbf{J}$ from further computation

Working of ART1

- Initial state: nodes on $F_1(\mathbf{b})$ and F_2 set to zeros
- Recognition phase:** determine the winner cluster for input s
 - $s \neq 0$ is applied to $F_1(\mathbf{a})$ and stay there (clamped)

$$\mathbf{G}_1 = 1 \quad (\because y = 0, s \neq 0)$$

$$\mathbf{G}_2 = 1 \quad (\because s \neq 0)$$
 - $F_1(\mathbf{b})$ is open to receive s

$$\mathbf{R} = 0 \quad (|\mathbf{x}|/|\mathbf{s}| = 1 > r)$$
 - F_2 is open to receive $\mathbf{x} \bullet \mathbf{b}_{\bullet j}$
 - determine winner J
 - $\mathbf{x}$ is tentatively classified to cluster J

$$\mathbf{Y}_J = 1 \quad \mathbf{Y}_{k \neq J} = 0$$

- **Comparison phase:**
 - $t_{j\bullet}$ is sent down from F_2
 - $G_1 = 0$ ($y \neq 0$)
 - new $\mathbf{x}$ appears on $F_1(\mathbf{b})$: $\mathbf{x}_i = \mathbf{s}_i \wedge \mathbf{t}_{ji}$ (2/3 rule)
 - if $\|\mathbf{x}\|/\|\mathbf{s}\| \geq r$ $\mathbf{R} = 0$
 - a resonance occurs, the classification is accepted
 - if $\|\mathbf{x}\|/\|\mathbf{s}\| < r$ $\mathbf{R} = 1$
 - classification is rejected
 - reset signal is sent (from $\mathbf{R}$ to F_2)
 - y_j is permanently disabled
 - all other y_k is set to zero
 - goes back to recognition phase
 - search for other possible match

- ***Weight update/adaptive phase***

- Initial weight: (no bias)

bottom up: $0 < \mathbf{b}_{ij}(0) < L / (L - 1 + n)$ (L usually 2)

top down: $\mathbf{t}_{ji}(0) = 1$

- When a resonance occurs with $\mathbf{y}_J$, update $\mathbf{b}_{\bullet J}$ and $\mathbf{t}_{J\bullet}$.

$\mathbf{s}$: current input (on $F_1(\mathbf{a})$)

$\mathbf{x}$: comparison result between $\mathbf{s}$ and $\mathbf{t}_{J\bullet}$ (on $F_1(\mathbf{b})$)

$$\mathbf{x}_i = \mathbf{s}_i \cdot \mathbf{t}_{Ji}$$

$$\text{new } \mathbf{t}_{Ji} = \mathbf{x}_i \quad \text{new } \mathbf{b}_{ij} = \frac{L \mathbf{x}_i}{L - 1 + \|\mathbf{x}\|}$$

- If k sample patterns are clustered to node $\mathbf{y}_J$, then

$\mathbf{t}_{\bullet J}$ = pattern whose 1's are common to all these k samples

$$\mathbf{t}_{\bullet J} = \mathbf{s}(1) \wedge \mathbf{s}(2) \dots \mathbf{s}(k)$$

$$\mathbf{b}_{\bullet J}(\text{new}) \neq 0 \text{ iff } \mathbf{x}_i \neq 0 \text{ only if } \mathbf{s}_i \neq 0$$

$$\mathbf{t}_{J\bullet} = \mathbf{x}, \mathbf{b}_{\bullet J} \text{ is a normalized } \mathbf{t}_{\bullet J}$$

- Winner may shift:

$$L = 2, \mathbf{b}_{ij}(0) = 0.2$$

$$\mathbf{s}(1) = (1 \ 1 \ 1 \ 0) \quad \mathbf{t}_{1\bullet} = (1 \ 1 \ 1 \ 0)$$

$$\mathbf{s}(2) = (1 \ 1 \ 0 \ 0) \quad \mathbf{t}_{1\bullet} = (1 \ 1 \ 0 \ 0)$$

$$\mathbf{s}(3) = (1 \ 0 \ 0 \ 0) \quad \mathbf{t}_{1\bullet} = (1 \ 0 \ 0 \ 0)$$

$$\mathbf{s}(1) = (1 \ 1 \ 1 \ 0) \quad \frac{\|\mathbf{x}\|}{\|\mathbf{s}\|} = \frac{1}{3} < r$$

- What to do when failed to classify into any existing cluster?
 - report failure/treat as outlier
 - add a new cluster node $\mathbf{y}_{m+1}$ with

$$\mathbf{b}_{i,m+1} = \frac{L}{L-1+n}, \quad \mathbf{t}_{m+1,i} = 1$$

Notes

1. Classification as a search process
2. No two classes have the same ***b*** and ***t***
3. Different ordering of sample input presentations may result in different classification.
4. Increase of ρ increases # of classes learned, and decreases the average class size.
5. Classification may shift during search, will reach stability eventually.
6. ART2 is the same in spirit but different in details.