

## Neuronale Netze (SS 2002), 22.4.

### Perceptron for non linearly separable data (continued):

- **Hitting set problem:** Given a set  $S = \{s_1, \dots, s_n\}$ , a set of subsets  $C = \{c_1, \dots, c_m\}$  with  $c_i \subset S$ ,  $k \in \mathbb{N}$ , does there exist  $k$  points in  $S$  such that every  $c_i$  is hit by at least one of the  $k$  points

(In real life: Find a fixed number of representatives such that each group of interest is covered.)

We have shown that the problem is NP-complete via reduction from SAT. This holds even under the restriction  $|c_1| = |c_2| = \dots$

- **Perceptron training in the presence of errors:** Given an input dimension  $n$ , a training set  $P \subset \{0, 1\}^n \times \{0, 1\}$ , and  $k \in \mathbb{N}$ . Decide whether a perceptron can be found which has at most  $k$  errors on  $P$ .

We have shown that this problem is NP-complete via a (complicated) reduction from the hitting set problem.

$\Rightarrow$  optimum training of a perceptron is not that easy in principle!

**Rosenblatt-Perceptron** – an easy extension of the perceptron such that we can still use the perceptron learning algorithm for training:

- The Rosenblatt-perceptron computes a function

$$f : \{0, 1\}^{n \times m} \rightarrow \{0, 1\}, \vec{x} \mapsto p(f_1(\vec{x}), \dots, f_M(\vec{x}))$$

where  $p$  is a standard perceptron and  $f_i$  are **masks/receptors** implementing some function.

Idea: The  $f_i$  implement local problem dependent feature extraction, e.g. smoothing, line extraction,  $\dots$ . The  $f_i$  are fixed, only  $p$  is trained accordingly. The hope is, that for most realistic functions  $f$  simple standard receptors  $f_i$  can be found such that the problem becomes linearly separable.

- If there is no restriction on the receptors  $f_i$ , every function  $f$  can be implemented with a Rosenblatt-perceptron – just choose  $f_1 = f$ .

- A receptor  $f_i$  has a limited diameter  $k$ , iff indices  $(j, l)$  exist such that  $f_i$  depends only on the input coefficients

$$\begin{array}{ccc}
 x_{(j,l)} & \dots & x_{(j,l+k-1)} \\
 \vdots & & \vdots \\
 x_{(j+k-1,l)} & \dots & x_{(j+k-1,l+k-1)}
 \end{array}$$

- Hope: standard functions  $f$  can be made linearly separable only applying diameter-restricted receptors.

Note: applying some filters for preprocessing in vision (smoothing, line-extraction, ...) corresponds to diameter-restricted masks. This is part of the standard preprocessing for real problems in vision – although the problems become unfortunately not linearly separable.

- A Rosenblatt-perceptron with diameter-restricted receptors cannot recognize connected patterns!

⇒ Nice try, but unfortunately not enough!