Goal: A brief overview unsupervised learning and multi-layer perceptrons (MLPs).

MLP: Perceptron, weights, non-linear activation function, feed-forward topology.

Let X be the stimuli space and Y be the (discrete) set of class labels.

Supervised Learning Problem: To classify<sup>1</sup> stimuli based on example stimulus-class pairs.

- Input: labeled training examples  $\{(x_1, y_1), ..., (x_n, y_n)\}$  drawn from P(x, y), and unlabeled stimuli  $\{x_{n+1}, ...\}$  to be classified.
- Output: Best-guesses for class labels  $\{y_{n+1},...\}$ , or more generally estimates of the probabilities  $P(y|x_{n+1})$ , etc.
- Evaluation: Mean-square error of predicted to actual class labels, or expected perplexity.
- Example: A child learning sees images  $x_i$  and is told by their parent the name  $y_i$  of that type of object.
- Implementation: An MLP with any number of layers, x presented to first layer, weights trained via back-propagation of error between output layer and y.

Unsupervised Learning Problem: To learn class structure from stimuli.

- Input: unlabeled examples  $\{x_1, ..., x_n\}$  drawn from marginal P(x).
- Output: Inferred class structure Y and or more estimates of the probabilities  $P(y|x_i)$ .
- Evaluation: Mutual information between learned class structure and actual structure.
- Example: A feral child sees images  $x_i$  but has no parent to tell them the names of things.
- Implementation: An MLP with each layer trained to satisfy suitable intrinsic statistical criteria (such as fidelity and sparsity). Methods include sparse autoencoders, restricted Boltzmann machines, and perhaps a multi-neuron version of Chklovskii's rule.

Semisupervised Learning Problem: To classify stimuli given many unlabeled examples and limited labeled examples.

- **Input:** That of both supervised and unsupervised cases.
- Output: Same as supervised case.
- Evaluation: Same as supervised case.
- Example: A child who sees many images only some of which are named for them.
- Implementation: The MLP of unsupervised case plus a final supervised stage.

<sup>&</sup>lt;sup>1</sup>This discussion is in terms of classification, but the concepts can be easily extended to regression, reenforcement learning, and other frameworks.