ML

- · Supervised learning: prediction based on examples of correct behavior
- Unsupervised learning: no explicit target, only data, goal to model/discover
- · Semi-supervised learning: sup-

- plement limited annotations with unsupervised learning
- · Active learning: learn to query the examples actually needed for learning
- Transfer learning: how to apply what you have learned from A to B

Linear Regression Linear Perception Algorithm without

for t = 1, ..., T do **do**

· Reinforcement learning: learning to act, not just predict; goal to optimize the consequences of actions

Initialize $\theta \leftarrow 0(vector)$

```
for i = 1,...,n then do
         if y^i * (\theta * x^{(i)}) \le 0 then
             update \theta = \theta + y^{(i)}x^{(i)}
         end if
    end for
end for
```

Linear Perception Algorithm with Off-

Initialize $\theta \leftarrow 0(vector); \theta_0 \leftarrow 0(scalar)$

```
for t = 1,..., T do do
    for i = 1, ..., n then do
         if y^{i} * (\theta * x^{(i)} + \theta_{0}) \le 0 then
             update \theta = \theta + y^{(i)}x^{(i)}
             update \theta_0 = \theta_0 + y^{(i)}
         end if
    end for
end for
```

Decision Boundary

The decision boundary is the set of points which satisfy $\theta * x^{(i)} + \theta_0 = 0$

Margin Boundary

The margin boundary is the set of points which satisfy $\theta * x^{(i)} + \theta_0 = \pm 1$