Let $\hat{L}_w(f) = \sum_{i=1}^N w_i L(y_i, f(x^i))$ *weighted loss*

Algorithm AdaBoost

Assume $\mathcal{B}$ contains functions b taking values in $[-1, 1]$ or $\{\pm 1\}$

Input M, labeled training set $\mathcal{D}$

Initialize $f = 0$

$w_i^1 = \frac{1}{N}$ weight of datapoint x_i

for $k = 1, 2, \ldots M$

1. "learn classifier b^k for $\mathcal{D}$ with weights w^k" $b^k = \text{argmin}_{\mathcal{B}} \hat{L}_{w^k}(b)$
2. compute error $\varepsilon^k = \sum_{i=1}^N w_i^k \frac{1 - y_i b^k(x_i)}{2}$
3. set $\beta^k = \frac{1}{2} \ln \frac{1-\varepsilon^k}{\varepsilon^k}$
4. compute new weights $w_i^{k+1} = \frac{1}{Z^k} w_i^k e^{-\beta^k y_i b^k(x_i)}$ where Z^k is the normalization constant that makes $\sum_i w_i^{k+1} = 1$

Output $f(x) = \sum_{k=1}^M \beta^k b^k(x)$