HW# 2

Entrega: segunda-feira, 12/10/2015 (23:59) **Atenção**: Justifique todas suas respostas.

1. (**Problem 2.1**) In $E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}$, set $\delta = 0.03$ and let

$$\epsilon(M, N, \delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}$$

.

- (a) For M = 1, how many examples do we need to make $\epsilon \leq 0.05$?
- (b) For M = 100, how many examples do we need to make $\epsilon \leq 0.05$?
- (c) For M = 10000, how many examples do we need to make $\epsilon \leq 0.05$?
- 2. (**Problem 2.5**) prove by induction that $\sum_{i=0}^{D} {N \choose i} < N^D + 1$, hence

$$m_{\mathcal{H}}(N) < N^{d_{\text{VC}}} + 1$$

.

- 3. **Problem 2.14** Let $\mathcal{H}_1, \mathcal{H}_2, ..., \mathcal{H}_k$ be K hypothesis sets with finite VC dimension d_{VC} . Let $\mathcal{H} = \mathcal{H}_1 \cup \mathcal{H}_2 \cup ... \cup \mathcal{H}_k$ be the union of these models.
 - (a) Show that $d_{VC}(\mathcal{H}) < K(d_{VC} + 1)$.
 - (b) Suppose that ℓ satisfies $2^{\ell} < 2K\ell^{d_{VC}}$. Show that $d_{VC}(\mathcal{H}) \leq \ell$.
 - (c) Hence, show that

$$d_{\text{VC}}(\mathcal{H}) \le \min(K(d_{\text{VC}} + 1), 7(d_{\text{VC}} + K)\log_2(d_{\text{VC}}K)).$$

That is, $d_{VC}(\mathcal{H}) =)(\max(d_{VC}, K) \log_2 \max(d_{VC}, K))$ is not too bad.

- 4. **Problem 2.20** There are a number of bounds on the generalization error ϵ , all holding with probability at least 1δ .
 - (a) Original VC-bound:

$$\epsilon \le \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(2N)}{\delta}}$$

(b) Rademacher penalty Bound:

$$\epsilon \le \sqrt{\frac{2\ln(2Nm_{\mathcal{H}}(N))}{N}} + \sqrt{\frac{2}{N}\ln\frac{1}{\delta}} + \frac{1}{N}$$

(c) Parrondo and Van den Broek:

$$\epsilon \le \sqrt{\frac{1}{N} \left(2\epsilon + \ln \frac{6m_{\mathcal{H}}(2N)}{\delta} \right)}$$

(d) Devroye:

$$\epsilon \le \sqrt{\frac{1}{2N} \left(4\epsilon(1+\epsilon) + \ln \frac{4m_{\mathcal{H}}(N^2)}{\delta} \right)}$$

Note that (c) and (d) are implicit bounds in ϵ . Fix $d_{\rm VC}=50$ and $\delta=0.05$ and plot these bounds as a function of N. Which is best?

5. (**Problem 2.22**) When there is noise in the data, $E_{\text{out}}(g) = \mathbb{E}_{x,y}[(g(x) - y(x))^2]$, where $y(x) = f(x) + \epsilon$. If ϵ is a zero mean noise random variable with variance σ^2 , show that the bias variance decomposition becomes

$$\mathbb{E}_{\mathcal{D}}[E_{\text{out}}(g)] = \sigma^2 + \text{bias} + \text{var.}$$