

Experts and Multiplicative Weights

slides from Avrim Blum

Using "expert" advice

Say we want to predict the stock market.

- We solicit n "experts" for their advice. (Will the market go up or down?)
- We then want to use their advice somehow to make our prediction. E.g.,

Expt 1	Expt 2	Expt 3	neighbor's dog	truth
down	up	up	up	up
down	up	up	down	down
...

Basic question: Is there a strategy that allows us to do nearly as well as best of these in hindsight?

["expert" = someone with an opinion. Not necessarily someone who knows anything.]

Simpler question

- We have n "experts".
- One of these is perfect (never makes a mistake). We just don't know which one.
- Can we find a strategy that makes no more than $\lg(n)$ mistakes?

Answer: sure. Just take majority vote over all experts that have been correct so far.

➤ Each mistake cuts # available by factor of 2.

➤ Note: this means ok for n to be very large.

➤ What is no expert is perfect?

What if no expert is perfect?

Intuition: Making a mistake doesn't completely disqualify an expert. So, instead of crossing off, just lower its weight.

Weighted Majority Alg:

- Start with all experts having weight 1.
- Predict based on weighted majority vote.
- Penalize mistakes by cutting weight in half.

	prediction				correct
	Y	N	Y	N	
weights	1	1	1	1	
predictions	Y	Y	Y	N	Y
weights	1	1	1	.5	
predictions	Y	N	N	Y	N
weights	1	.5	.5	.5	

Analysis: do nearly as well as best expert in hindsight

- M = # mistakes we've made so far.
- m = # mistakes best expert has made so far.
- W = total weight (starts at n).
- After each mistake, W drops by at least 25%. So, after M mistakes, W is at most $n(3/4)^M$.
- Weight of best expert is $(1/2)^m$. So,

$$(1/2)^m \leq n(3/4)^M$$

$$(4/3)^M \leq n2^m$$

$$M \leq 2.4(m + \lg n)$$

So, if m is small, then M is pretty small too.

Randomized Weighted Majority

$2.4(m + \lg n)$ not so good if the best expert makes a mistake 20% of the time. Can we do better? Yes.

- Instead of taking majority vote, use weights as probabilities. (e.g., if 70% on up, 30% on down, then pick 70:30) Idea: smooth out the worst case.
- Also, generalize $\frac{1}{2}$ to $1 - \epsilon$.

Solves to: $M \leq \frac{-m \ln(1 - \epsilon) + \ln(n)}{\epsilon} \approx (1 + \epsilon/2)m + \frac{1}{\epsilon} \ln(n)$

$M = \text{expected \#mistakes}$ $M \leq 1.39m + 2 \ln n \quad \leftarrow \epsilon = 1/2$

$M \leq 1.15m + 4 \ln n \quad \leftarrow \epsilon = 1/4$

$M \leq 1.07m + 8 \ln n \quad \leftarrow \epsilon = 1/8$

Analysis

- Say at time t we have fraction F_t of weight on experts that made mistake.
- So, we have probability F_t of making a mistake, and we remove an ϵF_t fraction of the total weight.

$$- W_{\text{final}} = n(1 - \epsilon F_1)(1 - \epsilon F_2) \dots$$

$$- \ln(W_{\text{final}}) = \ln(n) + \sum_t [\ln(1 - \epsilon F_t)] \leq \ln(n) - \epsilon \sum_t F_t$$

$$= \ln(n) - \epsilon M.$$

(using $\ln(1-x) < -x$)
($\sum F_t = E[\# \text{ mistakes}]$)

- If best expert makes m mistakes, then $\ln(W_{\text{final}}) > \ln((1-\epsilon)^m)$.
- Now solve: $\ln(n) - \epsilon M > m \ln(1-\epsilon)$.

$$M \leq \frac{-m \ln(1-\epsilon) + \ln(n)}{\epsilon} \approx (1 + \epsilon/2)m + \frac{1}{\epsilon} \log(n)$$