

CPS 196.2

Proper scoring rules

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How to incentivize a weather forecaster

$$P(\text{☀}) = .7$$

$$P(\text{☁}) = .3$$

$$P(\text{☀}) = .8$$

$$P(\text{☁}) = .2$$



- Forecaster's bonus can depend on
 - Prediction
 - Actual weather on predicted day
- Reporting true beliefs should maximize expected bonus

A first attempt

- Suppose we reward the forecaster as follows:
 - Suppose outcome i happened
 - Pay the forecaster the probability he assigned to i
- Notation (in binary setting):
 - x = whether the event (say, rain) happens
 - 1 if rain, 0 if sun
 - q = probability (of rain) reported by forecaster
 - p = forecaster's private (true) probability (of rain)
- So, forecaster receives
$$xq + (1-x)(1-q)$$
- Forecaster's expected payoff for reporting q :
$$pq + (1-p)(1-q)$$
- How does forecaster choose q to maximize expected payoff?

A different reward function

- How about: $xq - q^2/2$
- Forecaster's expected payoff: $pq - q^2/2$
- Derivative w.r.t. q : $p - q$
 - Setting to 0 gives $q=p$
 - Note second derivative is negative
- Forecaster is (strictly) incentivized to tell the truth!
- We say that $xq - q^2/2$ is a **proper scoring rule**
- Little funny: asymmetric between $x=0$ and $x=1$

Brier (aka. quadratic) scoring rule

- How about: $1 - (x - q)^2$
- Forecaster's expected payoff:
- $1 - p(1 - q)^2 - (1 - p)q^2$
- Derivative w.r.t. q : $-2pq + 2p - 2(1-p)q = 2p - 2q$
 - Setting to 0 gives $q=p$
 - Note second derivative is negative
- So Brier is also proper

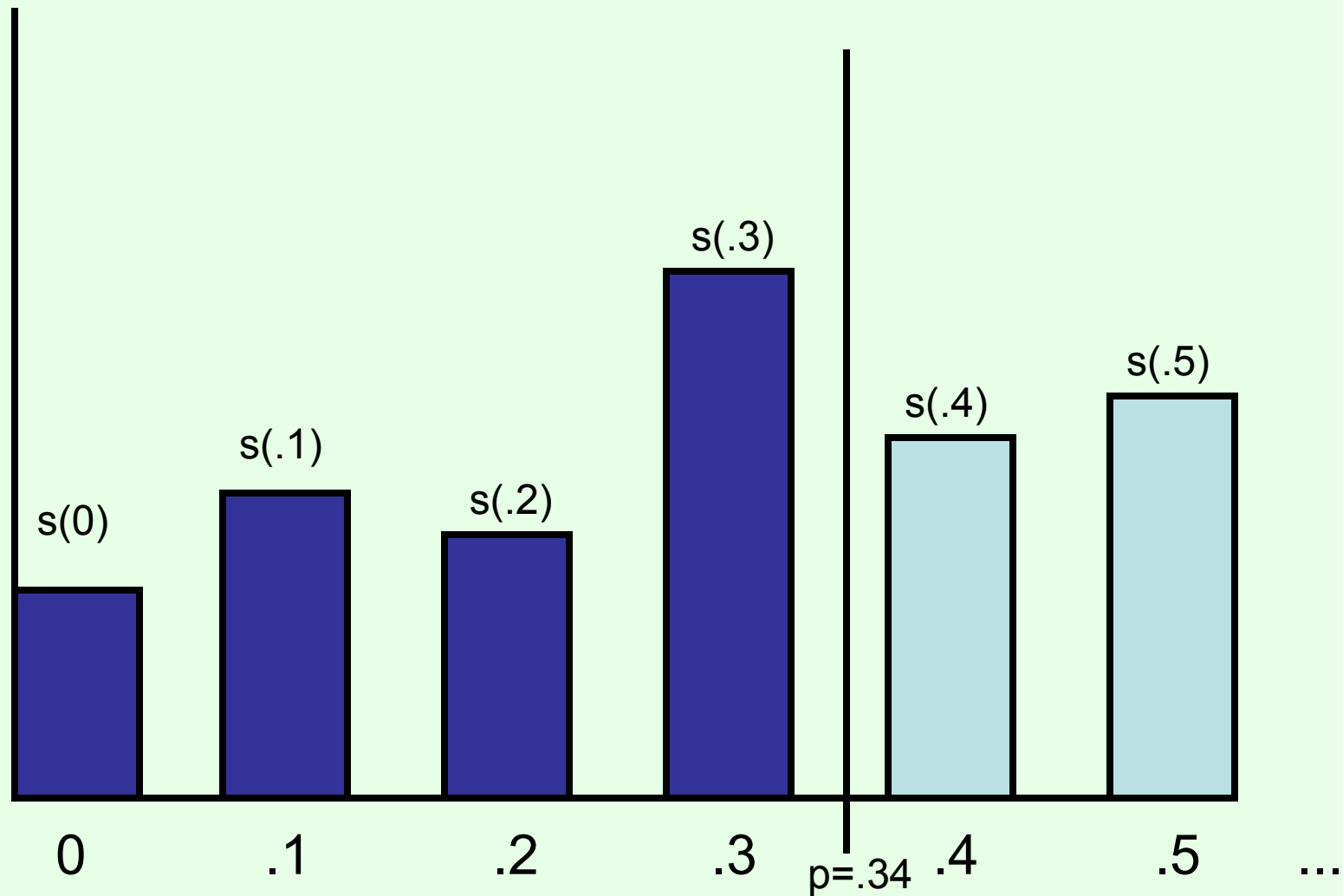
Logarithmic scoring rule

- How about $x \log q + (1-x) \log (1-q)$
- Forecaster's expected payoff:
- $p \log q + (1-p) \log (1-q)$
- Derivative w.r.t. q : $p/q - (1-p)/(1-q)$
 - Setting to 0 gives $q=p$
 - Note second derivative is negative
- So logarithmic is also proper

A securities-based interpretation

- Let's say we allow the forecaster to buy "rain securities"
- Each rain security pays off 1 if it rains, 0 otherwise
- Price starts at 0;
- After the forecaster buys $s(0)$ securities, we increase the price to .1;
- After the forecaster buys $s(.1)$ securities, we increase the price to .2;
- Etc.
- Forecaster should keep buying until price = p
 - (regardless of function s)

A securities-based interpretation



Forecaster pays $s(0)*0 + s(.1)*.1 + s(.2)*.2 + s(.3)*.3$,
expects $.34*(s(0)+s(.1)+s(.2)+s(.3))$

Formulas

- If the forecaster buys up to price q , he will end up buying

$$\sum_{y=0}^q s(y)$$

securities (y is the price)

- He will pay

$$\sum_{y=0}^q s(y)y$$

- He expects a payoff of

$$p^* \sum_{y=0}^q s(y)$$

Making things continuous

- If the forecaster buys up to price q , he will end up buying

$$\int_{y=0}^q s(y)dy$$

securities

- He will pay

$$\int_{y=0}^q s(y)ydy$$

- He expects a payoff of

$$p^* \int_{y=0}^q s(y)dy$$

Example

- $s(y) = 1$
- If the forecaster buys up to price q , he will pay
$$\int_{y=0}^q s(y)ydy = \int_{y=0}^q ydy = q^2/2$$
- He will receive
$$x^* \int_{y=0}^q s(y)dy = x^* \int_{y=0}^q dy = xq$$
- Total payoff = $xq - q^2/2$
- This was the proper scoring rule we started with!

Another example

- $s(y) = y$
- If the forecaster buys up to price q , he will pay $\int_{y=0}^q s(y)ydy = \int_{y=0}^q y^2dy = q^3/3$
- He will receive $x^* \int_{y=0}^q s(y)dy = x^* \int_{y=0}^q ydy = xq^2/2$
- Total payoff = $xq^2/2 - q^3/3$
- Expected payoff = $pq^2/2 - q^3/3$
- Derivative w.r.t. $q = pq - q^2$
 - Expected payoff maximized at $q=p$
- I.e. $xq^2/2 - q^3/3$ is also a proper scoring rule
 - Not surprising, since of course you want to keep buying up to where the price hits p

Yet another example

- $s(y) = e^y$
- If the forecaster buys up to price q , he will pay $\int_{y=0}^q s(y)ydy = \int_{y=0}^q e^y y dy = (q-1)e^q$
- He will receive $x^* \int_{y=0}^q s(y)dy = x^* \int_{y=0}^q e^y dy = xe^q$
- Total payoff = $xe^q - (q-1)e^q$
- Expected payoff = $pe^q - (q-1)e^q$
- Derivative w.r.t. $q = pe^q - (q-1)e^q - e^q$
 - Expected payoff maximized at $q=p$
- I.e. $xe^q - (q-1)e^q$ is also a proper scoring rule
 - Not surprising, since of course you want to keep buying up to where the price hits p

More than two outcomes

$$P(\text{☀}) = .5$$

$$P(\text{☁☔}) = .3$$

$$P(\text{☁⚡}) = .2$$

$$P(\text{☀}) = .8$$

$$P(\text{☁☔}) = .1$$

$$P(\text{☁⚡}) = .1$$



- Can extend all the above by simply asking for separate predictions for whether there will be sun, whether there will be rain, whether there will be lightning

Market scoring rules

- Extend proper scoring rules to multiple forecasters
- Each forecaster is allowed to move current distribution
- Gets rewarded according to new distribution, *minus* reward for previous distribution
- Say we use proper scoring rule $xq - q^2/2$

time


$$P(\text{cloud with rain}) = .3$$



$$x(.3) - (.3)^2/2$$


$$P(\text{cloud with rain}) = .5$$



$$x(.5) - (.5)^2/2 - x(.3) + (.3)^2/2$$


$$P(\text{cloud with rain}) = .4$$



$$x(.4) - (.4)^2/2 - x(.5) + (.5)^2/2$$