#### **Announcements and Such**

- Administrative Stuff
  - HW #5 will be graded soon (and I will post solutions soon)
  - HW #6 is due next Friday (April 22)
    - \* Consists of two (sets of) probability problems: one involving general algebraic reasoning, one involving numerical calculation.
  - I have posted a *Practice Final Exam*. We will go over this Practice Final in class on our last class day next Tuesday (4/19).
  - I will also be doing *course evaluations* on Tuesday
- Unit #4 *Probability & Inductive Logic, Continued* 
  - Measuring Factor #2 measures of confirmation (relevance)
  - Prospects for measuring "Overall Argument Strength"?
  - Probabilism and the Accuracy of Credences
  - Time-Permitting: The Dutch Book Argument for Probabilism

### Three Grades of Measurement

• Suppose we are measuring some numerical quantity. Two examples: the temperature of an object o  $[\mathfrak{t}(o)]$  and the degree to which E confirms H  $[\mathfrak{c}(H,E)]$ . Each of these cases involves three grades of measurement.

#### **Qualitative Measurement**

- \* **Temperature**. This first grade of temperature measurement involves one object *o* being *warm* (or *cold*). This will correspond to the temperature of *o* being *above some threshold t*.
  - "Object o is warm"  $\mapsto t(o) > t$ .
- \* **Confirmation**. This first grade of confirmation measurement involves an argument being *strong* (or *weak*) in the Factor #2 sense. This will correspond to the degree of confirmation that *E* provides for *H* being *above some threshold t*.
  - "E: H is strong"  $\mapsto c(H, E) > t$ .

#### **Comparative Measurement**

- \* **Temperature**. This first grade of temperature measurement involves one object  $o_1$  being warm er (or cold er) than another object  $o_2$ .
  - "Object  $o_1$  is warmer than object  $o_2$ "  $\mapsto \mathfrak{t}(o_1) > \mathfrak{t}(o_2)$ .
- \* **Confirmation**. This involves one argument  $E_1 : H_1$  being strong er (or weak er) than another argument  $E_2 : H_2$ .
  - " $E_1$  :  $H_1$  is stronger than  $E_2$  :  $H_2$ "  $\mapsto \mathfrak{c}(H_1, E_1) > \mathfrak{c}(H_2, E_2)$ .

#### **Numerical Measurement**

- \* **Temperature**. This involves an object *o* having a precise numerical temperature.
  - "Object o is 32 degrees Fahrenheit"  $\mapsto t(o) = 32^{\circ}$  Fahrenheit.
- \* **Confirmation**. This involves an argument E : H having a precise numerical degree of confirmation/strength (in the Factor #2 sense).
  - "The degree to which *E* confirms *H* is 1/2."  $\mapsto \mathfrak{c}(H, E) = 1/2$ .

### **Measuring Factor 2: Degrees of Confirmation I**

- *Dozens* of relevance/confirmation measures have been proposed in the literature. Here are the four most popular measures (each defined on a [-1, +1] scale, for ease of comparison).
  - The *Difference*:  $d(H, E) = Pr(H \mid E) Pr(H)$
  - The *Ratio*:  $r(H, E) = \frac{\Pr(H \mid E) \Pr(H)}{\Pr(H \mid E) + \Pr(H)}$
  - The *Likelihood-Ratio*:  $l(H, E) = \frac{\Pr(E \mid H) \Pr(E \mid \sim H)}{\Pr(E \mid H) + \Pr(E \mid \sim H)}$
  - The *Normalized-Difference*:

$$s(H,E) = \Pr(H \mid E) - \Pr(H \mid \sim E) = \frac{1}{\Pr(\sim E)} \cdot d(H,E)$$

• *A fortiori, all* Bayesian confirmation measures agree on *qualitative* judgments like "*E* confirms/disconfirms/is irrelevant to *H*". But, these measures *disagree* with each other in various ways — *comparatively*.

# **Measuring Factor 2: Degrees of Confirmation III**

- There is a relatively simple way of narrowing the field of competing measures of degree of confirmation, which is based on *thinking of inductive logic as a generalization of deductive logic*.
- The likelihood-ratio measure *l* stands out from the other relevance measures in the literature, since *l* is the only relevance measure that gets the (non-trivial) deductive cases right (as cases of *extreme relevance*).
- That is, l is the only measure (defined on the scale [-1, +1]) that satisfies:

• Here, we assume that  $\mathfrak{c}$  is *defined*, which constrains the unconditional Pr's.

### **Measuring Factor 2: Degrees of Confirmation V**

• Here's how our 4 relevance measures handle non-trivial deductive cases.

• 
$$l(H, E) =$$

$$\begin{cases}
+1 & \text{if } E \vDash H, \Pr(E) > 0, \Pr(H) \in (0, 1) \\
-1 & \text{if } E \vDash \sim H, \Pr(E) > 0, \Pr(H) \in (0, 1)
\end{cases}$$

• 
$$d(H, E) = \begin{cases} \Pr(\sim H) & \text{if } E \vDash H, \Pr(E) > 0 \\ -\Pr(H) & \text{if } E \vDash \sim H, \Pr(E) > 0 \end{cases}$$

• 
$$r(H, E) = \begin{cases} \frac{1 - \Pr(H)}{1 + \Pr(H)} & \text{if } E \vDash H, \Pr(E) > 0, \Pr(H) > 0 \\ -1 & \text{if } E \vDash \sim H, \Pr(E) > 0, \Pr(H) > 0 \end{cases}$$
  
•  $s(H, E) = \begin{cases} \Pr(\sim H \mid \sim E) & \text{if } E \vDash H, \Pr(E) \in (0, 1) \\ -\Pr(H \mid \sim E) & \text{if } E \vDash \sim H, \Pr(E) \in (0, 1) \end{cases}$ 

• 
$$s(H, E) = \begin{cases} \Pr(\sim H \mid \sim E) & \text{if } E \vDash H, \Pr(E) \in (0, 1) \\ -\Pr(H \mid \sim E) & \text{if } E \vDash \sim H, \Pr(E) \in (0, 1) \end{cases}$$

 $\bullet$  From an inductive-logical point of view, this favors l over the other measures. Other considerations can also be used to narrow the field.

### Can We Measure Argument Strength (Numerically)? I

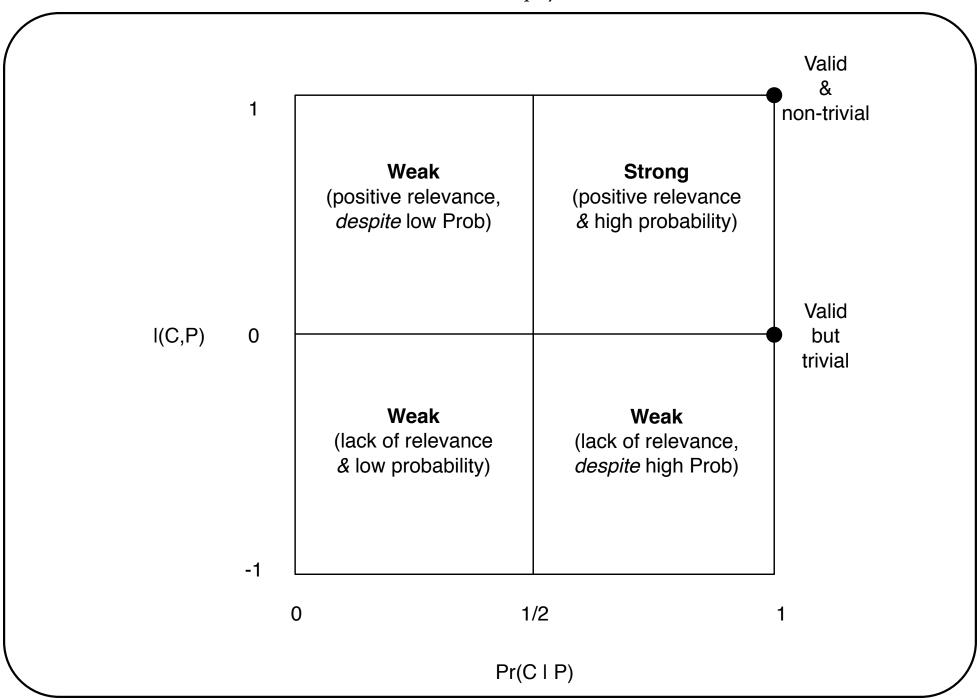
- We know how to measure Factor #1 this is just the conditional probability of the conclusion, given the premise:  $Pr(C \mid P)$ .
- We have some idea of how we might go about measuring Factor #2-a measure like l(C,P) seems a plausible candidate. Let's run with that.
- This allows us to give a *numerical* version of our "Two-Factor" Chart for graphing the two components of argument strength (next slide).
- Every argument will have associated with it an *ordered pair/vector*:  $\langle \Pr(C \mid P), l(C, P) \rangle$ , which records values for both Factors.
- However, it is not at all clear how we might *combine* these two measures to yield a *single measure* of *overall* argument strength.
- Presumably, such a measure would be *some function* f *of*  $Pr(C \mid P)$  and l(C, P). The challenge is to say *which function* f *is.* Let's think about this, in terms of our *three grades of measurement*.

• We've already given a definition of *qualitative* argument strength.

That's the *first grade of measurement* for "overall argument strength".

**Proposal** #3. An argument P : C is *inductively strong* iff

- (1) *C* is probable, given *P*, i.e.,  $Pr(C \mid P) > \frac{1}{2}$ , and
- (2) P is positively relevant to C, i.e.,  $Pr(C \mid P) > Pr(C)$ .
- This places a strong constraint on the shape of f. Specifically, it requires that f be above some threshold t in the upper-right quandrant of our 4-quadrant chart, and below t in the other three compartments.
- We can visualize f as adding a *third dimension* to our 4-quadrant chart (imagine a z-axis, coming out of the chart). The height of each point in this third dimension will correspond to the value of f(x, y).
- Of course, this qualitative constraint is not the end of the story. To get a better grip on f, we'd need to think about its *comparative* structure. This would involve thinking about various  $pairs \langle x, y \rangle$  and their (intuitive) comparative relationship to each other...



#### Probabilism and The Accuracy of Credences I

- Many philosophers have argued for **Probabilism**, which is the claim that one's degrees of confidence (*i.e.*, one's credences) *should obey the probability calculus*. I will discuss one argument for probabilism.
- In epistemology (the theory of knowledge and rational belief), it is typical to suppose that *accuracy* in one's judgments is a virtue.
- For instance, when it comes to (qualitative) *belief*, it is better to have true beliefs than false beliefs. If a belief is false, then it *misrepresents* the world, and this is generally agreed to be (epistemically) *bad*.
- Something similar can be said for credences. Here is a principle.

  The Principle of Gradational Accuracy (qualitative rendition). One ought to be more confident in truths than in falsehoods.
- Ideally, one would assign maximal confidence to all the truths and minimal confidence to all the falsehoods (think: omniscient agents).

## Probabilism and The Accuracy of Credences II

- Of course, it would be far too strong to require all rational agents to live up to this ideal. But, we can use this ideal notion to generate an interesting argument for probabilism.
- Let's call the ideal credence function (in a possible world) the vindicated credence function. I will use  $v_w(\cdot)$  to denote this ideal function.

$$v_w(p) = \begin{cases} 1 & \text{if } p \text{ is true in } w, \\ 0 & \text{if } p \text{ is false in } w. \end{cases}$$

- We can use  $v_w(\cdot)$  to state a quantitative form of the PGA.

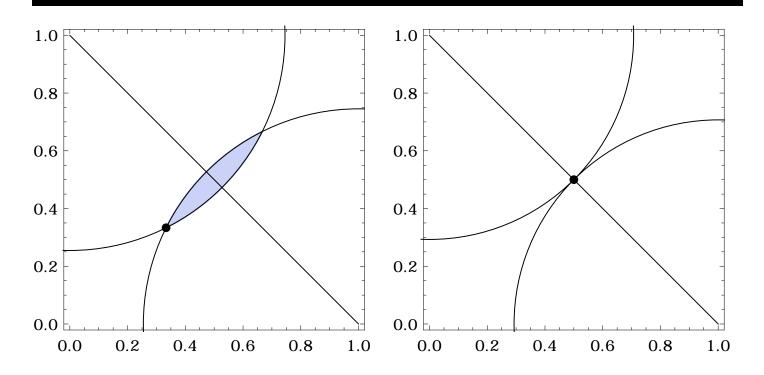
  The Principle of Gradational Accuracy (PGA, quantitative rendition).

  The closer a credence function  $b(\cdot)$  is to  $v_w(\cdot)$ , the better.
- To precisify PGA, we need a way to measure the *distance* between a credence function  $b(\cdot)$  and the vindicated/ideal function  $v_w(\cdot)$ .

### Probabilism and The Accuracy of Credences III

- Because we are only dealing with finite probability spaces,  $b(\cdot)$  and  $v_w(\cdot)$  will always be representable as *finite vectors of real numbers*.
- So, distance between  $b(\cdot)$  and  $v_w(\cdot)$  is just distance between finite vectors of real numbers. A very natural way to measure the distance between such vectors is *via* (squared) *Euclidean distance*.
- To make things easy, let's focus on the simplest possible example. Suppose we're assigning credences over a language with one atomic sentence: P. This means we'll have just *two states*:  $\{P, \sim P\}$ .
- So, any assignment of credence in this case will consist of vector containing two numbers:  $\langle b(P), b(\sim P) \rangle$ . This means we can visualize all such credences *via* a two-dimensional plot.
- On the next slide, I use such a plot to explain the simplest case of what I will call *the accuracy dominance argument for probabilism*.

### Probabilism and The Accuracy of Credences IV



• The diagonal lines are the *probabilistic b*'s (on  $\langle P, \sim P \rangle$ ). The point  $\langle 1, 0 \rangle$  ( $\langle 0, 1 \rangle$ ) corresponds to the values assigned by  $v_w(\cdot)$  in the  $P(\sim P)$  world.

**Theorem** (de Finetti). b is non-probabilistic  $\Leftrightarrow$  there exists a  $b'(\cdot)$  which is (Euclidean) closer to  $v_w(\cdot)$  in every possible world.

• The plot on the left (right) explains the  $\Rightarrow$  ( $\Leftarrow$ ) direction.

# The Dutch Book Argument for Probabilism I

- The key assumptions/set-up of the Dutch Book argument are as follows:
  - For each proposition p that our agent (Mr. B) entertains at t, Mr. B must announce a number q(p) called his *betting quotient* on p, at t and *then* Ms. A (the bookie) will choose the *stake*  $\mathfrak{s}$  of the bet.
  - |s| should be small in relation to Mr. B's total wealth (more on this later). But, it can be positive or negative (so, she can "switch sides").

Mr. B's payoff (in \$) for a bet about 
$$p = \begin{cases} \mathfrak{s} - q(p) \cdot \mathfrak{s} & \text{if } p \text{ is true.} \\ -q(p) \cdot \mathfrak{s} & \text{if } p \text{ is false.} \end{cases}$$

- NOTE: If  $\mathfrak{s} > 0$ , then the bet is *on* p, if  $\mathfrak{s} < 0$ , then the bet is *against* p.
- q(p) is taken to be a measure of Mr. B's *degree of belief* in p (at t).
- If there is a sequence of multiple bets on multiple propositions, then Mr. B's total payoff is the *sum* of the payoffs for each bet on each proposition. This is called "the package principle".

### The Dutch Book Argument for Probabilism II

- The **Dutch Book Theorem** (DBT) has four parts [3 axioms for  $Pr(\cdot)$  plus 1 definition of  $Pr(\cdot | \cdot)$ ]. In each part, we prove that if  $q(\cdot)$  [or  $q(\cdot | \cdot)$ ] *violates* the axioms (or defn.), then  $q(\cdot)$  is *in*coherent.
- If Mr. B violates Axiom 1, then his q is incoherent. Proof:
  - If q(p) = a < 0, then Ms. A sets  $\mathfrak{s} < 0$ , and Mr. B's payoff is  $\mathfrak{s} a\mathfrak{s} < 0$  if p, and  $-a\mathfrak{s} < 0$  if  $\sim p$ . [If  $q(p) \ge 0$ , then Mr. B's payoff is  $\mathfrak{s} q\mathfrak{s} \ge 0$  if  $\mathfrak{s} > 0$  and p is true, and  $-q\mathfrak{s} \ge 0$  if  $\mathfrak{s} < 0$  and  $\sim p$ , avoiding *this* Book.]
- If Mr. B violates Axiom 2, then his q is incoherent. Proof:
  - If Mr. B assigns  $q(\top) = a < 1$ , then Ms. A sets  $\mathfrak{s} < 0$ , and Mr. B's payoff is always  $\mathfrak{s} a\mathfrak{s} < 0$ , since  $\top$  cannot be false.
  - Similarly, if Mr. B assigns  $q(\top) = a > 1$ , then Ms. A sets  $\mathfrak{s} > 0$ , and Mr. B's payoff is always  $\mathfrak{s} a\mathfrak{s} < 0$ , since  $\top$  cannot be false.
    - \* NOTE: if  $q(\top) = 1$ , then Mr. B's payoff is always  $\mathfrak{s} \mathfrak{s} = 0$ , which avoids *this particular* Dutch Book.

## The Dutch Book Argument for Probabilism III

• Axiom 3 requires that

$$Pr(p \vee r) = Pr(p) + Pr(r)$$

if p and r are inconsistent (*i.e.*, if they can't both be true).

- The argument for this *additivity* axiom is more controversial. The main source of controversy is the "package principle".
- I will now go through the additivity case of the DBT.
- **Setup**: Let p and r be some pair of inconsistent propositions that the agent entertains at t. And, suppose Mr. B announces these q's:

$$q(p) = a$$
,  $q(r) = b$ , and  $q(p \lor r) = c$ , where  $c \ne a + b$ .

• This leaves Mr. B susceptible to a *Dutch Book*. Next: the proof of this case of the DBT (note how this presupposes the "package principle").

## The Dutch Book Argument for Probabilism IV

- Case 1: c < a + b. Ms. A asks Mr. B to make *all 3* of these bets ( $\mathfrak{s} = +\$1$ ):
  - 1. Bet a on p to win (1-a) if p, and to lose a if p.
  - 2. Bet b on r to win (1-b) if r, and to lose b if r.
  - 3. Bet \$(1-c) against  $p \vee r$  to win \$c if  $\sim (p \vee r)$ , and lose \$(1-c) o.w.
- Since p and r are mutually exclusive (by assumption of the additivity axiom), the conjunction p & r cannot be true.  $\therefore$  There are 3 cases:

Case	Payoff on (1)	Payoff on (2)	Payoff on (3)	Total Payoff
p & ~r	1-a	-b	-(1-c)	c-(a+b)
~p&r	-a	1-b	-(1-c)	c-(a+b)
~p & ~r	-a	-b	С	c-(a+b)

- Since c < a + b, c (a + b) is negative. So, Mr. B loses [c (a + b)].
- Case 2: c > a + b. Ms. A simply reverses the bets ( $\mathfrak{s} = -\$1$ ), and a parallel argument shows that the total payoff for Mr. B is \$-[c-(a+b)]<0.
- Note: he can avoid *this* Book, by setting c = a + b.

### The Dutch Book Argument for Probabilism V

- We also need to show that an agent's *conditional* betting quotients  $q(\cdot | \cdot)$  are coherent only if they satisfy our ratio definition of  $Pr(\cdot | \cdot)$ . There's a DBT for this too (note: this case *also* assumes the "package principle").
- Suppose Mr. B announces: q(p & r) = b, q(r) = c > 0, and  $q(p \mid r) = a$ . Ms. A asks Mr. B to make *all* 3 of these bets (stakes depend on quotients!):
- 1. Bet  $(b \cdot c)$  on p & r to win  $[(1-b) \cdot c]$  if p & r, and lose  $(b \cdot c)$  o.w. [s = c]
- 2. Bet  $\{(1-c)\cdot b\}$  against r to win  $\{(b\cdot c) \text{ if } r, \text{ and lose } \{(1-b)\cdot c\} \text{ o.w. } [\mathfrak{s}=b]$
- 3. Bet  $\$[(1-a)\cdot c]$  against p, conditional on r, to win  $\$(a\cdot c)$  if r & p, and lose  $\$[(1-a)\cdot c]$  if  $r \& \sim p$ . If  $\sim r$ , then the bet is *called off*, and payoff is \$0.  $[\mathfrak{s}=c]$

Case	Payoff on (1)	Payoff on (2)	Payoff on (3)	Total Payoff
p & r	$(1-b)\cdot c$	$-[(1-c)\cdot b]$	$-[(1-a)\cdot c]$	$(a \cdot c) - b$
~p & r	$-(b\cdot c)$	$-[(1-c)\cdot b]$	$a \cdot c$	$(a \cdot c) - b$
$\sim \gamma$	$-(b\cdot c)$	$b \cdot c$	0	0

• If  $a < \frac{b}{c}$ , then Mr. B loses *come what may*. If  $a > \frac{b}{c}$ , then Ms. A just asks Mr. B to take the other side on all three bets. So, coherence requires:  $q(p \mid r) = \frac{q(p\&r)}{a(r)}$ .