## The expected utility hypothesis

• A choice under risk is made by maximizing the expected utility,

$$a^* = \underset{a \in \{a_1, a_2, \dots\}}{\operatorname{argmax}} \mathbb{E}[U(a)]$$

- The objective function  $\mathbb{E}[U(a)]$  consists of two types of quantities:
  - -Pr(a): probability distribution
  - -U(a): utility function
- n.b. each possible choice  $a_1, a_2, \ldots$  is a distribution; i.e. we choose one of the random variables.  $U(\cdot)$  is a deterministic function, and U(a) may be seen as a random variable as well.

## The existence of probability distribution

• Under the assumptions As1-As5, for any event A, there exists a unique probability function Pr(A) satisfying  $A \sim_L G[0, Pr(A)]$  where G[a, b] is the event that a uniformly distributed random variable lies in the interval (a, b).

# Elicitation of probabilities

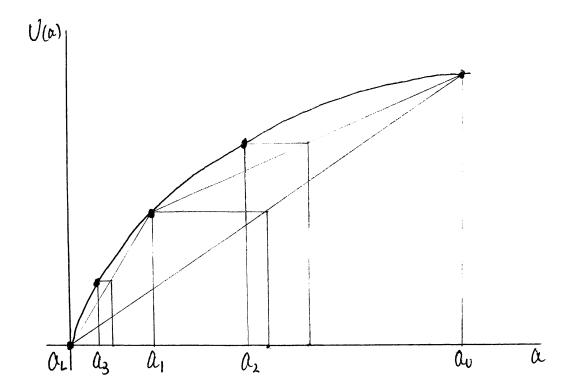
- Repeatable events
  - nonparametric statistics (e.g. quantile regression)
  - parametric statistics (e.g. MLE)
  - sample moments (e.g. mean, variance, etc.)
- Non-repeatable events
  - relying on the relative likelihood
  - reference lotteries
  - the fractile method

# The existence of utility function

- A2 (independence), e.g. Does pizza  $(p) \succ$  burrito (b) get reversed if ordered with horchata (h), i.e.  $\beta p + (1 \beta)h \prec \beta b + (1 \beta)h$ ?
- "U(a) is unique up to an affine transformation." Technically, the implication goes in both directions. That is, U(a) and V(a) represents the same preferences if and only if  $V(a) = \alpha + \beta U(a)$ ,  $(\beta > 0)$ . For a proof, "if" part is straightforward, while "only if" part requires some work. See some graduate textbook, e.g. Jehle & Reny (2011, p.108).

## Elicitation of preferences

- We may estimate  $U(\cdot)$  by using a questionnaire and tracing certainty equivalents, a concept you will learn next week.
- We sequentially determine the certainty equivalent of  $\mathbb{E}[U(a)]$  where a takes only two possible values with the equal probability. We start with an lower bound  $(a_L)$  and an upper bound  $(a_U)$ . Note that since an affine transformation of U(a) is allowed, we may set  $U(a_L) = 0$  and  $U(a_U) = 1$ . For example,
  - $-U(a_1) = \mathbb{E}[U(a)]$  where  $a \in \{a_L, a_U\}$
  - $-U(a_2) = \mathbb{E}[U(a)]$  where  $a \in \{a_1, a_U\}$
  - $-U(a_3) = \mathbb{E}[U(a)]$  where  $a \in \{a_L, a_1\}$



#### Solver in Excel

- A way to solve an equation or a constrained optimization problem in Excel
- If not found in "Data" tab, see the link