

## The expected utility hypothesis

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

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

- 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  $a_1, a_2, \dots$  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

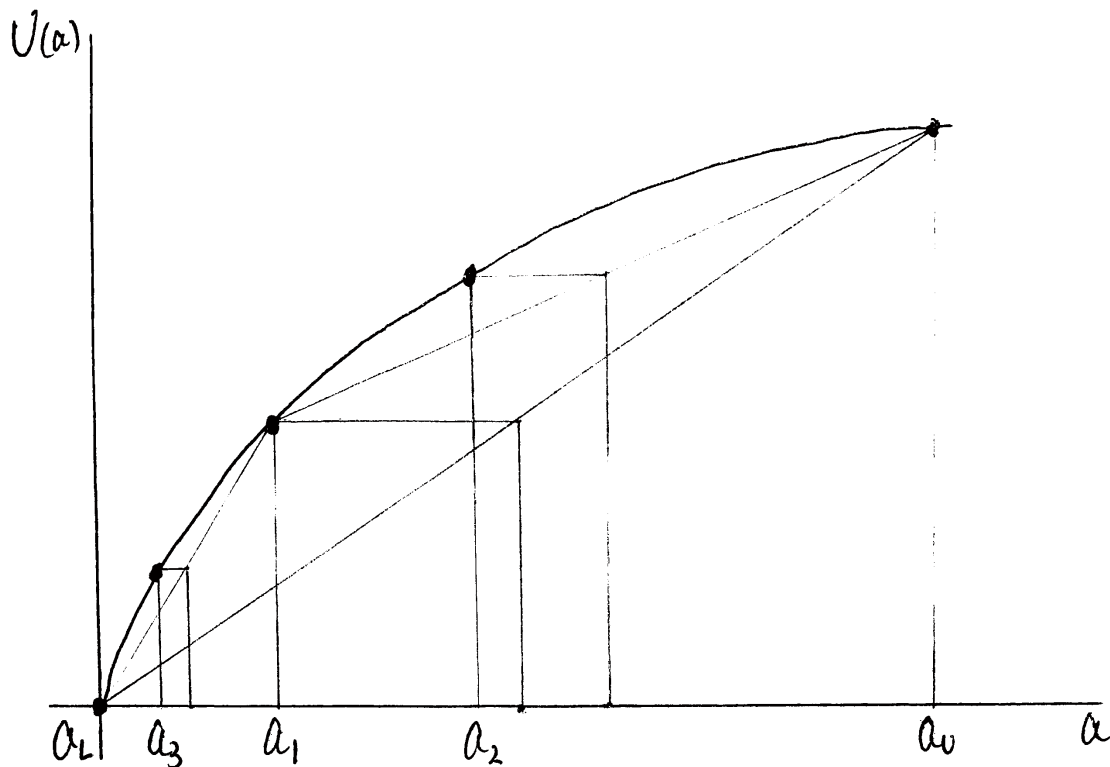
## The existence of utility function

- 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 "Data" tab, see the [link](#)