



# On envy-free allocations in large fair division problems with indivisible goods

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## Fair division of indivisible goods

- $m$  indivisible goods to allocate among  $n$  agents
- allocation  $\mathcal{A} = (\mathcal{A}_i)_{i=1 \dots n}$  is a partition of a set of goods
- $u_g^i$  is agent  $i$ 's satisfaction of a good  $g$
- utilities of agents are additive:

$$u^i(\mathcal{A}_i) = \sum_{g \in \mathcal{A}_i} u_g^i$$

- *For example, division of inheritance*

## Fairness criterion

- Envy-freeness:

Allocation  $\mathcal{A}$  is *envy-free* iff

$$u^i(\mathcal{A}_i) \geq u^i(\mathcal{A}_j) \quad \forall i, j = 1 \dots n$$

## Assumptions

- $u_g^i$  are i.i.d. random variables uniformly distributed on  $[0; 1]$
- the number of goods is large

## The question is:

How often does envy-free allocation of indivisible goods exist?

## Previous results:

J.P. Dickerson et al. show that the probability of envy-free allocation existence tends to 1 as the number of goods becomes large.

## Remark:

They do not obtain an explicit estimate on this probability in terms of  $n$  and  $m$ .

## Our aim:

- obtain explicit estimate
- introduce measure-concentration tools in large fair division problems

## Result

The probability of existence of envy-free allocation is greater, than

$$1 - n^2 \exp\left(\frac{-m}{4(n+1)^3}\right).$$

Note, that probability increases with increasing of number of goods.

## The main ideas of the proof

- **Utilitarian maximum**

Consider an allocation  $\mathcal{A}^{UT}$  in which each good  $g$  is given to an agent  $i$  who desires it most, in other words, agent  $i$  gets  $g$  iff  $u_g^i \geq u_g^j \quad \forall i, j = 1 \dots n$ .

$\mathcal{A}^{UT}$  can be really unfair.

- We use **measure concentration** tools to estimate the probability that agent  $i$  envies agent  $j$  in  $\mathcal{A}^{UT}$ . Measure concentration theory says that the "macroscopic" properties (that depends on a large number of random parameters) of large random objects are non-random, i.e., they are close to expected values with high probability.

## McDiarmid's inequality

- $\xi_1, \xi_2, \dots, \xi_N$  are independent random variables
- function  $f$  is such that for any fixed  $x_1, \dots, x_N$  the random variable  $f(x_1, \dots, x_{i-1}, \xi_i, x_{i+1}, \dots, x_N)$  belongs to interval of length  $c_i$

Then for any  $\varepsilon > 0$

$$\mathbb{P}(|f(\xi) - E(f(\xi))| \geq \varepsilon) \leq 2 \exp\left(-\frac{2\varepsilon^2}{\sum_{i=1}^N c_i^2}\right).$$

**Corollary:** the law of large numbers.

## Contact information

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## References

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