

Equitable and Efficient Healthcare Resource Distribution

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The Problem

- Healthcare resources are normally allocated to **maximize utility**.
 - As measured by health outcomes (QALYs, etc.)



The Problem

- Healthcare resources are normally allocated to **maximize utility**.
 - As measured by health outcomes (QALYs, etc.)
 - This can lead to **very unfair** resource distribution.



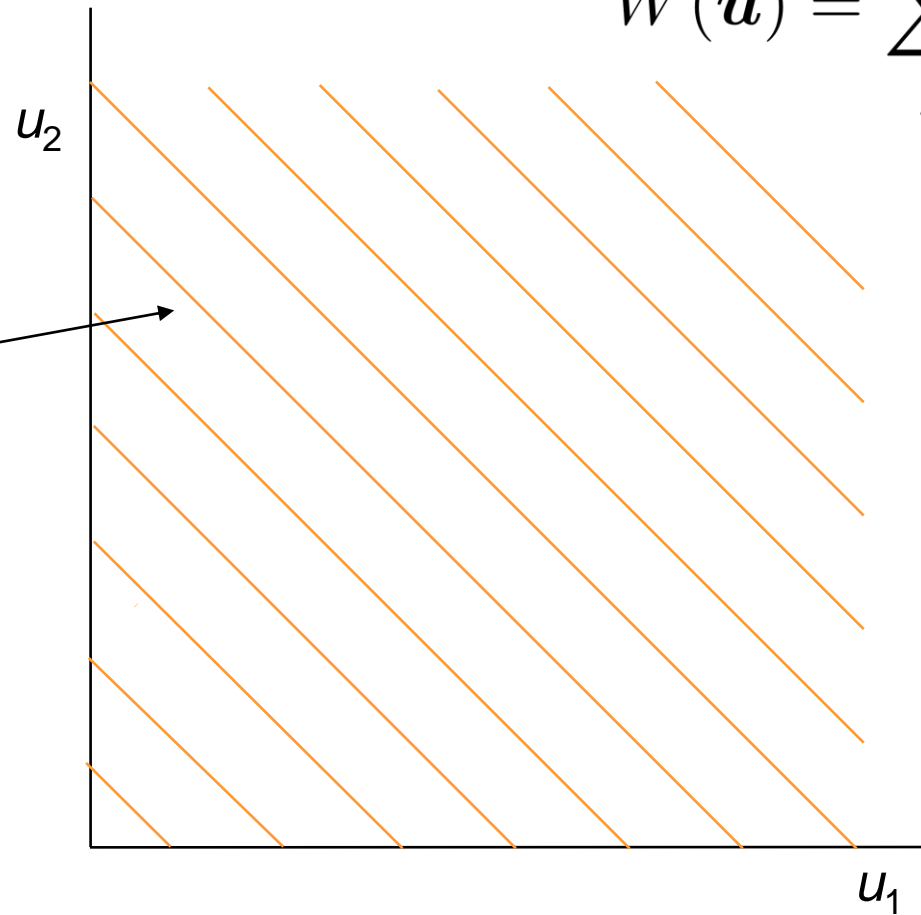
- For example...

The Problem

Utility maximizing
distribution
for 2 patients

$$W(\mathbf{u}) = \sum_j u_j$$

Utility contours
 $u_1 + u_2$



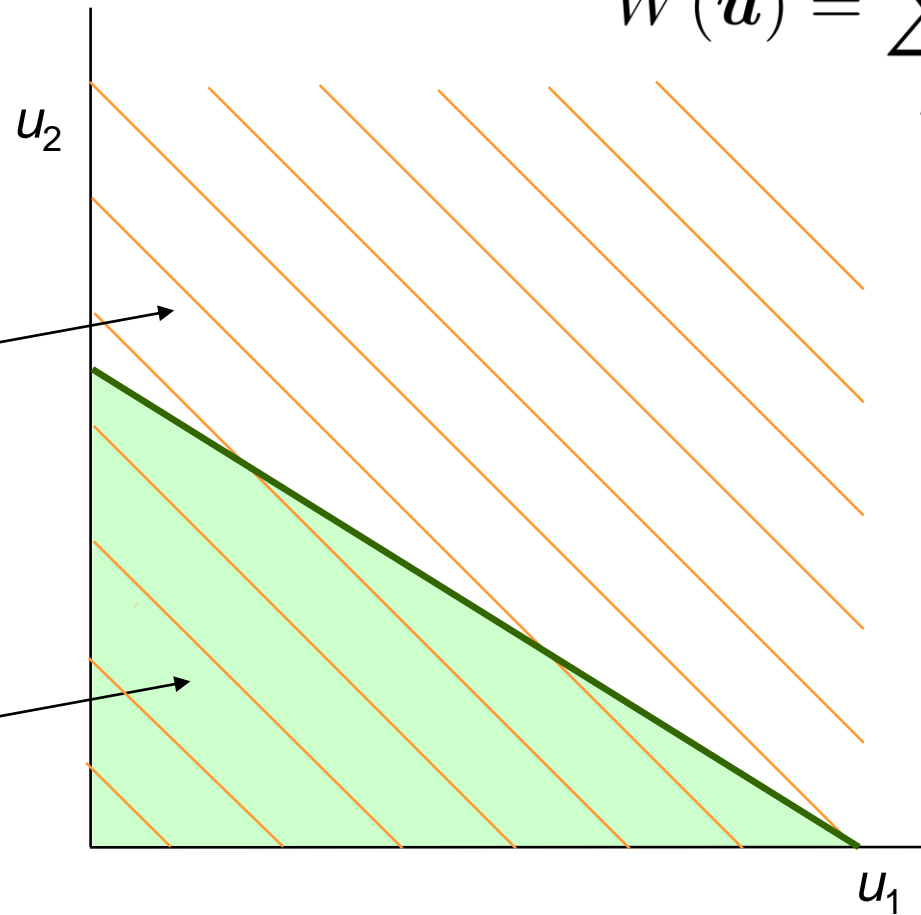
The Problem

Utility maximizing
distribution
for 2 patients,
subject to
budget constraint

$$W(\mathbf{u}) = \sum_j u_j$$

Utility contours
 $u_1 + u_2$

Feasible
region
 $a_1 u_1 + a_2 u_2 \leq B$



Patient 2 is more expensive to treat

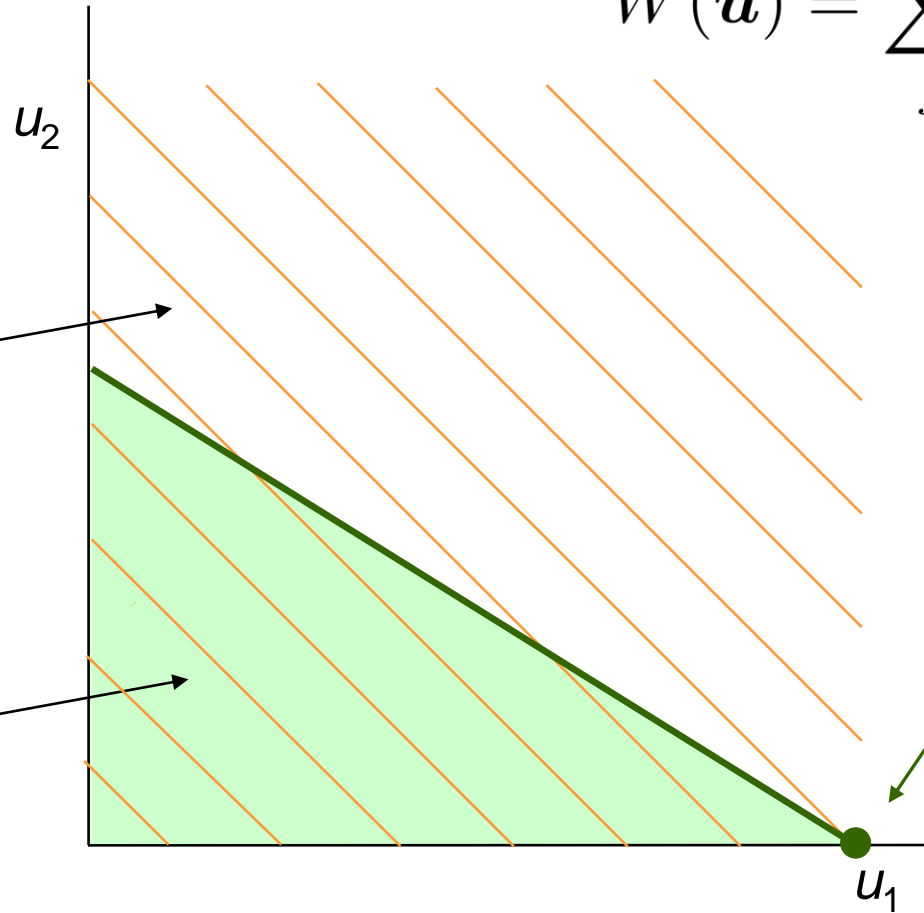
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Patient 1
gets
everything!

Patient 2 is more expensive to treat

The Problem

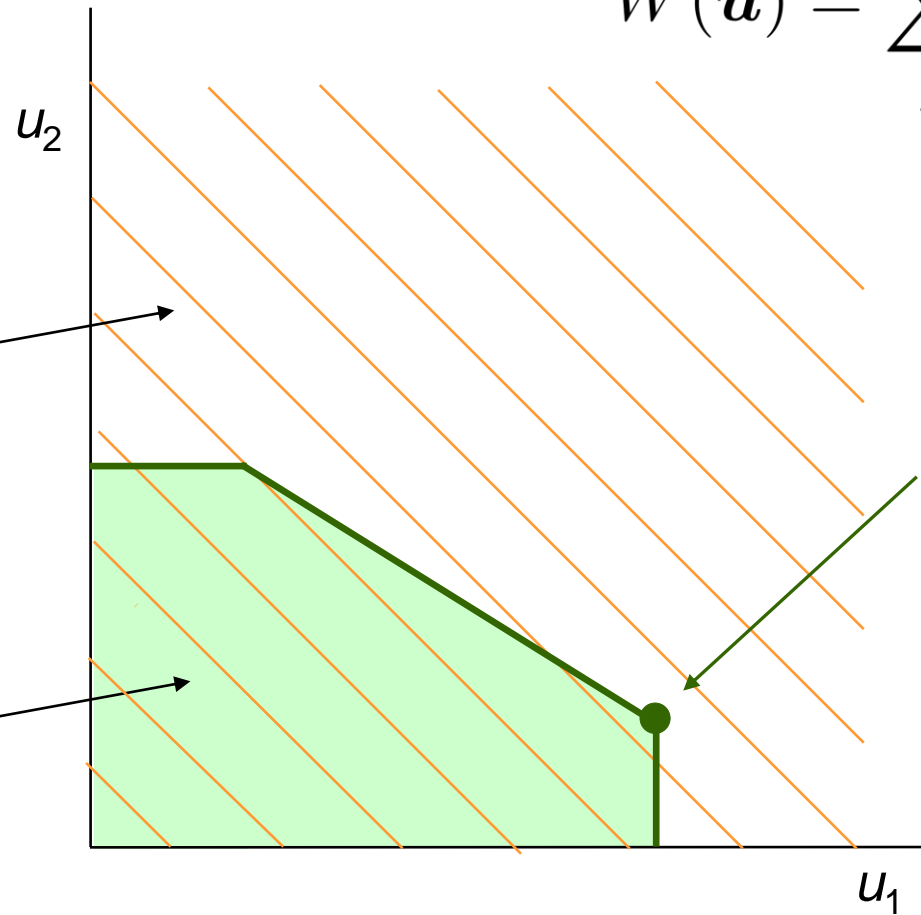
Add **bounds**
on individual
utilities...

$$W(\mathbf{u}) = \sum_j u_j$$

Utility contours
 $u_1 + u_2$

Patient 1
still gets
maximum
allotment

Feasible
region



The Problem

- True, these constraints are simplistic...
 - ...and such extreme solutions rarely occur in practice.

The Problem

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 - ...and such extreme solutions rarely occur in practice.
 - Yet the complexity of the constraints only **conceals the basic inadequacy** of the objective function!

The Problem

- True, these constraints are simplistic...
 - ...and such extreme solutions rarely occur in practice.
 - Yet the complexity of the constraints only **conceals the basic inadequacy** of the objective function!
 - We need an objective function that **incorporates equity** as well as efficiency.

Health Equity



The Problem

- Several **social welfare functions** have been designed for this purpose...

The Problem

- Several **social welfare functions** have been designed for this purpose...
- Yet many of these also lead to **extreme solutions**...
 - ...when maximized subject to **simple constraints**.
 - Extreme solutions may not often occur **in practice**...
 - ...but only **by accident**, not due to some **underlying concept of fairness** in the constraint set.

Research Program

- This suggests a research program.
 - Study the **structure of optimal solutions** using various social welfare functions (SWFs), subject to **simple constraints**.
 - Identify SWFs that yield **reasonable solutions**.



Research Program

- This suggests a research program.
 - Study the **structure of optimal solutions** using various social welfare functions (SWFs), subject to **simple constraints**.
 - Identify SWFs that yield **reasonable solutions**.
 - The results apply to **resource distribution in general**.



Generic Welfare Optimization Problem

$$\max_{\mathbf{x}} \left\{ W(\mathbf{U}(\mathbf{x})) \mid \sum_j x_j \leq B, \bar{\mathbf{c}} \leq \mathbf{x} \leq \bar{\mathbf{d}} \right\}$$

Generic Welfare Optimization Problem

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Resource
allocation
vector
 (x_1, \dots, x_n)
to patient
groups

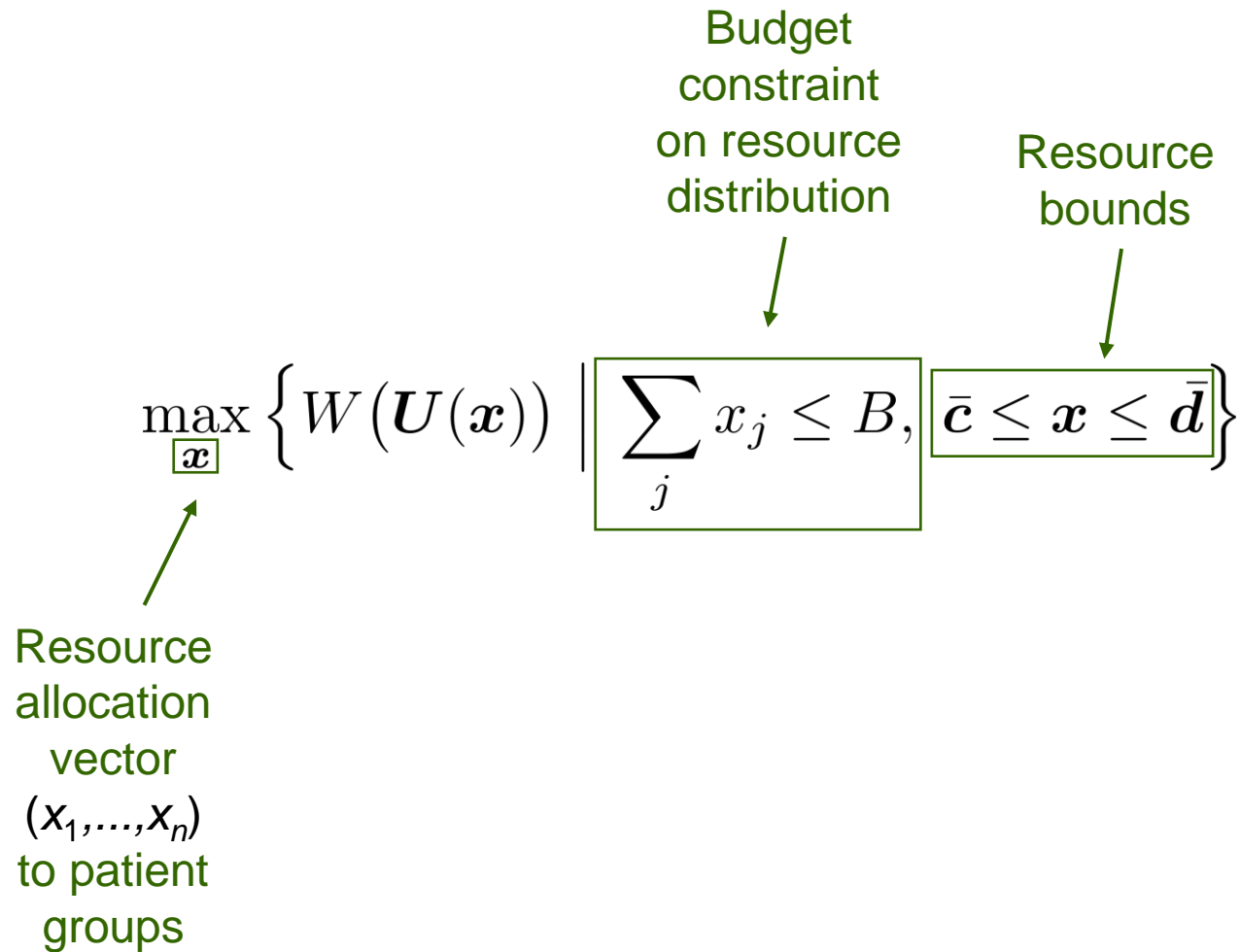
Generic Welfare Optimization Problem

Budget
constraint
on resource
distribution

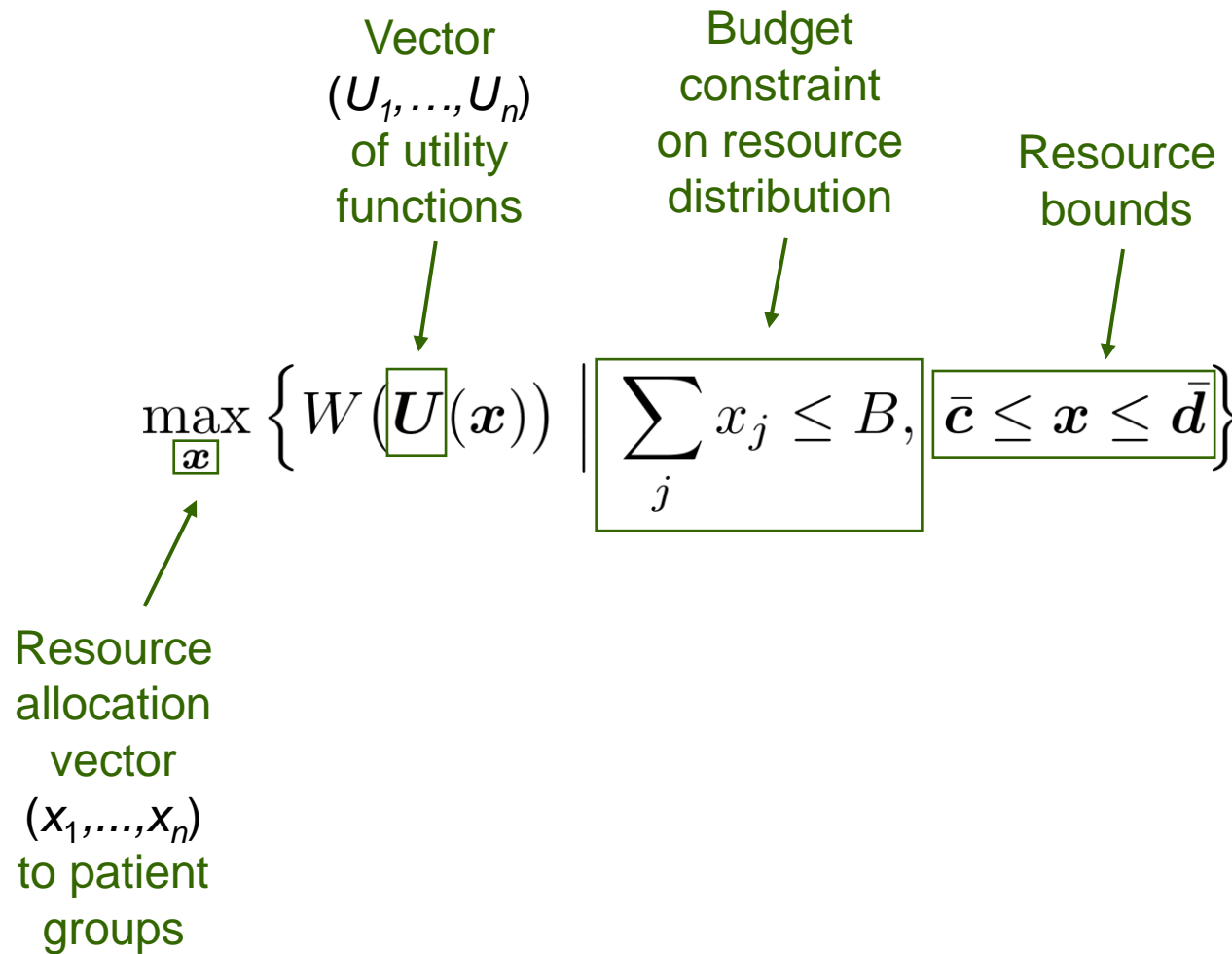
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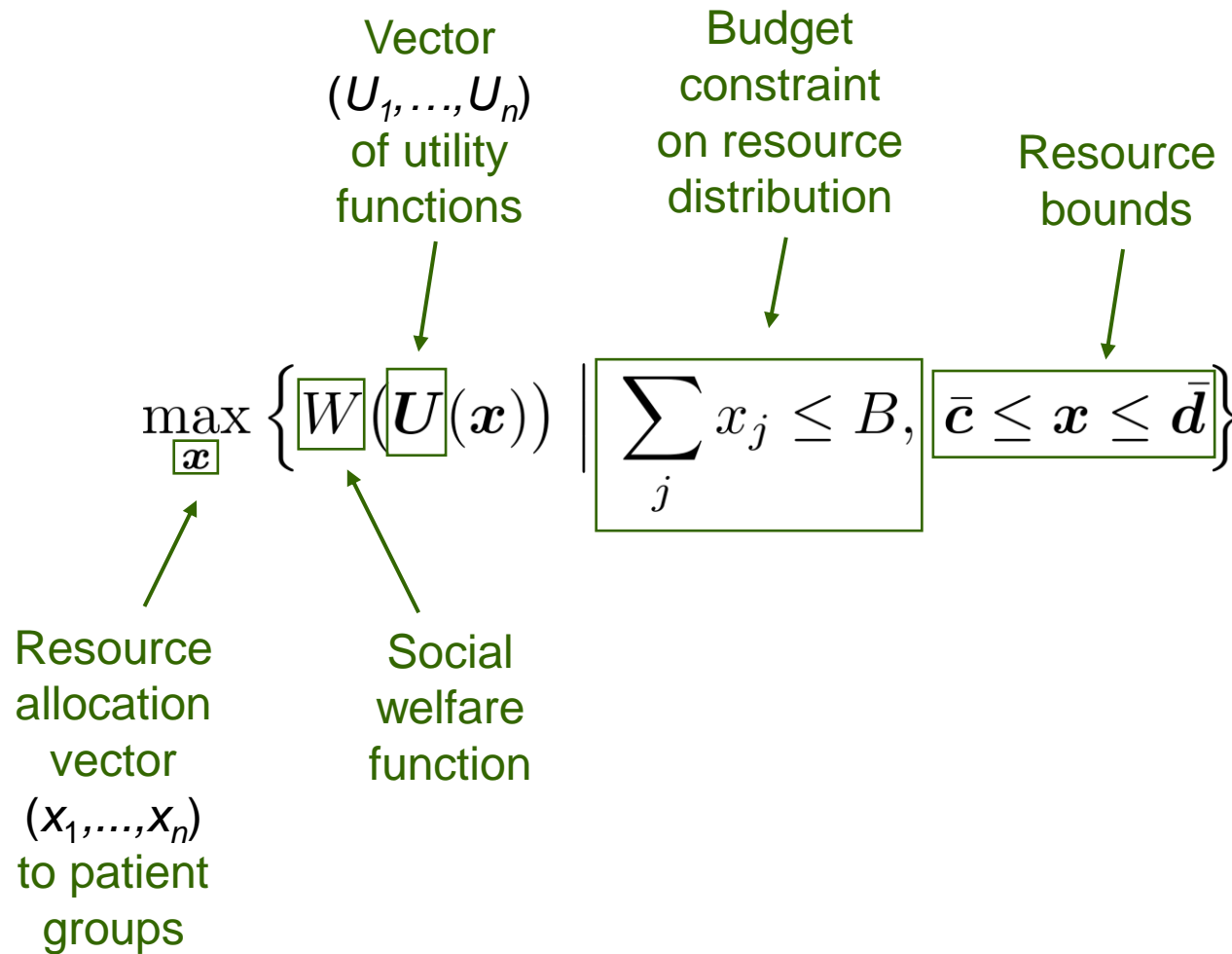
Generic Welfare Optimization Problem



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Generic Welfare Optimization Problem

We suppose $\mathbf{U}(\mathbf{x})$ has the form

$$U(\mathbf{x}) = (x_1/a_1, \dots, x_n/a_n), \quad \text{with } a_j > 0, \text{ all } j$$

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This allows us to get rid of \mathbf{x} :

$$\max_{\mathbf{u}} \{W(\mathbf{u}) \mid \mathbf{a}^\top \mathbf{u} \leq B, \mathbf{c} \leq \mathbf{u} \leq \mathbf{d}\}$$

where $c_j = \bar{c}_j/a_j, \quad d_j = \bar{d}_j/a_j$

Utility
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Generic Welfare Optimization Problem

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$$\text{where } c_j = \bar{c}_j/a_j, \quad d_j = \bar{d}_j/a_j$$

A large a_j means that patient group j is expensive to treat.

We assume $a_1 \leq \dots \leq a_n$

Healthcare Example

- Based on budget decisions in UK National Health Service
- Allocate treatment resources to patient groups
 - Groups characterized by disease and prognosis.
 - Based on cost and estimated QALY estimates with and without treatment*
 - We will solve this example later.**

*Data reflect a particular situation and are not valid in general.

**Solutions presented here should not be taken as a general recommendation for healthcare resource allocation.

QALY
& cost
data

Part 1

Intervention	Cost per person c_i (£)	QALYs gained q_i	Cost per QALY (£)	QALYs without intervention α_i	Subgroup size n_i
<i>Pacemaker for atrioventricular heart block</i>					
Subgroup A	3500	3	1167	13	35
Subgroup B	3500	5	700	10	45
Subgroup C	3500	10	350	5	35
<i>Hip replacement</i>					
Subgroup A	3000	2	1500	3	45
Subgroup B	3000	4	750	4	45
Subgroup C	3000	8	375	5	45
<i>Valve replacement for aortic stenosis</i>					
Subgroup A	4500	3	1500	2.5	20
Subgroup B	4500	5	900	3	20
Subgroup C	4500	10	450	3.5	20
<i>CABG¹ for left main disease</i>					
Mild angina	3000	1.25	2400	4.75	50
Moderate angina	3000	2.25	1333	3.75	55
Severe angina	3000	2.75	1091	3.25	60
<i>CABG for triple vessel disease</i>					
Mild angina	3000	0.5	6000	5.5	50
Moderate angina	3000	1.25	2400	4.75	55
Severe angina	3000	2.25	1333	3.75	60
<i>CABG for double vessel disease</i>					
Mild angina	3000	0.25	12,000	5.75	60
Moderate angina	3000	0.75	4000	5.25	65
Severe angina	3000	1.25	2400	4.75	70

QALY
& cost
data

Part 2

Intervention	Cost per person c_i (£)	QALYs gained q_i	Cost per QALY (£)	QALYs without intervention α_i	Subgroup size n_i
	22,500	4.5	5000	1.1	2
<i>Kidney transplant</i>					
Subgroup A	15,000	4	3750	1	8
Subgroup B	15,000	6	2500	1	8
<i>Kidney dialysis</i>					
<i>Less than 1 year survival</i>					
Subgroup A	5000	0.1	50,000	0.3	8
<i>1-2 years survival</i>					
Subgroup B	12,000	0.4	30,000	0.6	6
<i>2-5 years survival</i>					
Subgroup C	20,000	1.2	16,667	0.5	4
Subgroup D	28,000	1.7	16,471	0.7	4
Subgroup E	36,000	2.3	15,652	0.8	4
<i>5-10 years survival</i>					
Subgroup F	46,000	3.3	13,939	0.6	3
Subgroup G	56,000	3.9	14,359	0.8	2
Subgroup H	66,000	4.7	14,043	0.9	2
Subgroup I	77,000	5.4	14,259	1.1	2
<i>At least 10 years survival</i>					
Subgroup J	88,000	6.5	13,538	0.9	2
Subgroup K	100,000	7.4	13,514	1.0	1
Subgroup L	111,000	8.2	13,537	1.2	1

Healthcare Example

Budget constraint

$$\sum_j n_j c_j y_j \leq \bar{B}$$

The diagram shows the equation $\sum_j n_j c_j y_j \leq \bar{B}$ with three green arrows pointing to the variables n_j , c_j , and y_j . The arrow from n_j points to the text "Size of treatment group j ". The arrow from c_j points to the text "Unit cost of treatment j ". The arrow from y_j points to the text "Fraction of group treated".

Healthcare Example

Budget constraint

$$\sum_j n_j c_j y_j \leq \bar{B}$$

Size of
treatment
group j

Unit cost of
treatment j

Fraction
of group
treated

Utility function

$$u_i = q_i y_i + \alpha_i$$

Treatment
benefit
(QALYs)

QALYs
without
treatment

which implies $y_i = (u_i - \alpha_i) / q_i$

Healthcare Example

Budget constraint

$$\sum_j n_j c_j y_j \leq \bar{B}$$

Size of treatment group j (points to n_j)
 Unit cost of treatment j (points to c_j)
 Fraction of group treated (points to y_j)

Utility function

$$u_i = q_i y_i + \alpha_i$$

Treatment benefit (QALYs) (points to $q_i y_i$)
 QALYs without treatment (points to α_i)

which implies $y_i = (u_i - \alpha_i) / q_i$

So the optimization problem $\max_{\mathbf{u}} \{ W(\mathbf{u}) \mid \mathbf{a}^\top \mathbf{u} \leq B, \mathbf{c} \leq \mathbf{u} \leq \mathbf{d} \}$

becomes

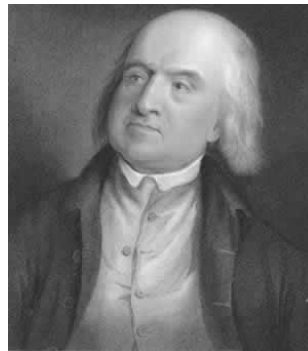
$$\max_{\mathbf{u}} \left\{ W(\mathbf{u}) \mid \sum_j \frac{n_j c_j}{q_j} u_j \leq \bar{B} + \sum_j \frac{n_j c_j \alpha_j}{q_j}; \alpha \leq \mathbf{u} \leq \mathbf{q} + \alpha \right\}$$

Utilitarian SWF

$$W(\mathbf{u}) = \sum_j u_j$$

Proposition. An optimal utilitarian distribution, subject to a resource constraint, allocates **all utility** to individual 1:

$$u_1 = B/a_1, \quad u_j = 0 \text{ for } j = 2, \dots, n$$

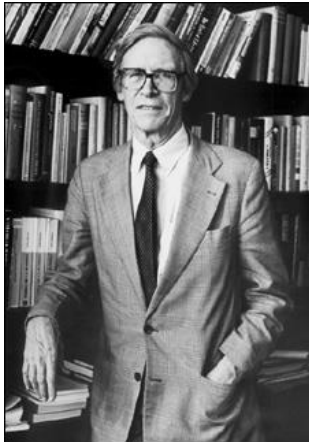


J. Bentham (1776)

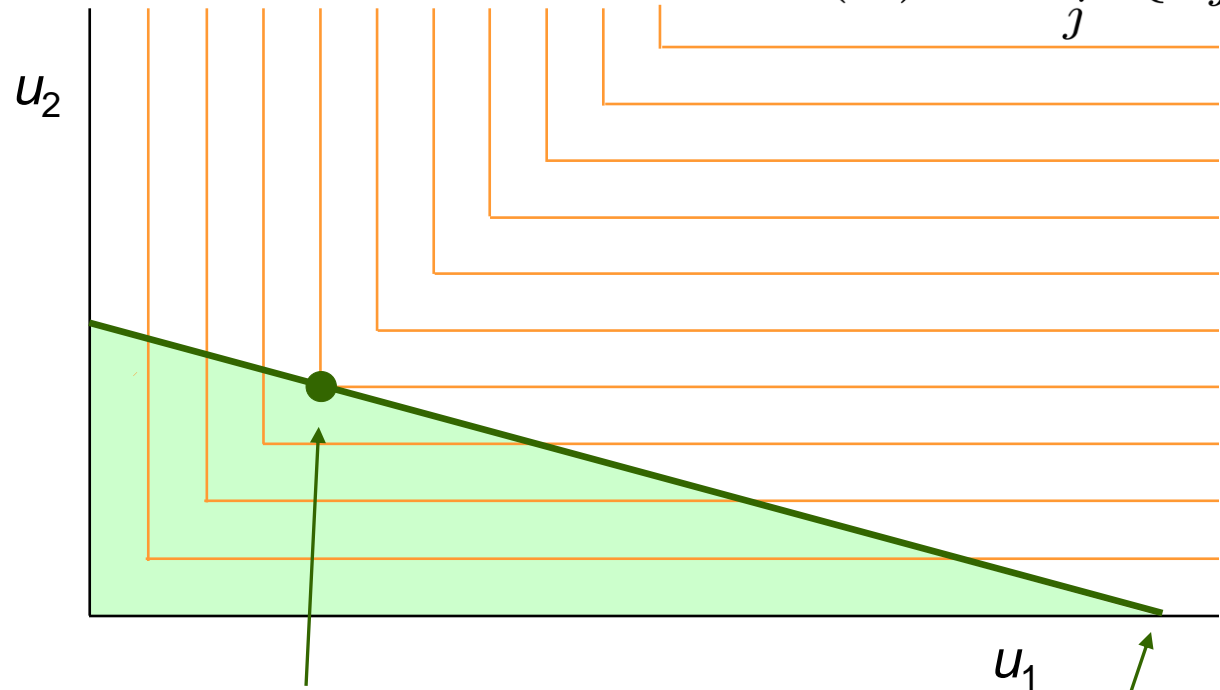
Maximin SWF

$$W(\mathbf{u}) = \min_j \{u_j\}$$

Optimal distribution
subject to budget
constraint



J. Rawls
(1971)



Utility equalized,
Patient 2 gets most
of the resources.

Substantial
sacrifice of
Patient 1

Maximin SWF

$$W(\mathbf{u}) = \min_j \{u_j\}$$

Proposition. An optimal maximin distribution, subject to a resource constraint, distributes utility **equally**:

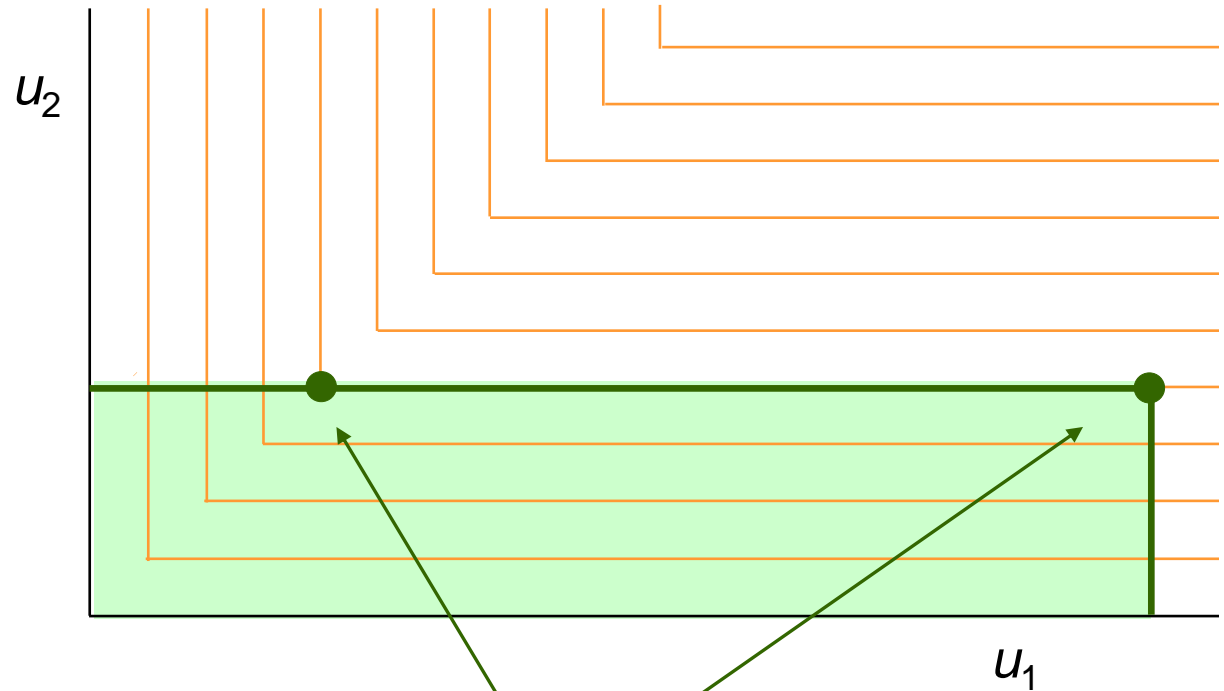
$$u_j = \frac{B}{\sum_i a_i} \text{ for } j = 1, \dots, n$$

Note: Rawls intended this criterion to apply only to the design of **social institutions** and the distribution of "**primary goods.**"

Maximin SWF

$$W(\mathbf{u}) = \min_j \{u_j\}$$

Optimal
distribution
subject to
resource
bounds

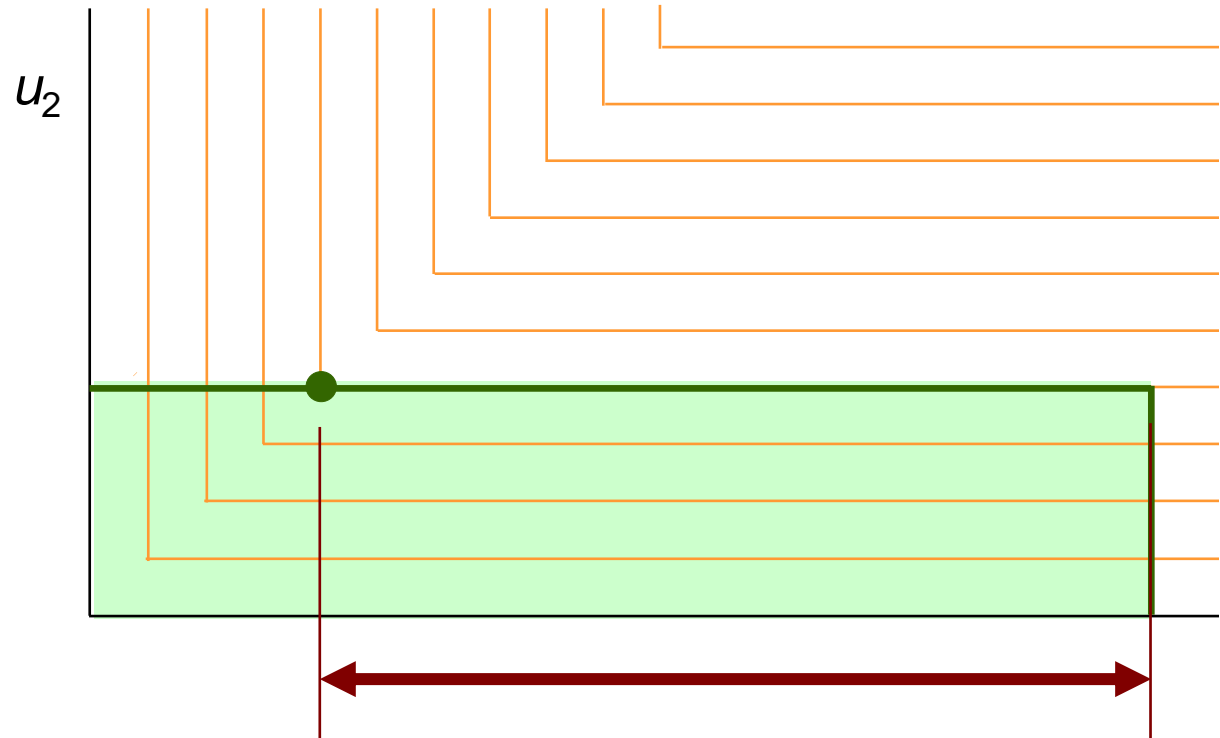


These solutions have same social welfare!

Maximin SWF

$$W(\mathbf{u}) = \min_j \{u_j\}$$

Optimal
distribution
subject to
resource
bounds



Maximin solution can waste
most of the resources!

Maximin SWF

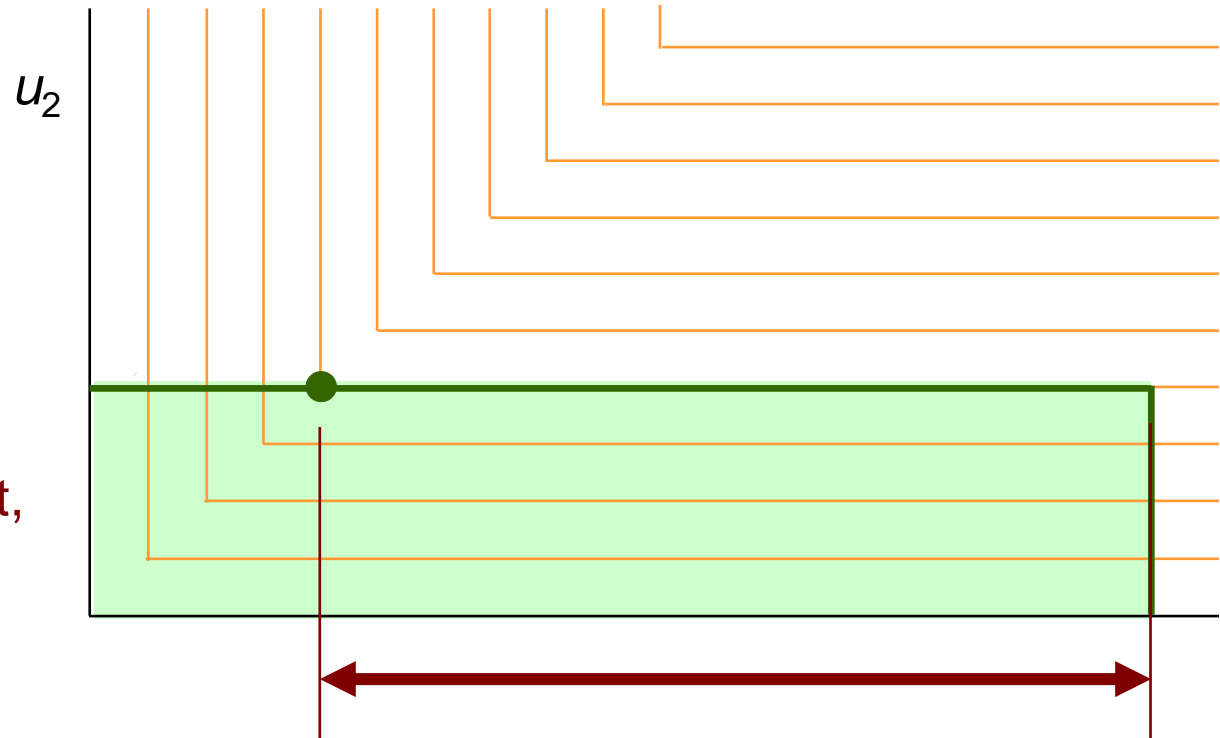
$$W(\mathbf{u}) = \min_j \{u_j\}$$

Optimal distribution
subject to
**resource
bounds**

Remedy: use
leximax solution

Maximize smallest,
then 2nd smallest,
etc.

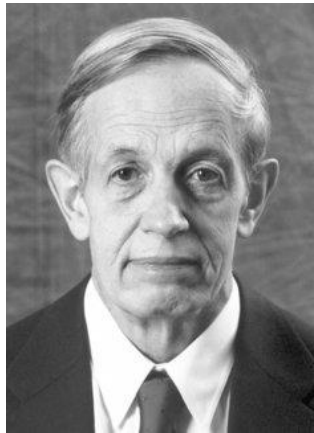
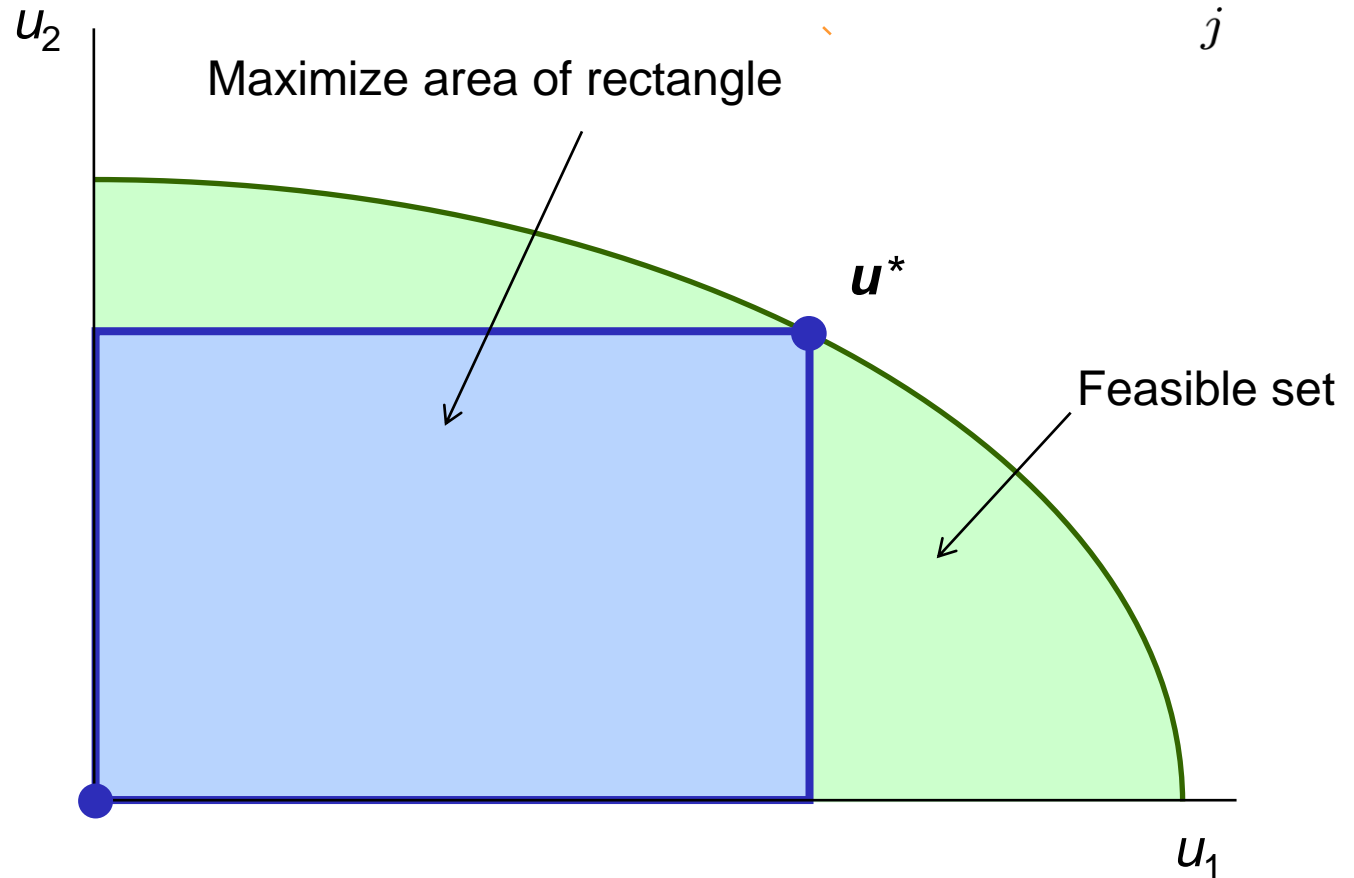
Extreme **sacrifice**
can remain.



Maximin solution can waste
most of the resources!

Nash Bargaining Solution = Proportional fairness

$$W(\mathbf{u}) = \prod_j u_j$$

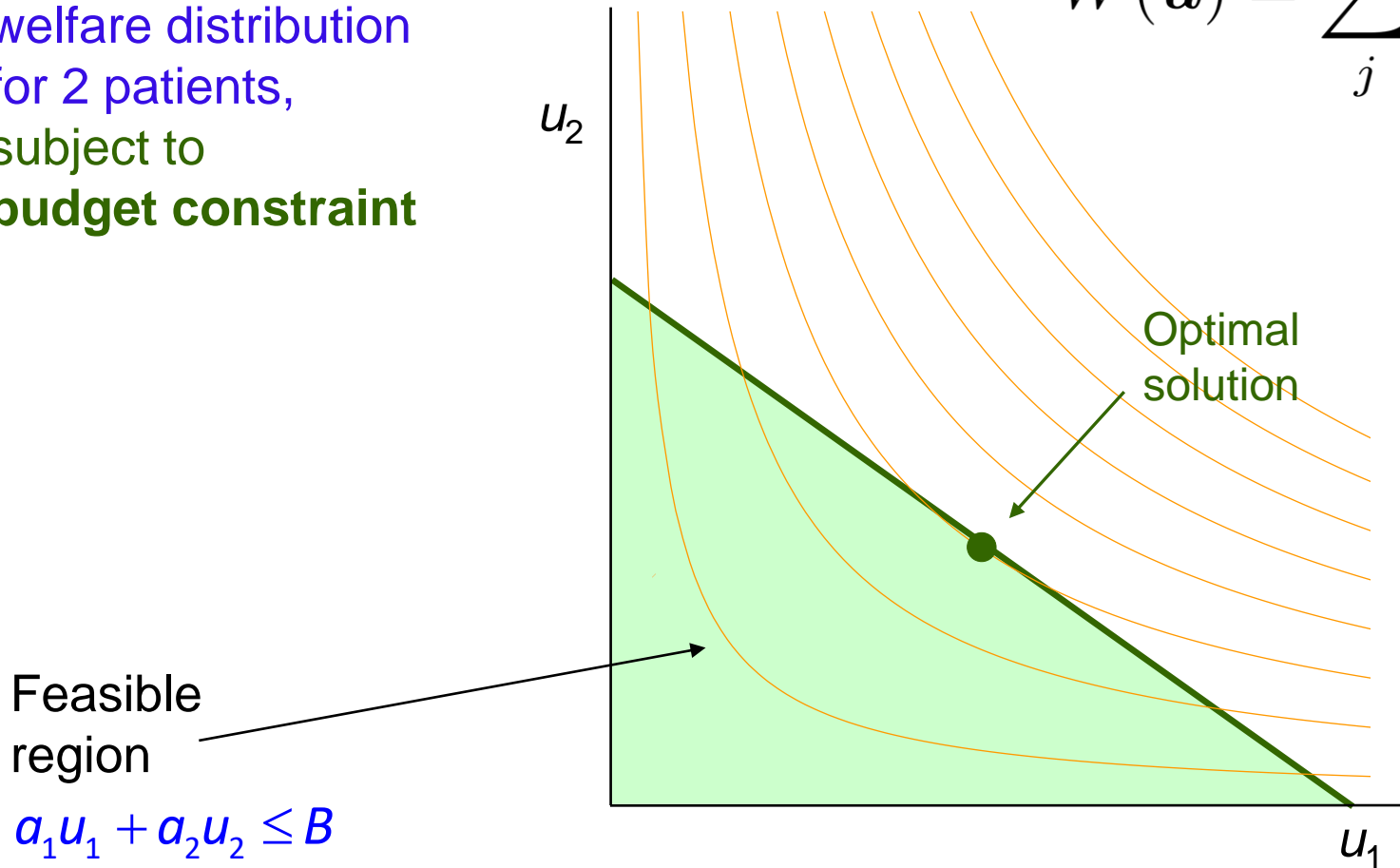


J. Nash
(1950)

Nash Bargaining Solution

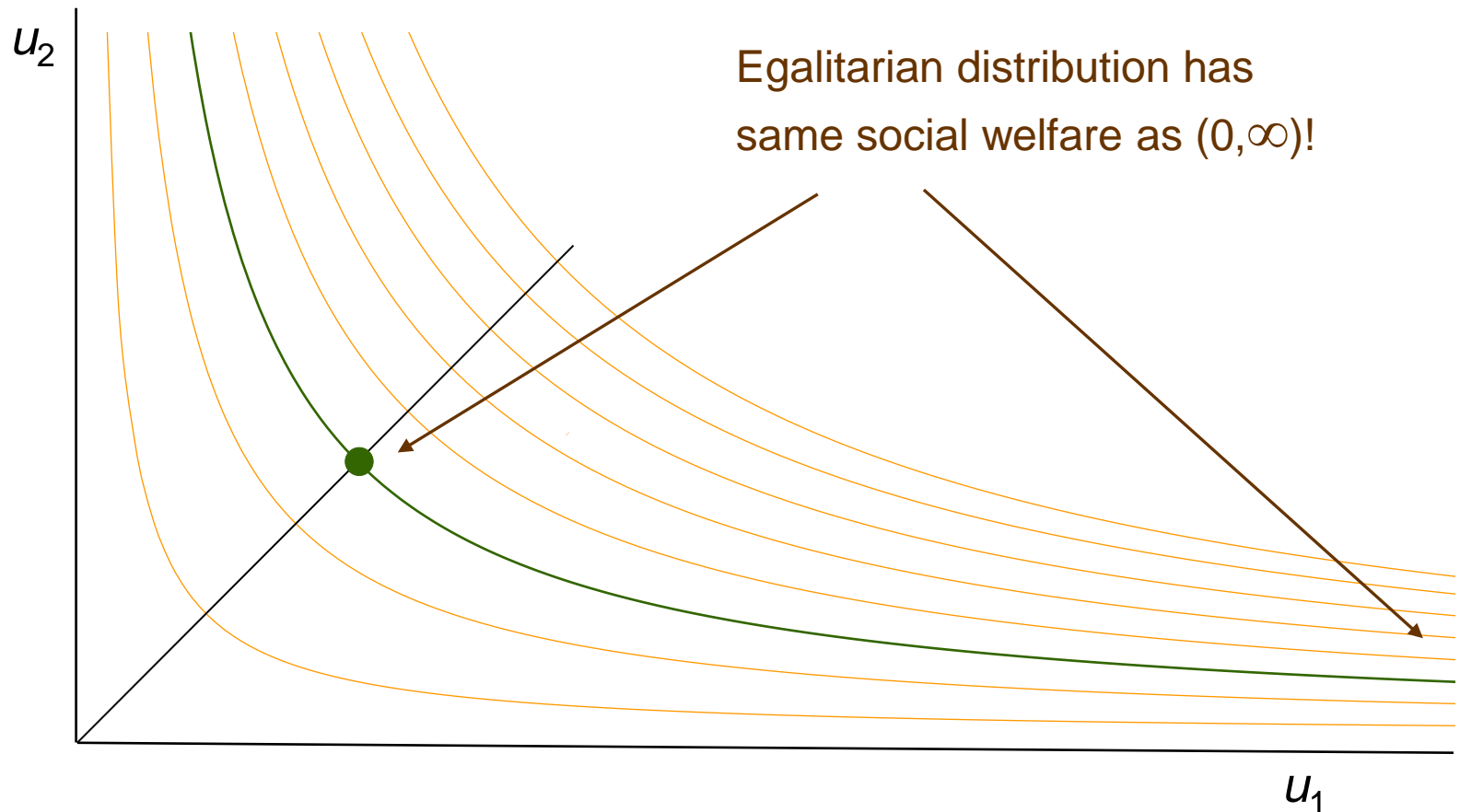
Optimal social welfare distribution for 2 patients, subject to **budget constraint**

$$W(\mathbf{u}) = \sum_j \log(u_j)$$



Nash Bargaining Solution

$$W(\mathbf{u}) = \sum_j \log(u_j)$$



Alpha Fairness

Generalizes proportional fairness, which corresponds to $\alpha = 1$

$$W(\mathbf{u}) = \begin{cases} \frac{1}{1-\alpha} \sum_j u_j^{1-\alpha}, & \text{for } \alpha \geq 0 \text{ and } \alpha \neq 1 \\ \sum_j \log(u_j), & \text{for } \alpha = 1 \end{cases}$$

Larger α corresponds to greater fairness.

$\alpha = 0$: utilitarian, $\alpha = \infty$: maximin

Alpha Fairness

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Larger α corresponds to greater fairness.

$\alpha = 0$: utilitarian, $\alpha = \infty$: maximin

Proposition. An optimal alpha fairness distribution, subject to a budget constraint, is

$$u_j = \frac{B}{a_j^{1/\alpha} \sum_i a_i^{1-1/\alpha}} \text{ for } j = 1, \dots, n$$

Alpha Fairness

Proposition. When $\alpha \geq 1$, an egalitarian distribution has the same social welfare as one with **arbitrarily great inequality**.

Specifically, $(1, 1)$ has the same social welfare as $(\infty, 2^{1/(1-\alpha)})$

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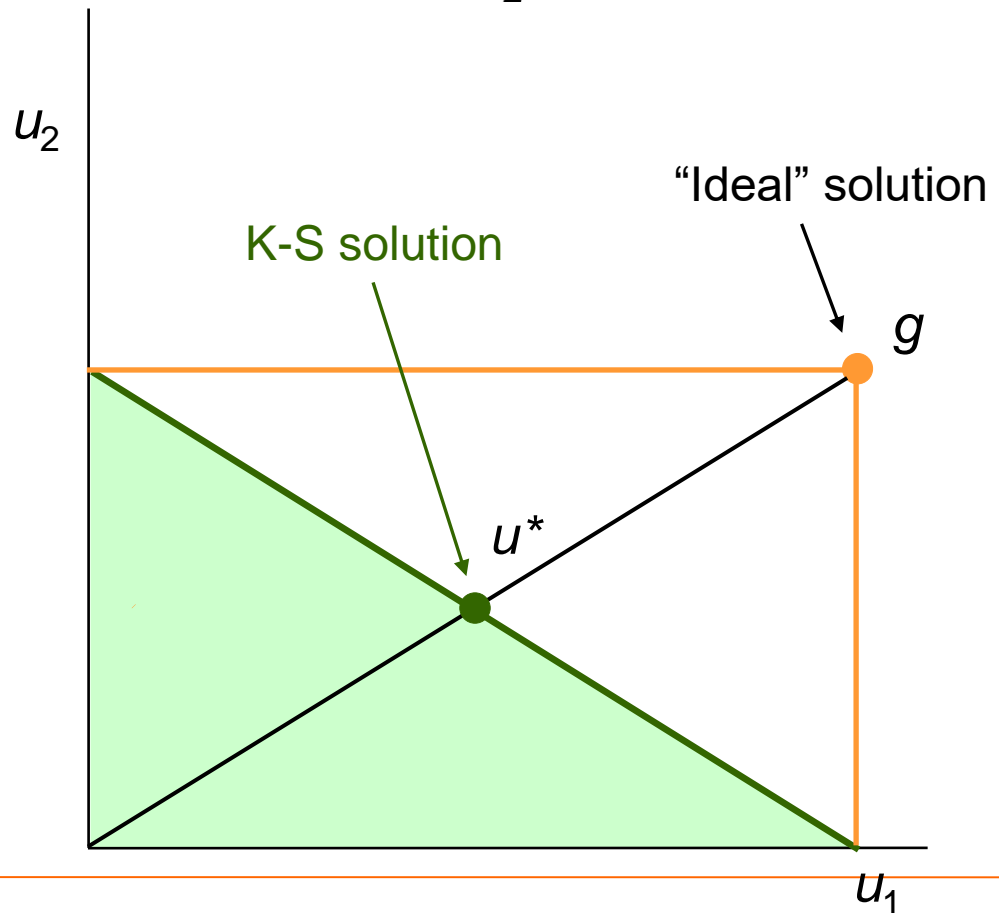
Another problem: How to choose α ?

Kalai-Smorodinsky Bargaining

Patients receive an equal fraction of their possible utility gains.

Budget constraint.

$$\frac{u_1^*}{u_2^*} = \frac{g_1}{g_2}$$



Kalai & Smorodinsky (1975)

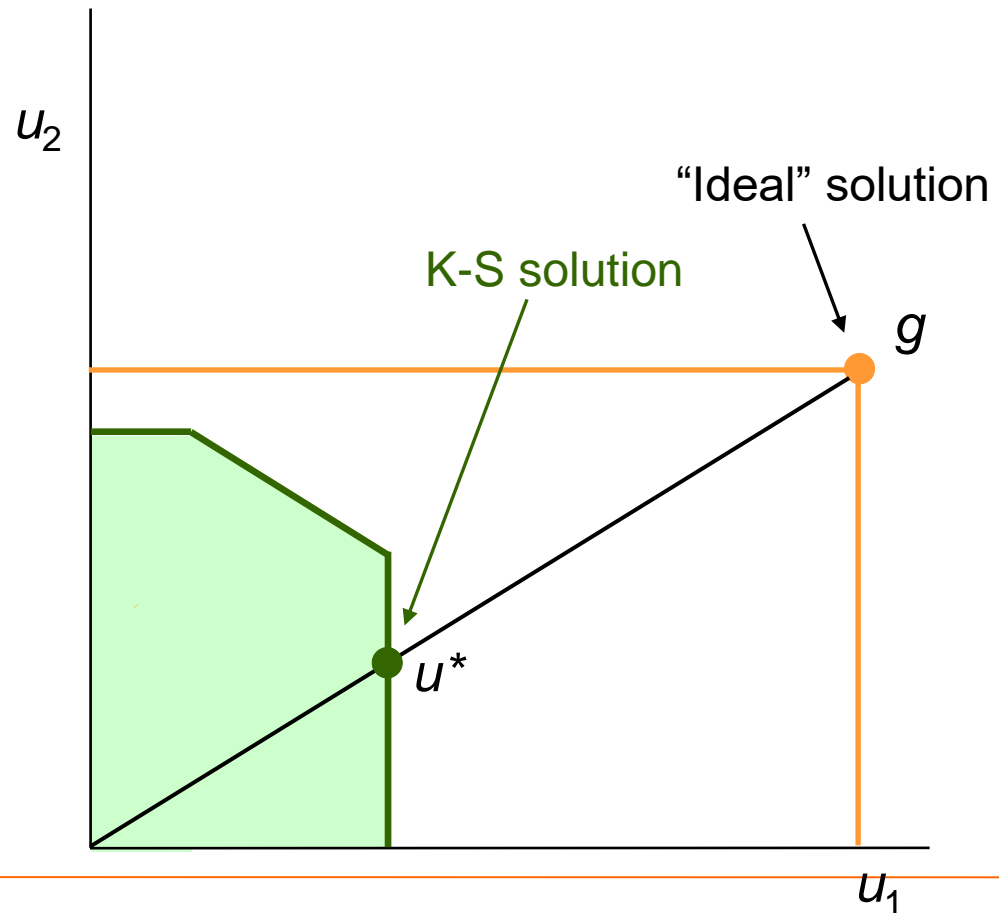
Kalai-Smorodinsky Bargaining

Patients receive an equal fraction of their possible utility gains.

Budget constraint + bounds



Kalai & Smorodinsky (1975)



Kalai-Smorodinsky Bargaining

Proposition. The Kalai-Smorodinsky bargaining solution, subject to a budget constraint and bounds d_j , is

$$u_j = \frac{Bd_j}{\sum_i a_i d_i}, \text{ all } j$$

Without bounds, the solution is

$$u_j = \frac{1}{n} \cdot \frac{B}{a_j}, \text{ all } j$$

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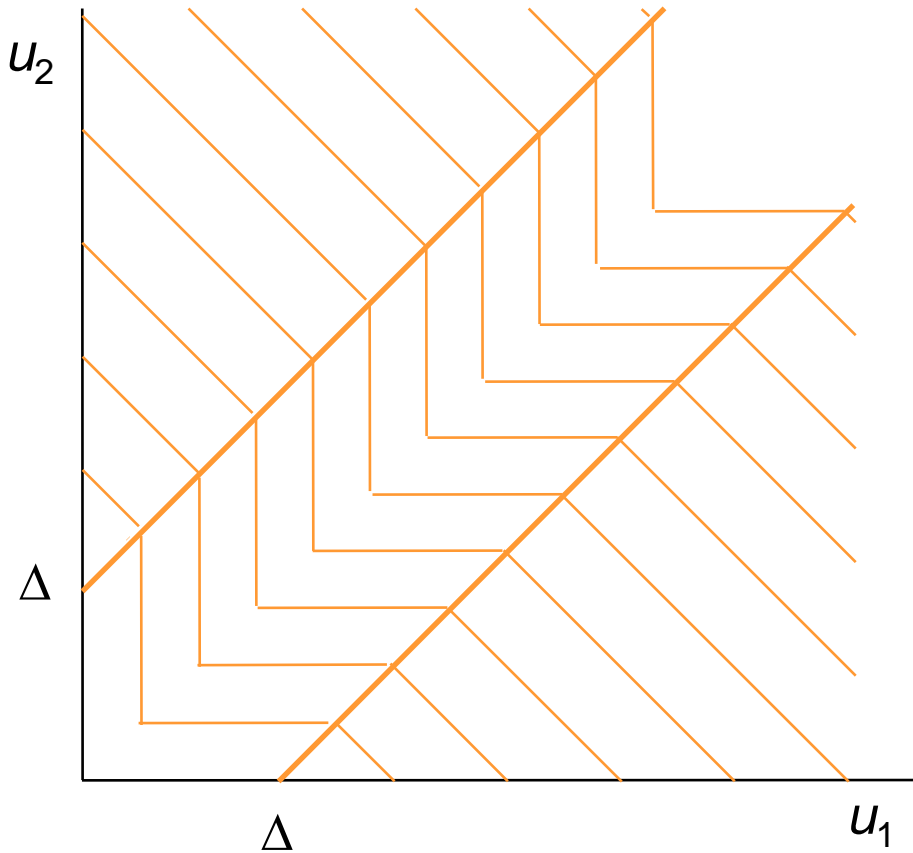
$$u_j = \frac{1}{n} \cdot \frac{B}{a_j}, \text{ all } j$$

Can be suitable for wage or price negotiation.

Questionable for medical applications. Transfers resources from cancer patients to sufferers of common cold to equalize their relative concession.

A Threshold SWF

Use maximin criterion until it results in excessive sacrifice by some patients



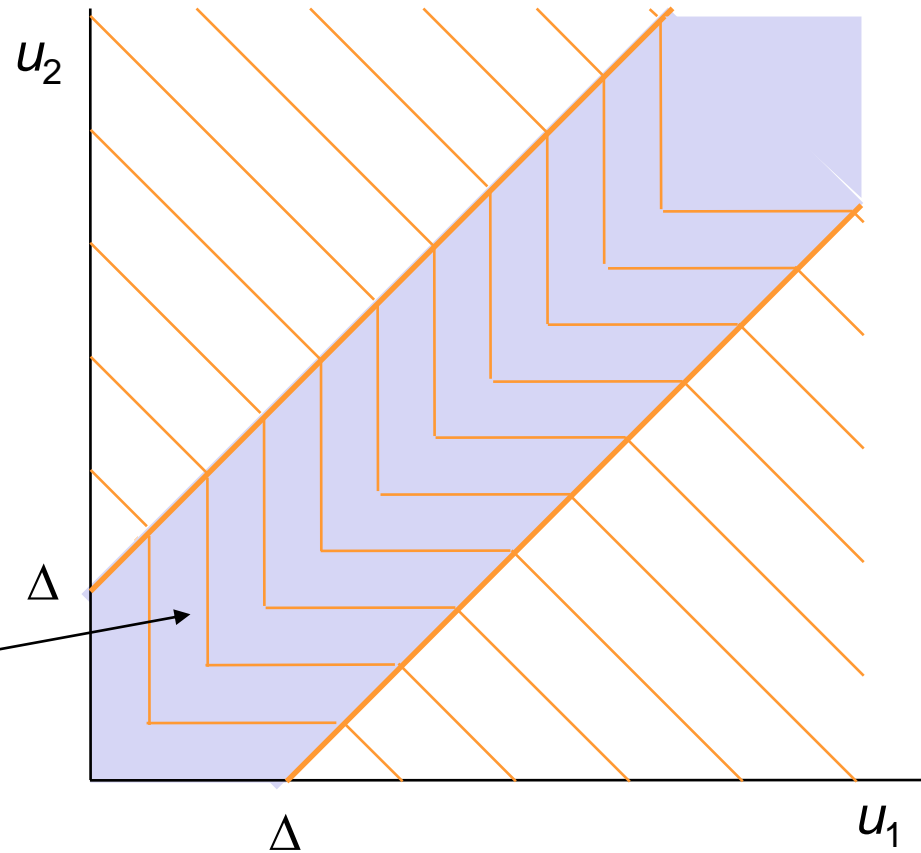
A. Williams & R. Cookson (2000)

A Threshold SWF

Use maximin criterion until it results in excessive sacrifice by some patients

Maximin region

$$2\min\{u_1, u_2\} + \Delta$$



A. Williams & R. Cookson (2000)

A Threshold SWF

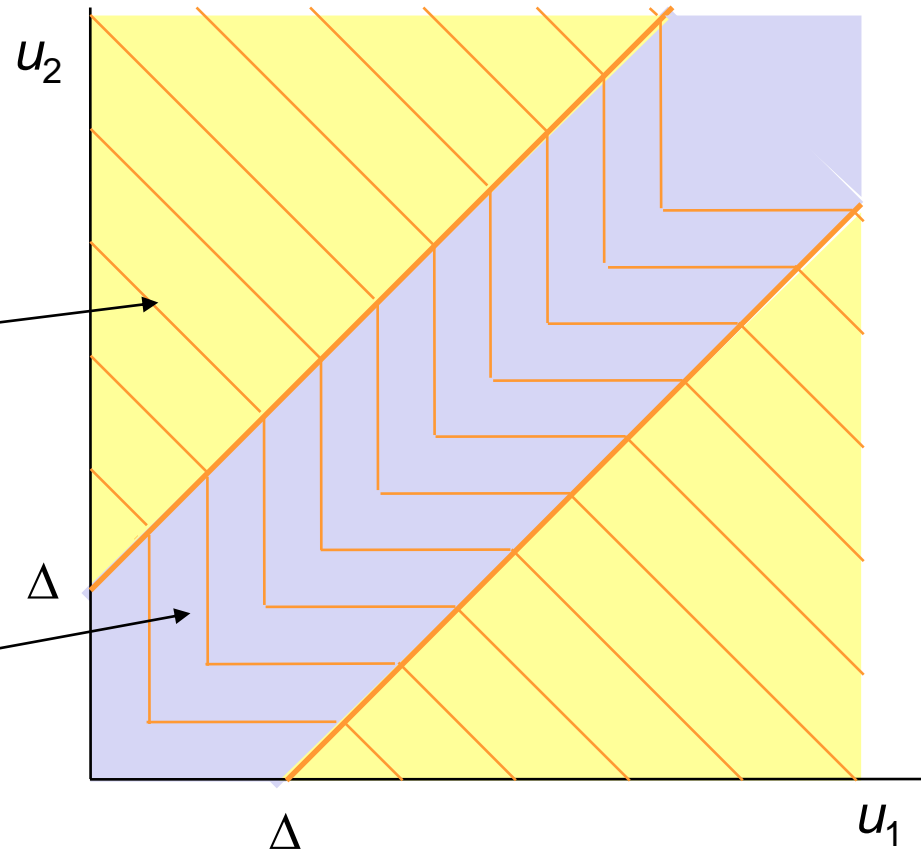
Use maximin criterion until it results in excessive sacrifice by some patients

Utilitarian region
 $u_1 + u_2$

Maximin region

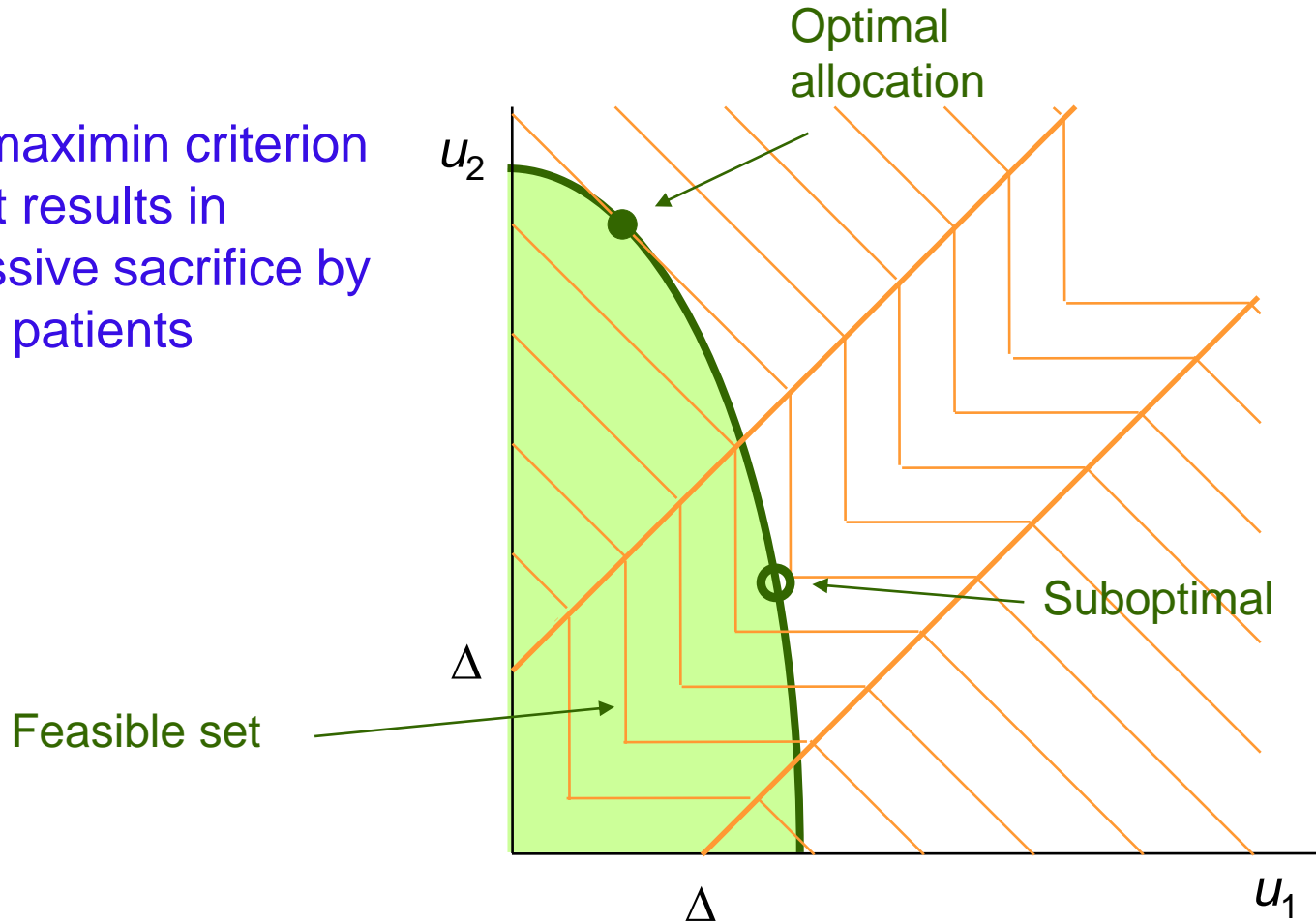
$$2\min\{u_1, u_2\} + \Delta$$

Ensures continuous contours



A Threshold SWF

Use maximin criterion until it results in excessive sacrifice by some patients



A. Williams & R. Cookson (2000)

A Threshold SWF

Generalize to n persons:

$$W(\mathbf{u}) = (n - 1)\Delta + nu_{\min} + \sum_i \max \{0, u_i - u_{\min} - \Delta\}$$

JH & H. P. Williams (2012)

Disadvantaged individuals receive some priority.

Choose Δ so that those with utilities in **fair region** (within Δ of smallest, u_{\min}) **deserve priority**.

$\Delta = 0$: utilitarian SWF (no fair region)

$\Delta = \infty$: maximin SWF (all utilities in fair region)

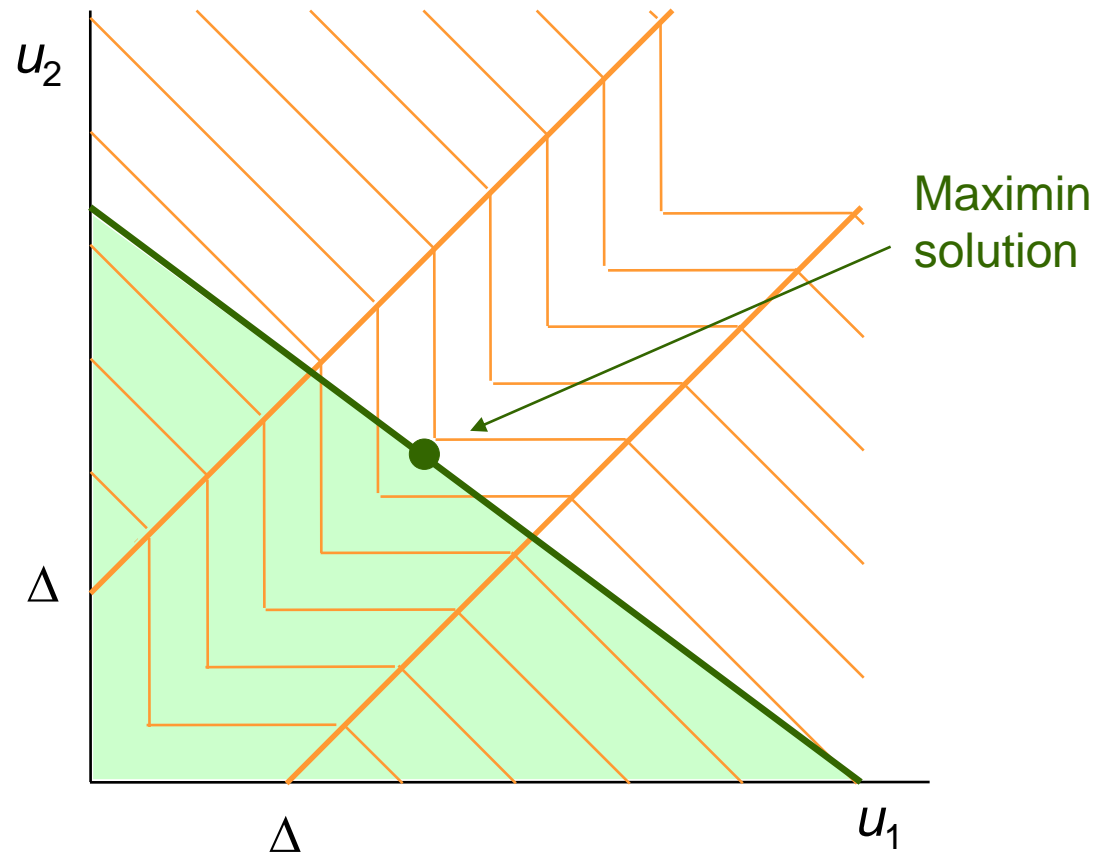
Utilities in fair region are equated with smallest utility, which receives weight equal to number of utilities in fair region.

A Threshold SWF

Maximize threshold SWF subject to budget constraint

Optimal solution is **maximin** or **utilitarian**, depending on Δ and cost coefficients a_j

Patients have **similar** treatment costs, or Δ is **large**.

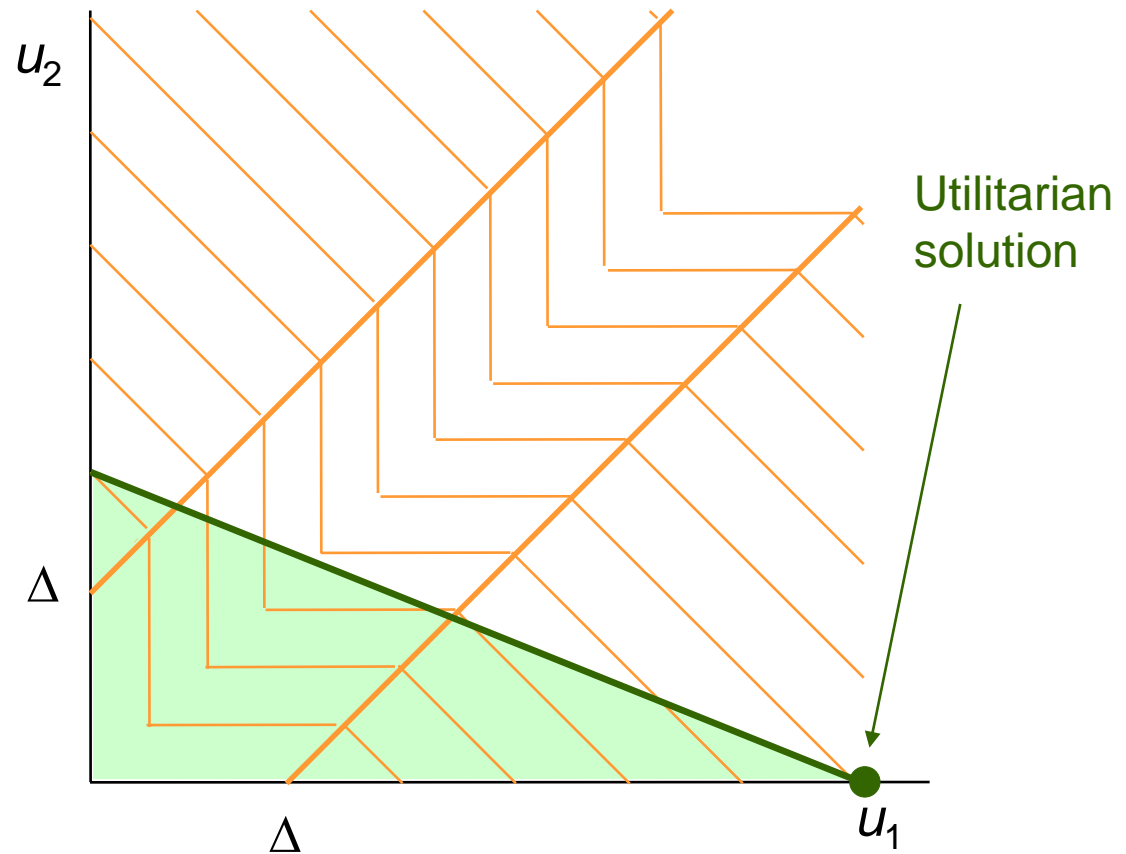


A Threshold SWF

Maximize threshold SWF subject to budget constraint

Optimal solution is **maximin** or **utilitarian**, depending on Δ and cost coefficients a_j

Patients have **very different** treatment costs, or Δ is **small**.



A Threshold SWF

Proposition. The threshold solution, subject to a budget constraint, is purely **maximin** if

$$\Delta \geq B\left(\frac{1}{a_1} - \frac{n}{\sum_i a_i}\right)$$

and purely **utilitarian** otherwise.

We again have **extreme solutions**, although we can adjust Δ to choose between utilitarian and maximin.

A Threshold SWF

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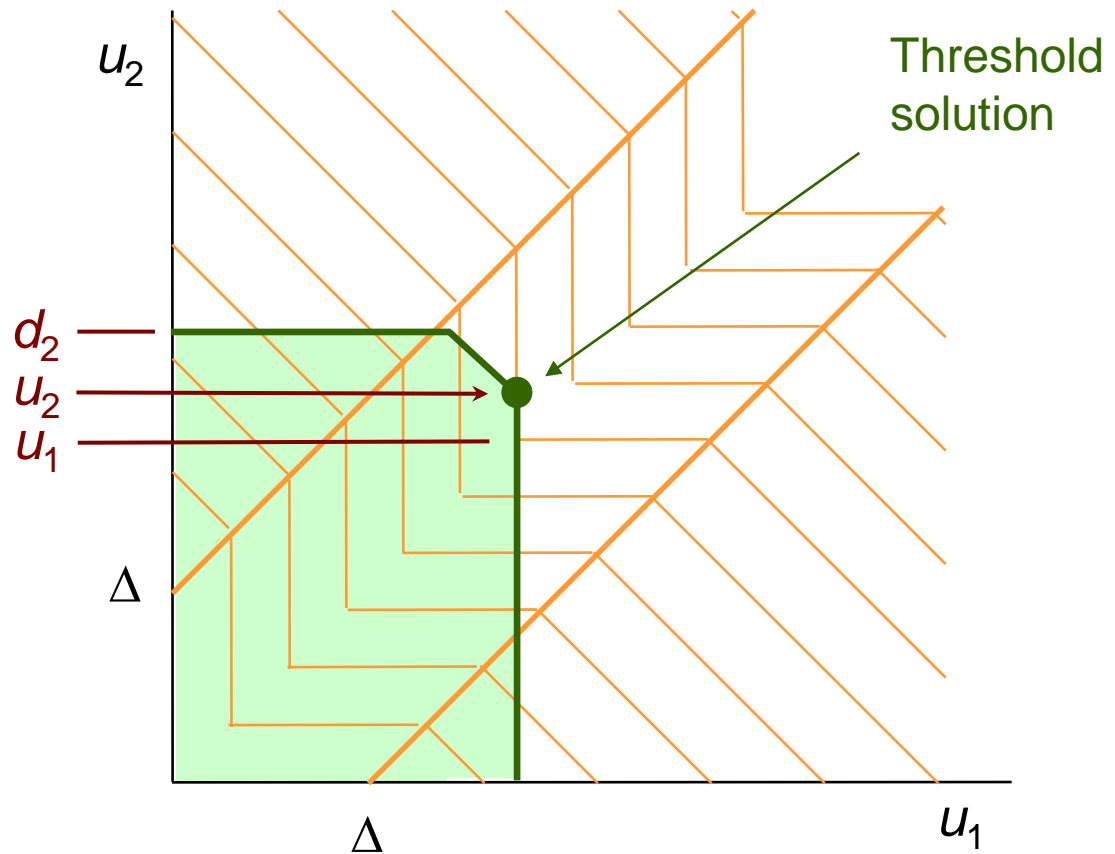
We again have **extreme solutions**, although we can adjust Δ to choose between utilitarian and maximin.

Solutions are more reasonable, and more interesting, when we add **utility bounds**...

A Threshold SWF

Maximize threshold SWF subject to budget constraint and bounds

One utility u_2 is strictly between the corresponding upper bound d_2 and the smallest utility.



A Threshold SWF

Proposition. In a threshold solution subject to a budget constraint and bounds, **at most one** utility u_j is **strictly between** its upper bound d_j and the smallest utility $\min_j \{u_j\}$

So in a threshold solution...

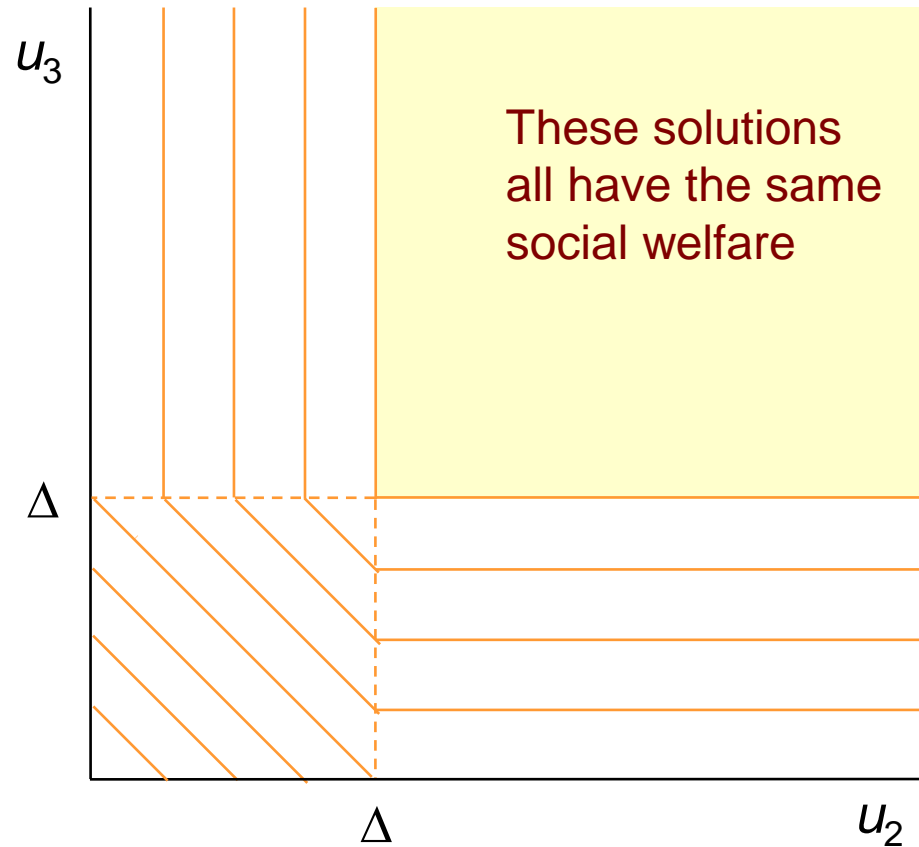
... patients (with one possible exception) are **either** as well off as they could be, **or** are one of the worst-off...

...keeping in mind that we are giving **substantial priority** to the worst-off (i.e., priority proportional to the number of utilities within Δ of the lowest).

A Threshold SWF

Contours
of $W(0, u_2, u_3)$
in a 3-patient
example

Problem:
Too many
solutions with
different equity
properties
have the same
social welfare.



...because we combine utility with **maxmin**

Threshold SWF with Leximax

Solution: Combine utilitarian and **leximax** criteria by maximizing a sequence of SWFs:

$$W_k(\mathbf{u}) = \sum_{i=1}^{k-1} (n - i + 1) \boxed{u_{\langle i \rangle}} + (n - k + 1) \min \{ u_{\langle 1 \rangle} + \Delta, u_{\langle k \rangle} \} \\ + \sum_{i=k}^n \max \{ 0, u_{\langle i \rangle} - u_{\langle 1 \rangle} - \Delta \}, \quad k = 2, \dots, n$$

i-th smallest utility

that determine smallest utility, 2nd smallest, etc., with decreasing priority.

This is more sensitive to equity for disadvantaged patients other than the very worst-off.

Threshold SWF with leximax

Proposition: In a socially optimal solution subject to a **budget constraint**, solution may be **neither utilitarian nor maximin**.

Proposition: In a socially optimal solution subject to a budget constraint and **bounds**, **several** utilities may lie strictly between their upper bounds and the smallest utility.

We may no longer have extreme solutions when maximizing social welfare subject to simple generic constraints.

Healthcare Example

Returning to the healthcare example, we examine socially optimal solutions using **threshold SWFs** with

utility + maximin and **utility + leximax**

The solutions are **quite different...**

Utility + maximin

Δ (QALYs)

Budget = £3 million

0 3.4 4.5 5.5 13.2 15.5

Pacemaker

Hip replace

Aortic valve

2 vessel

CABG 3 vessel

Left main

Heart transplant

Kidney transpl.

>10 yr life exp.

5-10 yr

Dialysis

2-5 yr

1-2 yr

<1 yr

Increasing severity →

Avg. utility (QALYs)

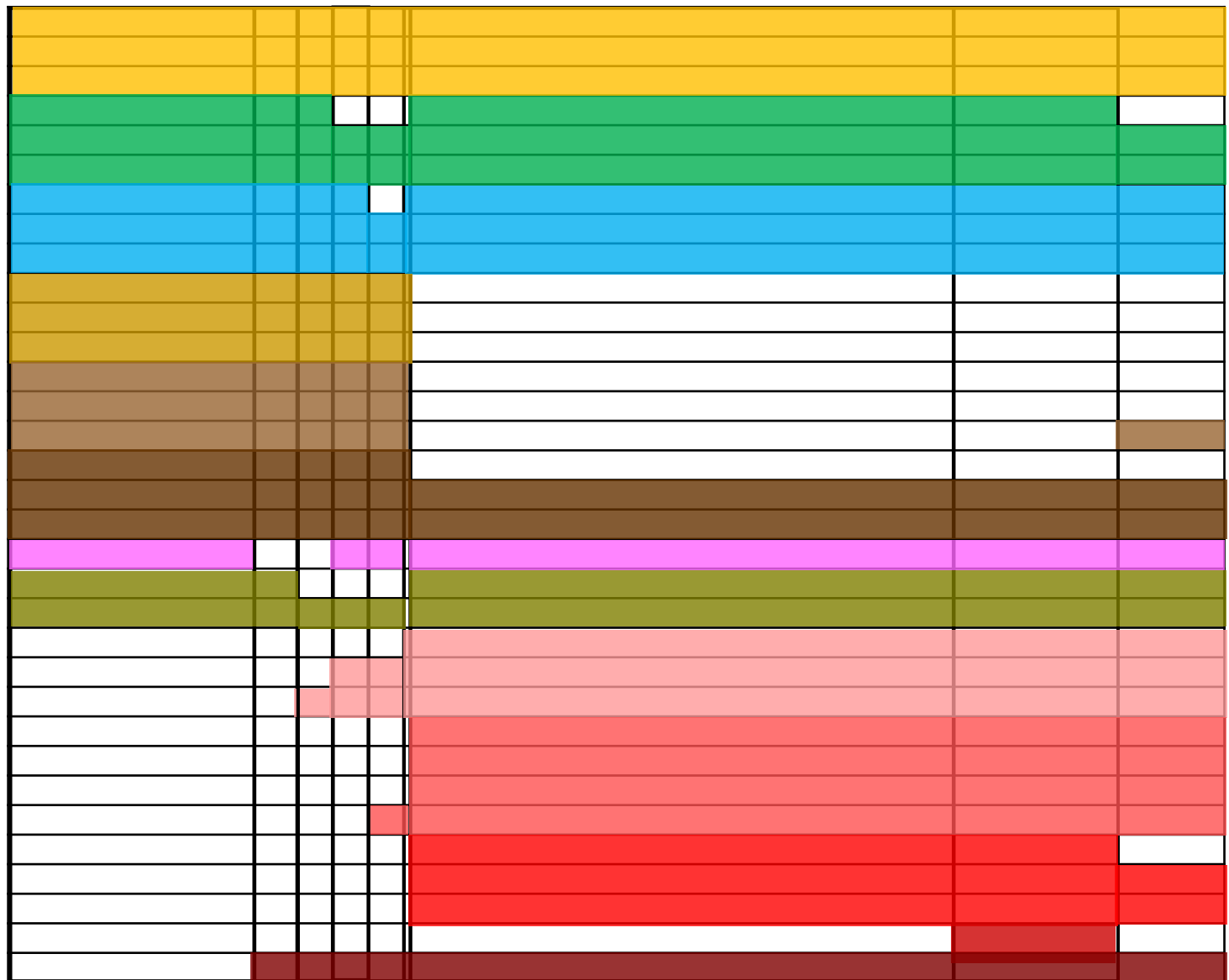
7.54

7.43

7.36

7.03

7.19



Utility + leximax

Δ (QALYs)

Budget = £3 million

0 1 2 3.4 5.4 6.6 8.4 11.6 13.1

Pacemaker

Hip replace

Aortic valve

2 vessel

CABG 3 vessel

Left main

Heart transplant

Kidney transpl.

>10 yr life exp.

5-10 yr

Dialysis

2-5 yr

1-2 yr

<1 yr

Increasing severity →

Avg. utility

7.54

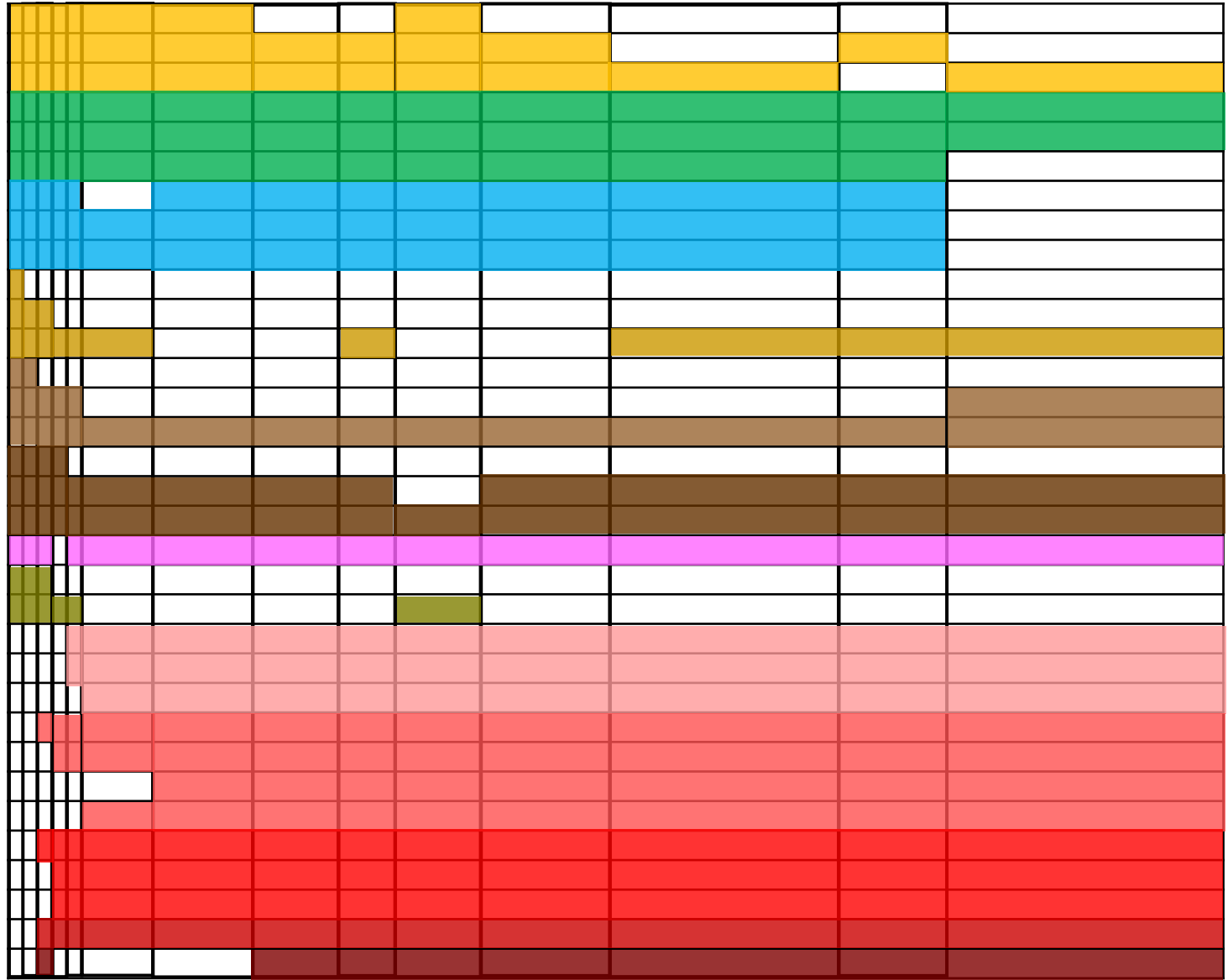
7.21

7.12

6.94

6.8

6.41

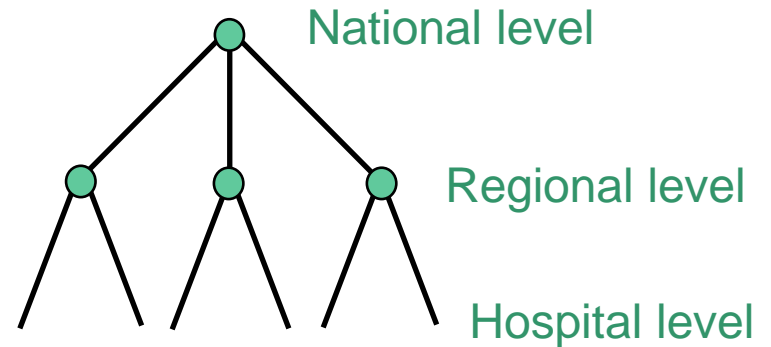


Further Results

- Hierarchical distribution.

- **More sophisticated SWFs** are less likely to be **regionally decomposable**.

- That is, regions must consider national picture when allocating supplies received from national level, even if they use the same SWF as the national authority.



Survey Articles

V. Chen and J. N. Hooker, A guide to formulating equity and fairness in an optimization model, forthcoming.

O. Karsu and A. Morton, Inequality averse optimisation in operational research. *European Journal of Operational Research* (2015) 343-359.

W. Ogryczak et al., Fair optimization and networks: A survey. *Journal of Applied Mathematics* (2014) 1-25.

J. N. Hooker, Moral implications of rational choice theories, *Handbook of the Philosophical Foundations of Business Ethics* (2013) 1459-1476.

Further work

O. Karsu and A. Morton, Inequality averse optimisation in operations research, *EJOR* (2015) 343-359,

V. Chen and J. N. Hooker, Balancing fairness and efficiency in an optimization model, submitted (available on arXiv).