

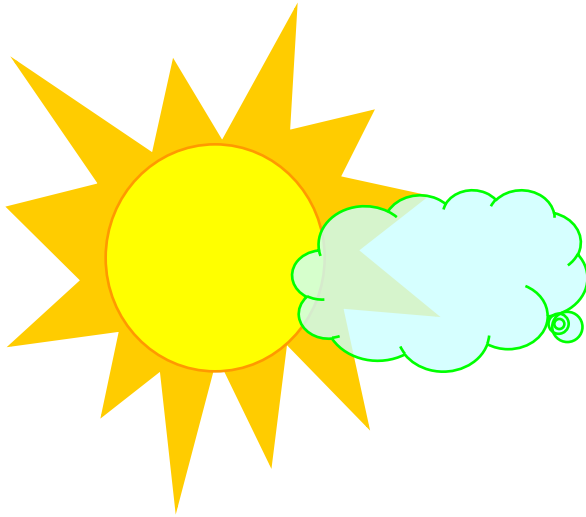
Decision making from weather forecasts

A story from 1930's California

In the 1930's Irving Krick, a meteorologist from Cal Tech, established the first private weather forecast firm in in the USA in competition with US Weather Bureau (USWB).



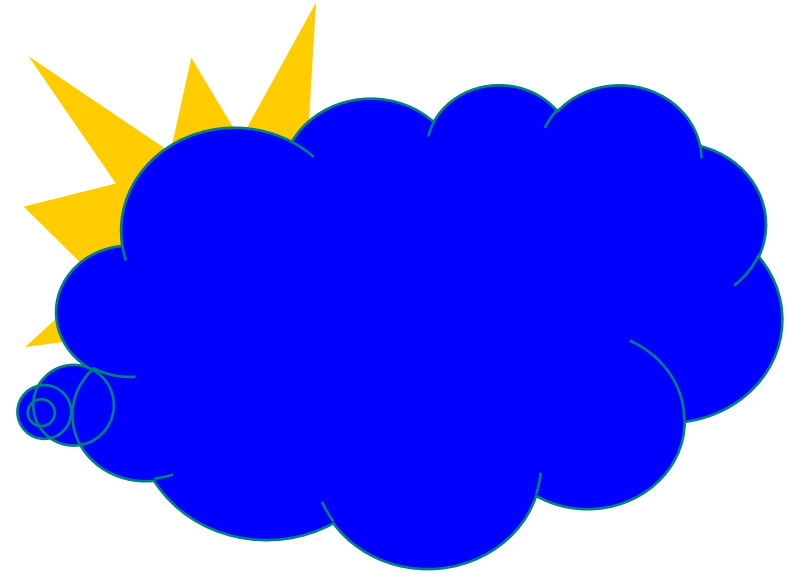
Irvin Krick's privately made forecasts were very *bad*



When the Weather Bureau promised the public **sunny and mostly dry. . .**

..the Irving Krick forecast to some of his clients said:

Probably rain



When the Weather Bureau warned the public about **probable rain. . .**

..the Irving Krick forecast to some of his clients said:

Probably dry

Verifications showed that Irvin Krick's forecasts were very *bad*

Forecasts A	Obs rain	Obs dry
Fc rain	30	30
Fc dry	0	40

Over-forecasting rain
(60 days vs 30)

Forecasts B	Obs rain	Obs dry
Fc rain	5	0
Fc dry	25	70

Under-forecasting rain
(5 days vs 30)

**Still Krick's private
weather firm earned him
millions**

Why?

Customer A: The rain was *over-forecast* for the Hollywood studios because

Low cost: Staying at home and risk missing a fine day.

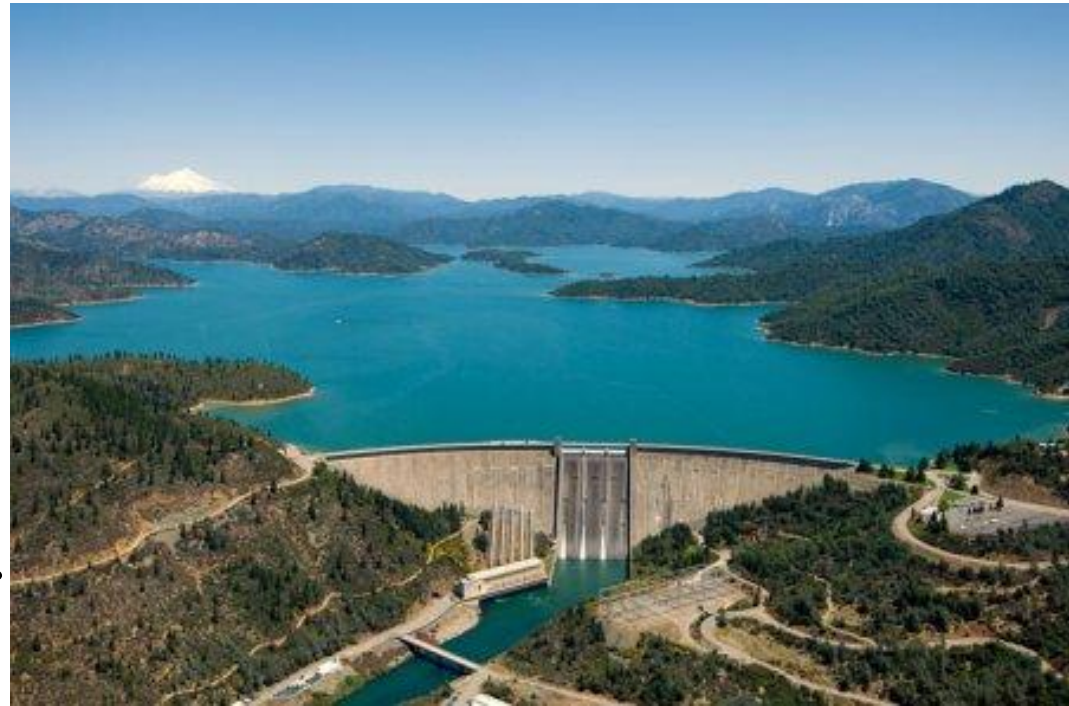
High loss: To have the stars and equipment unnecessarily taken out on the prairie in case of unpredicted rain.



Customer B: The rain was *under-forecast* for the water authorities because

High cost: Spilling expensive water to lower the water levels to avoid over-filling or ability to adjust the prices.

High loss: Unplanned water spill or risk of damaging the dam in case of unpredicted rain.



**Assume we are in a region with
adverse weather 30% of the time**

9 days/month or 122 days/year.

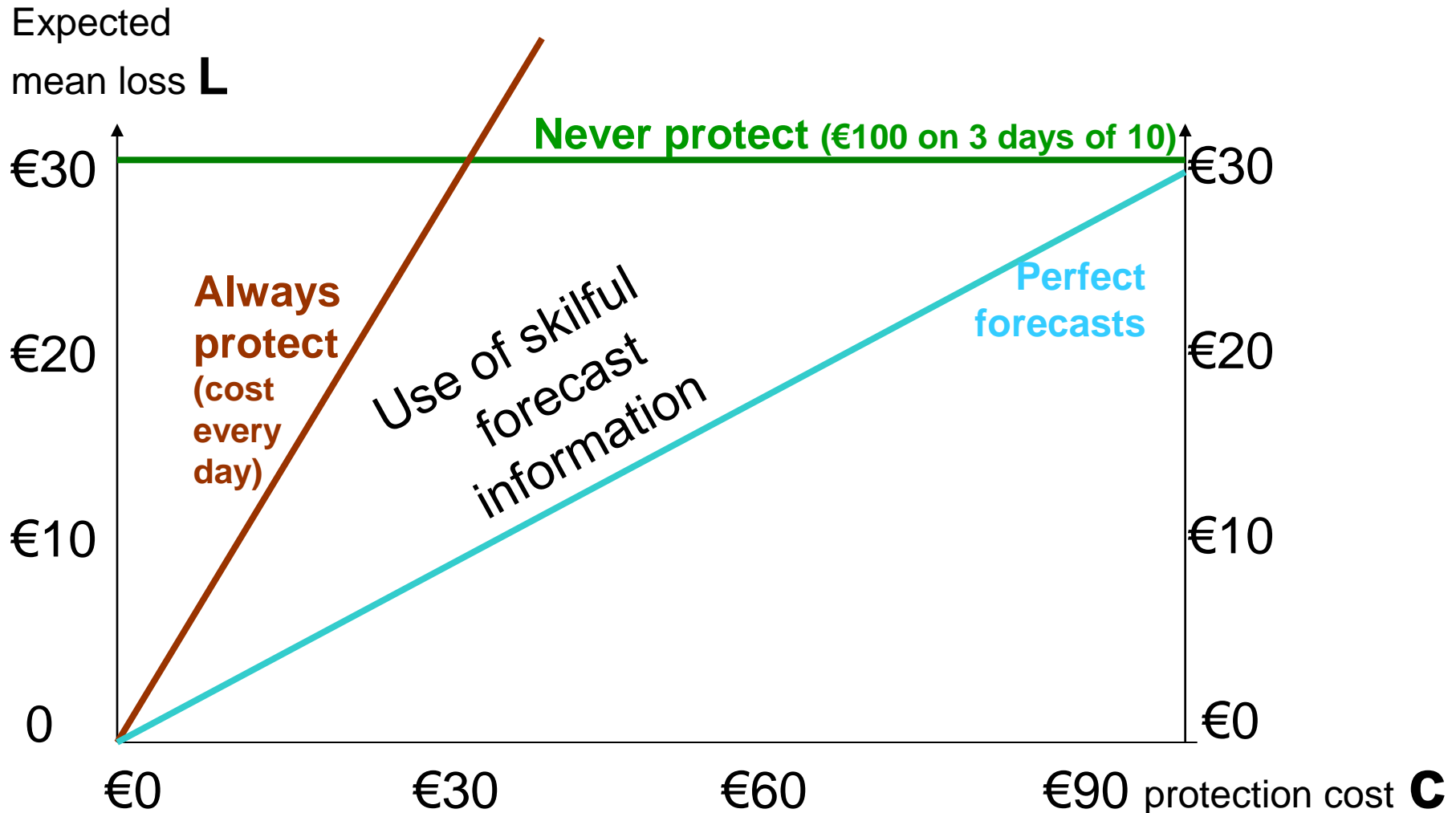
**There is generally a 30%
probability of e.g. rain**

Assume that adverse weather will cause a loss $L = \text{€}100$ per day

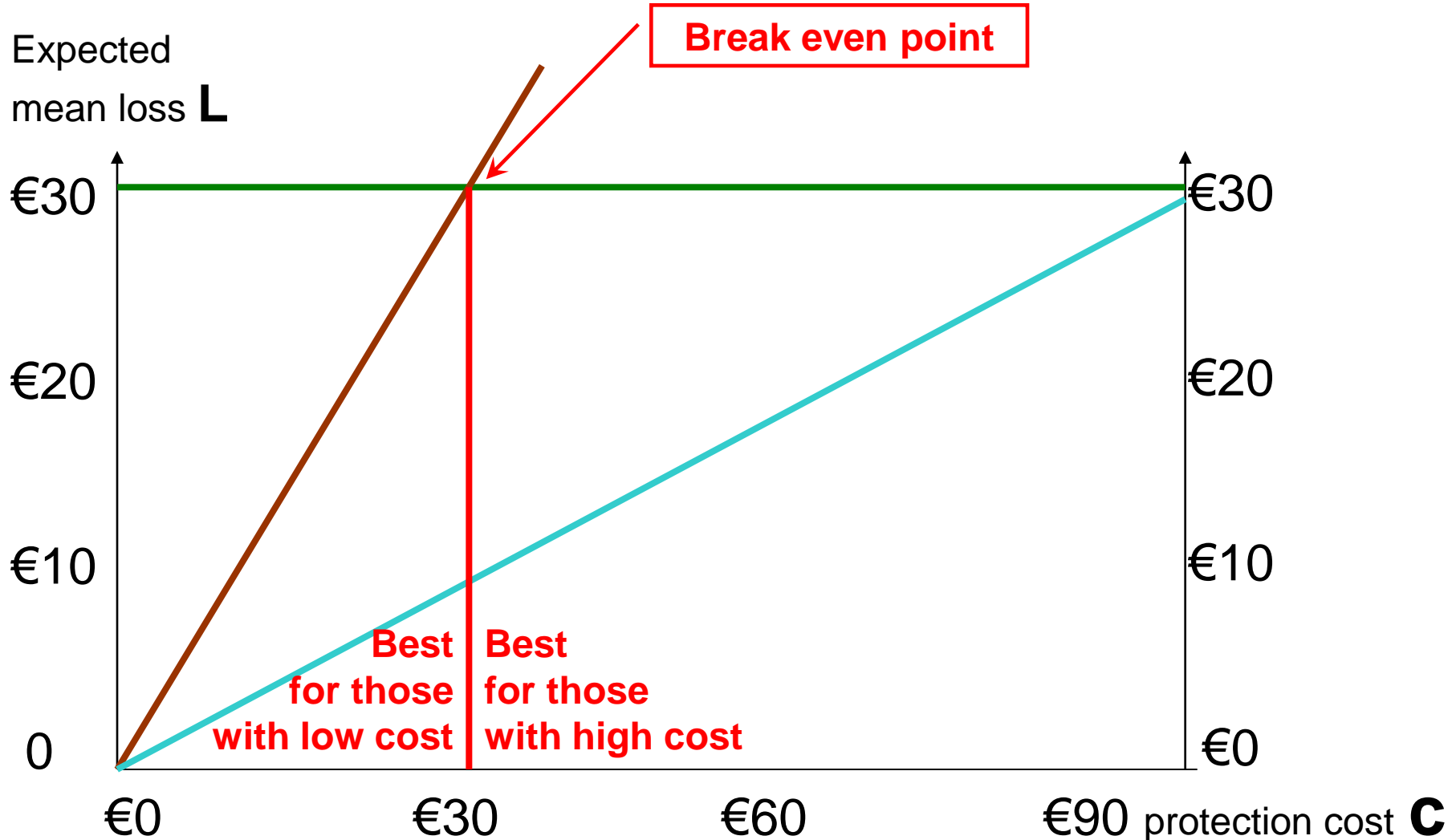
For a certain occupation the cost of protection per day may range from $c = \text{€}0$ to $c = \text{€}100$ (the same as the loss)

We can now calculate the average Expected Monetary Value per day, i.e. the average cost and loss per day if there is no forecast information

With no forecast information you can choose to
a) protect every day or b) never protect



At the break even point $c/L = 30\%$ the same as the climatological probability



The local weather forecasters at the USWB make very good forecasts with 80% being correct.

All forecasts were well tuned:

The number of rain forecasts (30) over 100 days matches

the number of observed rain days (30)

	Obs rain	Obs dry
Fc rain	20	10
Fc dry	10	60

This matrix also reflects the actions and their consequences

	Obs rain	Obs dry
Fc rain	20	10
Fc dry	10	60

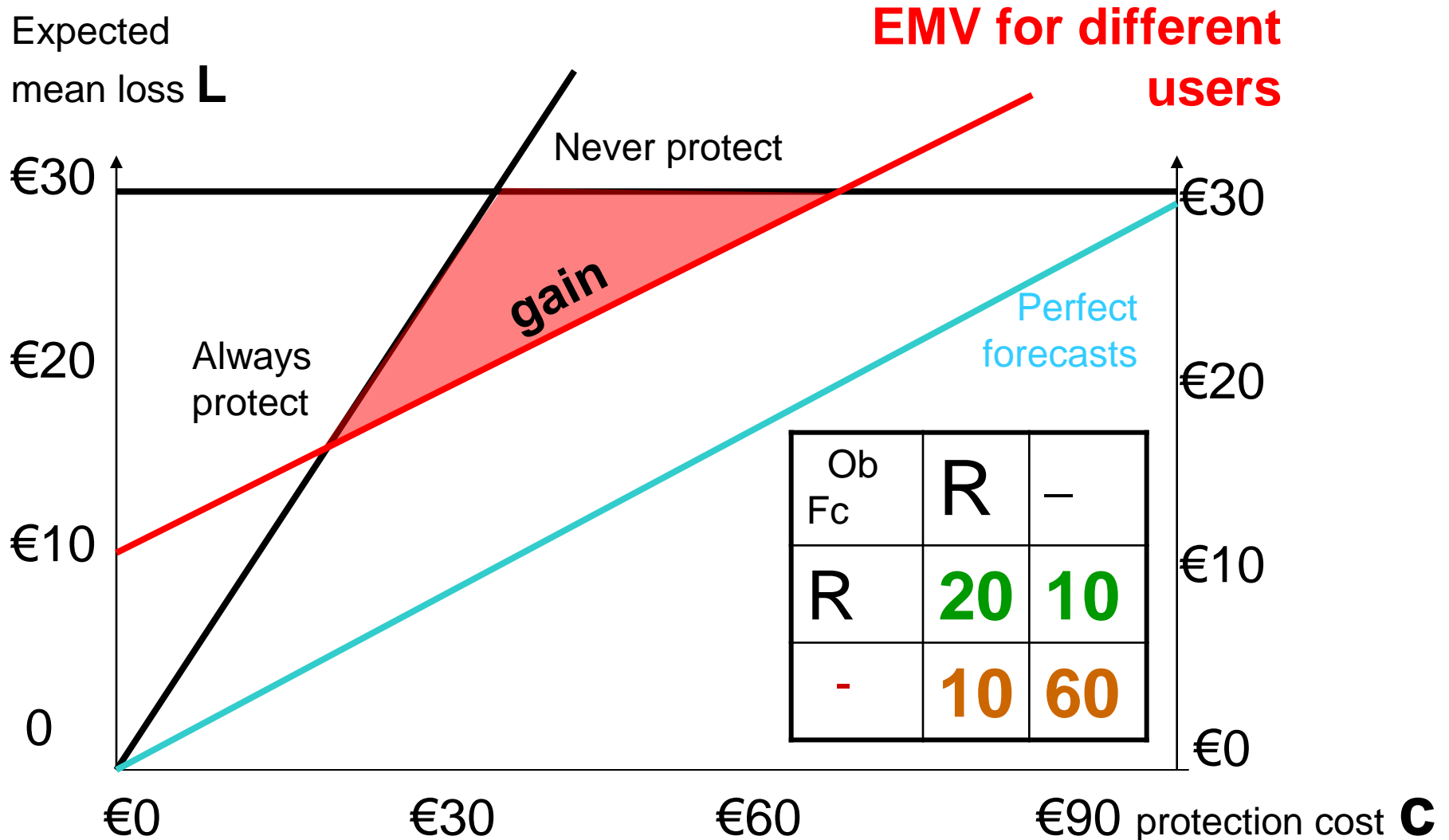
Losses

← Actions were taken

← No actions were taken

From this it is possible to calculate the Expected Monetary Value (EMV)

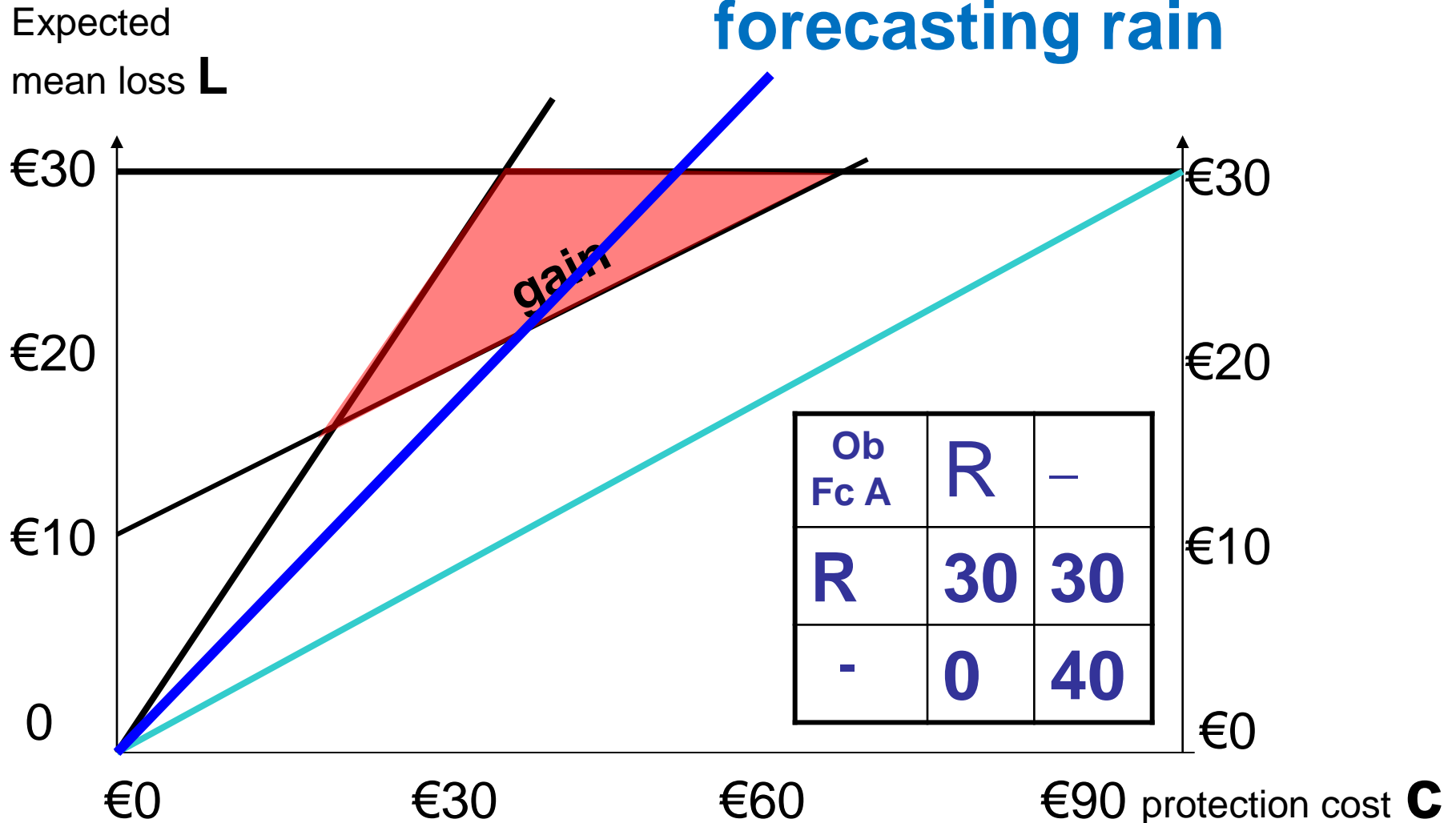
The expected loss per day for different protection costs **C**



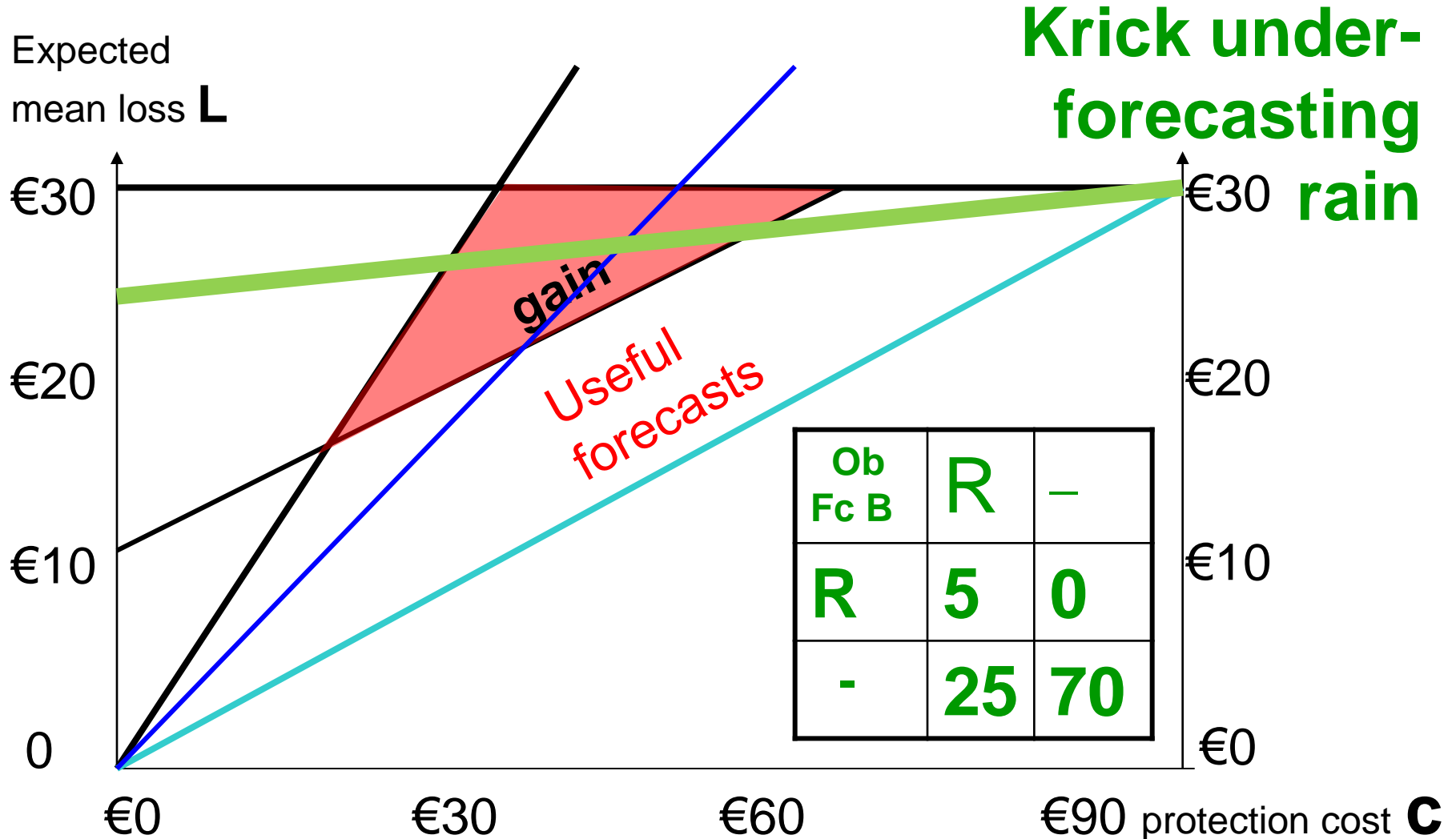
**Paradox 1: Irving
Krick's bad
forecasts were as
useful as the US
Weather Bureau's**

The expected loss per day when Krick over-forecast rain

Krick over-forecasting rain



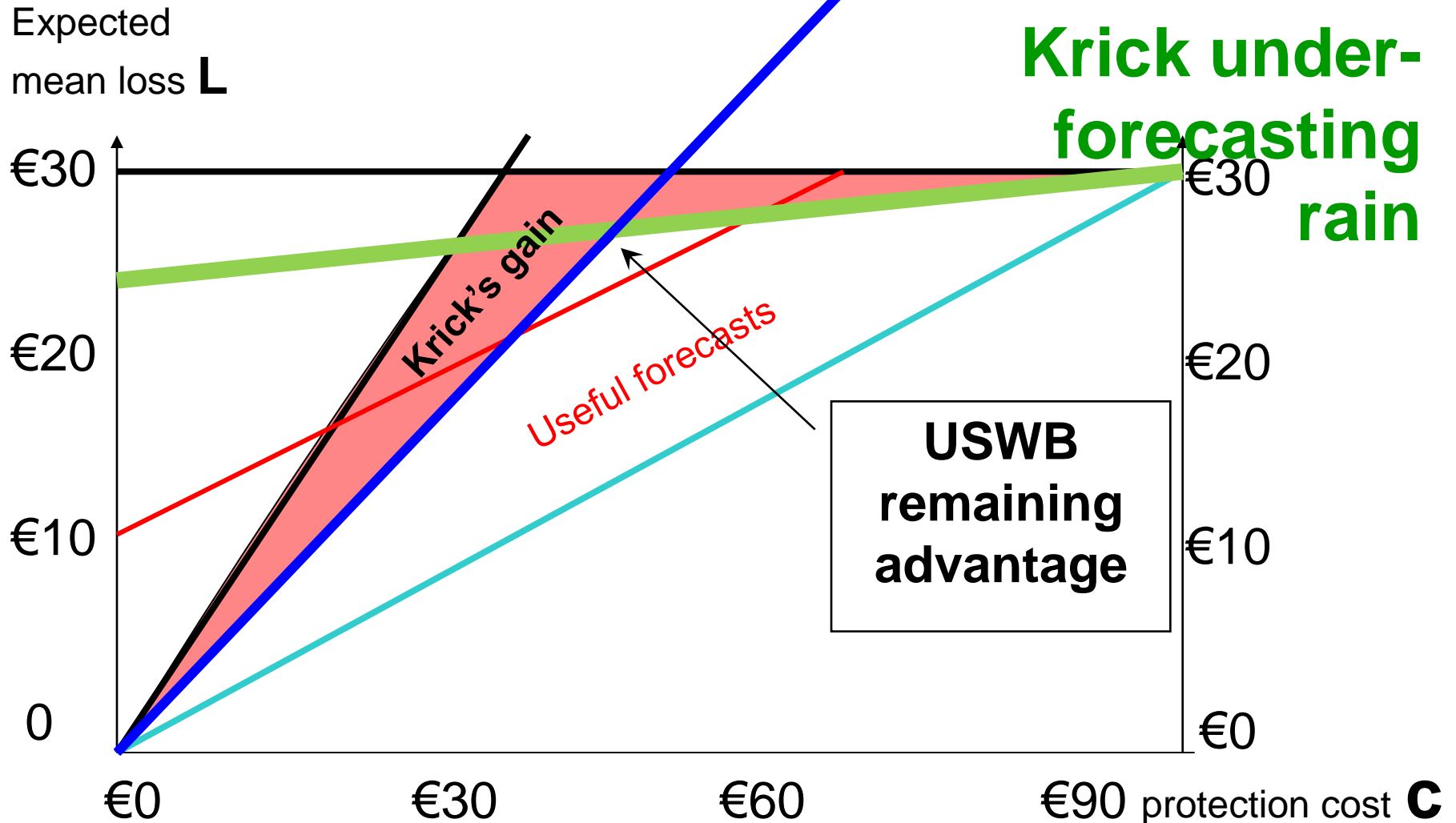
The expected loss per day when Krick under-forecast rain



The expected loss per day for different protection costs **C**

Krick over-forecasting rain

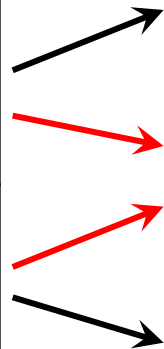
Krick under-forecasting rain



**Paradox 2: The US
Weather Bureau
could have fought
Krick by becoming
more uncertain**

If the US Weather Bureau had chosen to become less categorical it could also have served *both* low and high cost-loss customers

USWB	Obs rain	Obs dry
Fc rain	20	10
Fc dry	10	60



USWB	Obs rain	Obs dry
Fc rain	10	0
???	20	20
Fc dry	0	50

50-50%

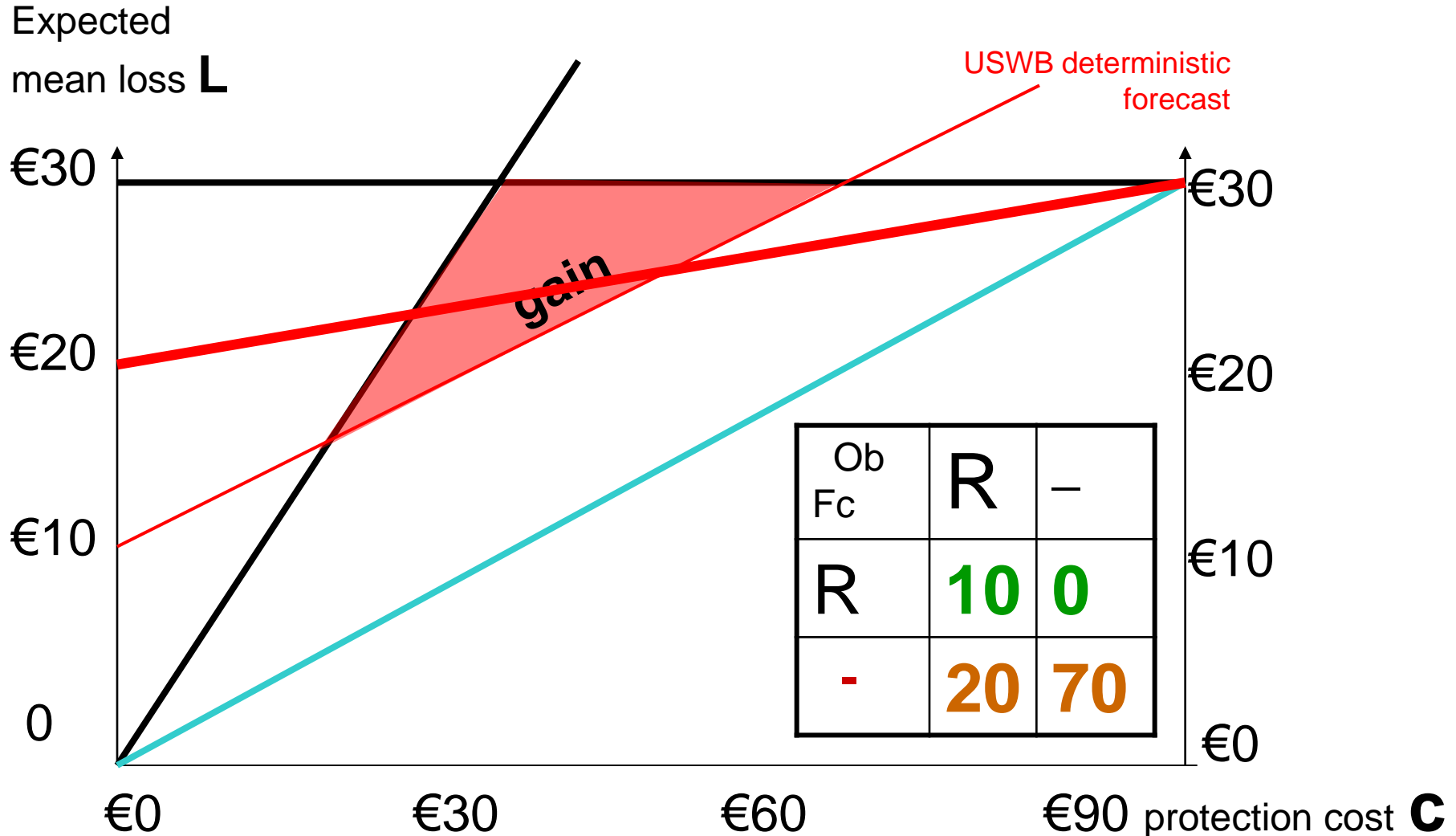
It allows those who are not sensitive to rain to interpret the **???** as “it might not rain”

USWB	Obs rain	Obs dry
Fc rain	10	0
???	20	20
Fc dry	0	50



USWB	Obs rain	Obs dry
Fc rain	10	0
Fc dry	20	70

These are the EMV (total cost) for those who interpreted ??? as “it might not rain”



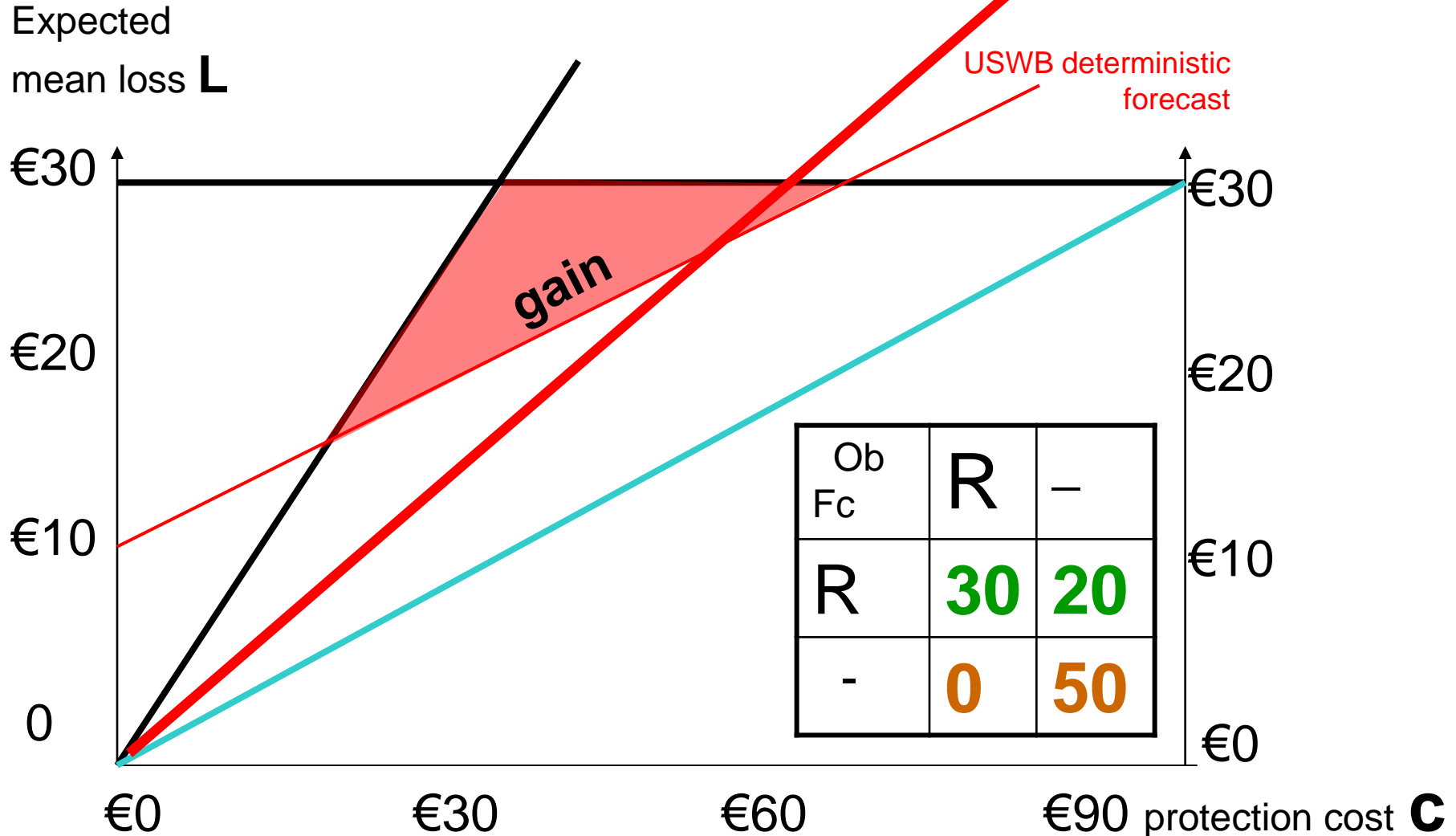
It allows those who are sensitive to rain to interpret the **???** as “it might rain”

USWB	Obs rain	Obs dry
Fc rain	10	0
???	20	20
Fc dry	0	50

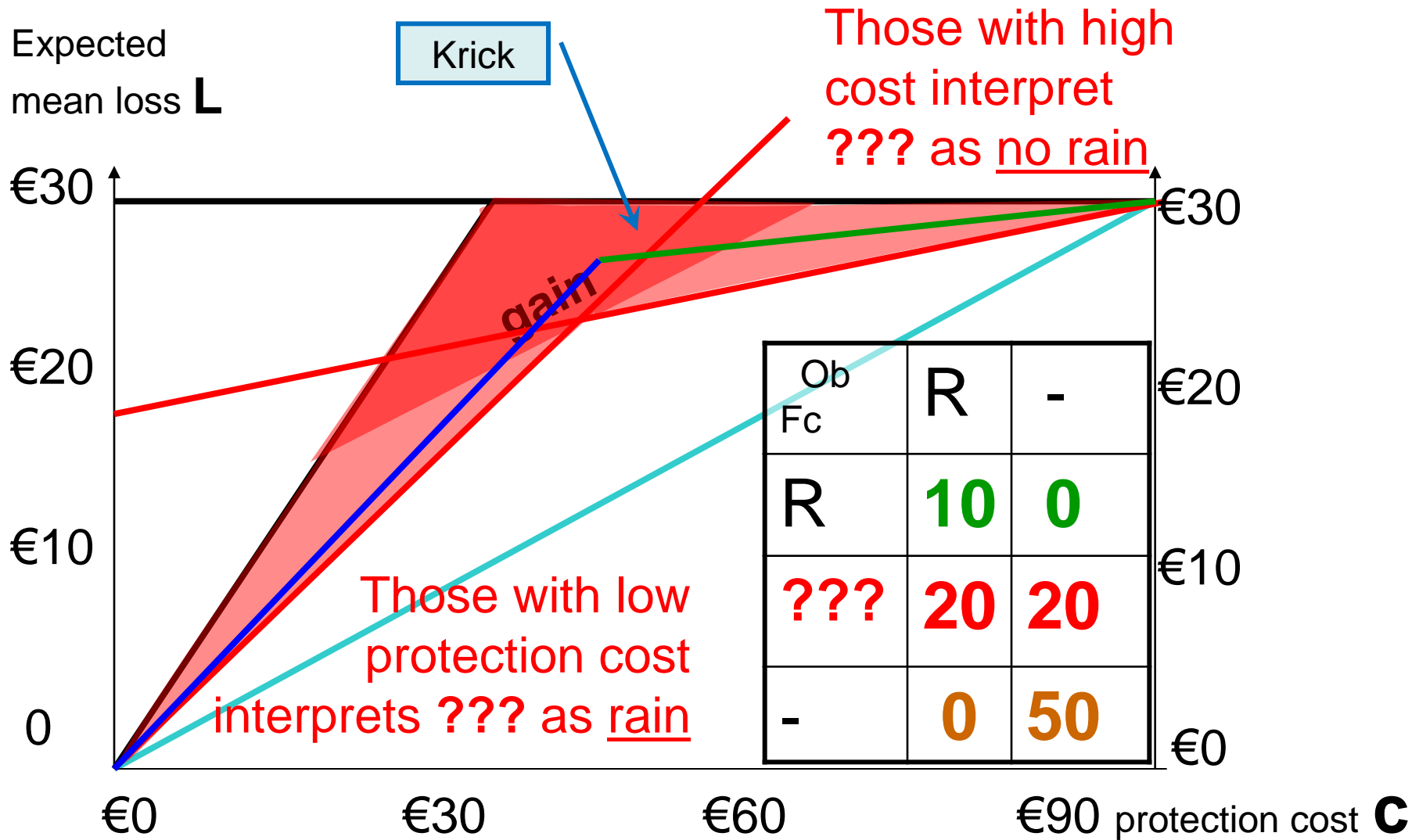


USWB	Obs rain	Obs dry
Fc rain	30	20
Fc dry	0	50

These are the EMV (total cost) for those who interpreted ??? as “it might rain”



And them put them together . . .



This is the “sensitivity to rain” approach:

Categorical

Ob Fc	R	-
R	20	10
-	10	60

Non-categorical

Ob Fc	R	-
R	10	0
??	20	20
-	0	50

This is the matrix for those

Ob Fc	R	-
R	30	20
-	0	50

with low protection cost

This is the matrix for those

Ob Fc	R	-
R	10	0
-	20	70

with high protection cost

...assume rain
to be on the
safe side!

This is the “not sensitivity to rain” approach

Categorical

Ob Fc	R	-
R	20	10
-	10	60

Non-categorical

Ob Fc	R	-
R	10	0
??	20	20
-	0	50

This is the matrix for those

Ob Fc	R	-
R	30	20
-	0	50

with low protection cost

This is the matrix for those

Ob Fc	R	-
R	10	0
-	20	70

with high protection cost

...I can afford to
be hit by the
odd shower!

Paradox 3: The US
Weather Bureau could
have **defeated** Krick by
applying probability
forecasting

Can we quantify the ??? uncertainty?

Categorical

Obs Fc	R	-
R	20	10
-	10	60

Non-categorical

Obs Fc	R	-
R	10	0
???	20	20
-	0	50

Probabilistic

Obs Prob%	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50

What to do with a probability p ?

1. If you do nothing there is a chance p to lose L .
2. On average the loss will be pL (“risk”)
3. If you take protective action it will cost c
4. Only if $p \cdot L > c$ is it worth while to take action
5. The “break even” point is $p = c/L$

Decision matrix for people with c/L almost 100%

Ob Prob	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50

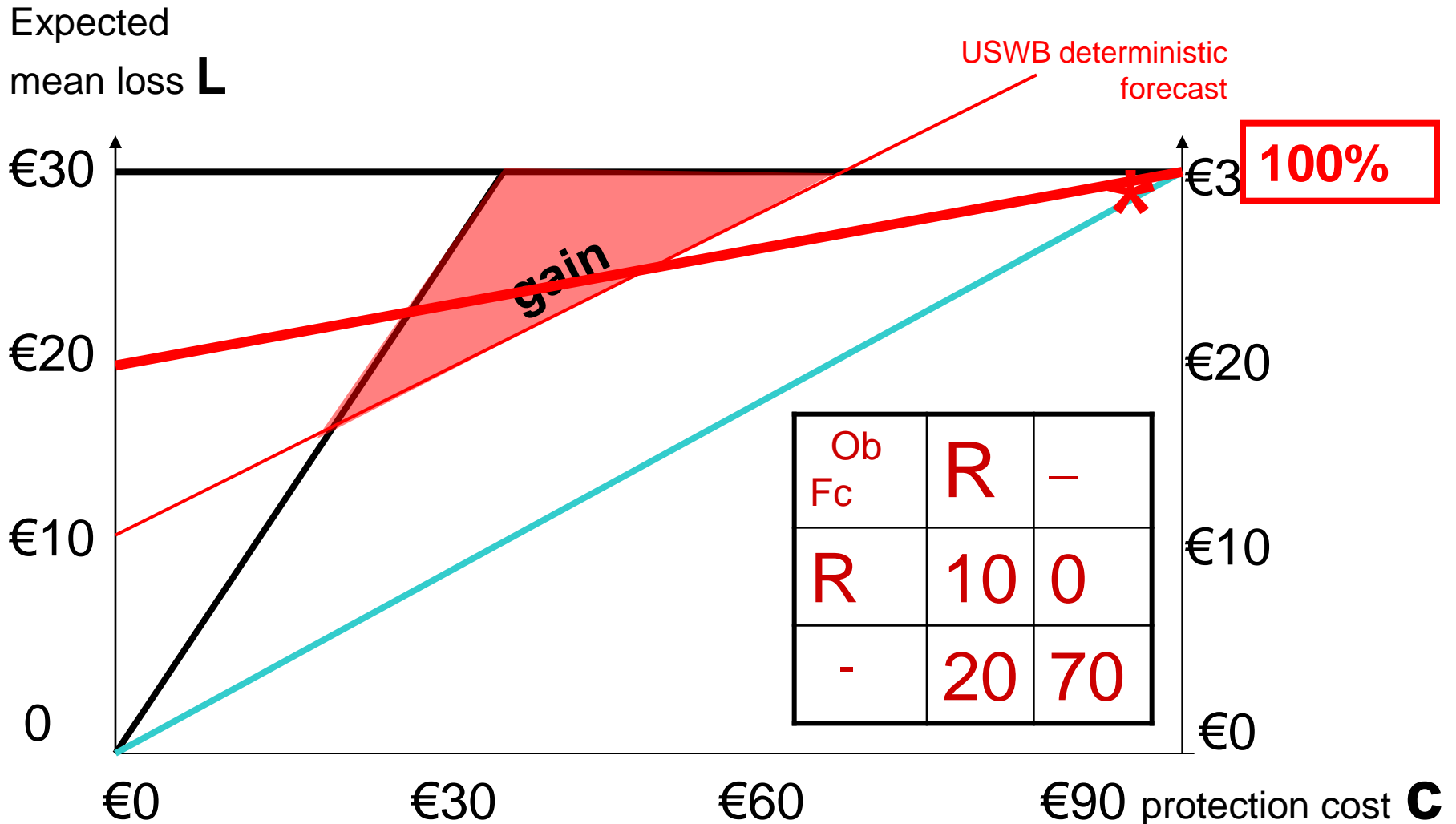


Ob Fc	R	-
R	10	0
-	20	70

Decision matrix

Probability matrix

Gains for people with c/L almost 100%



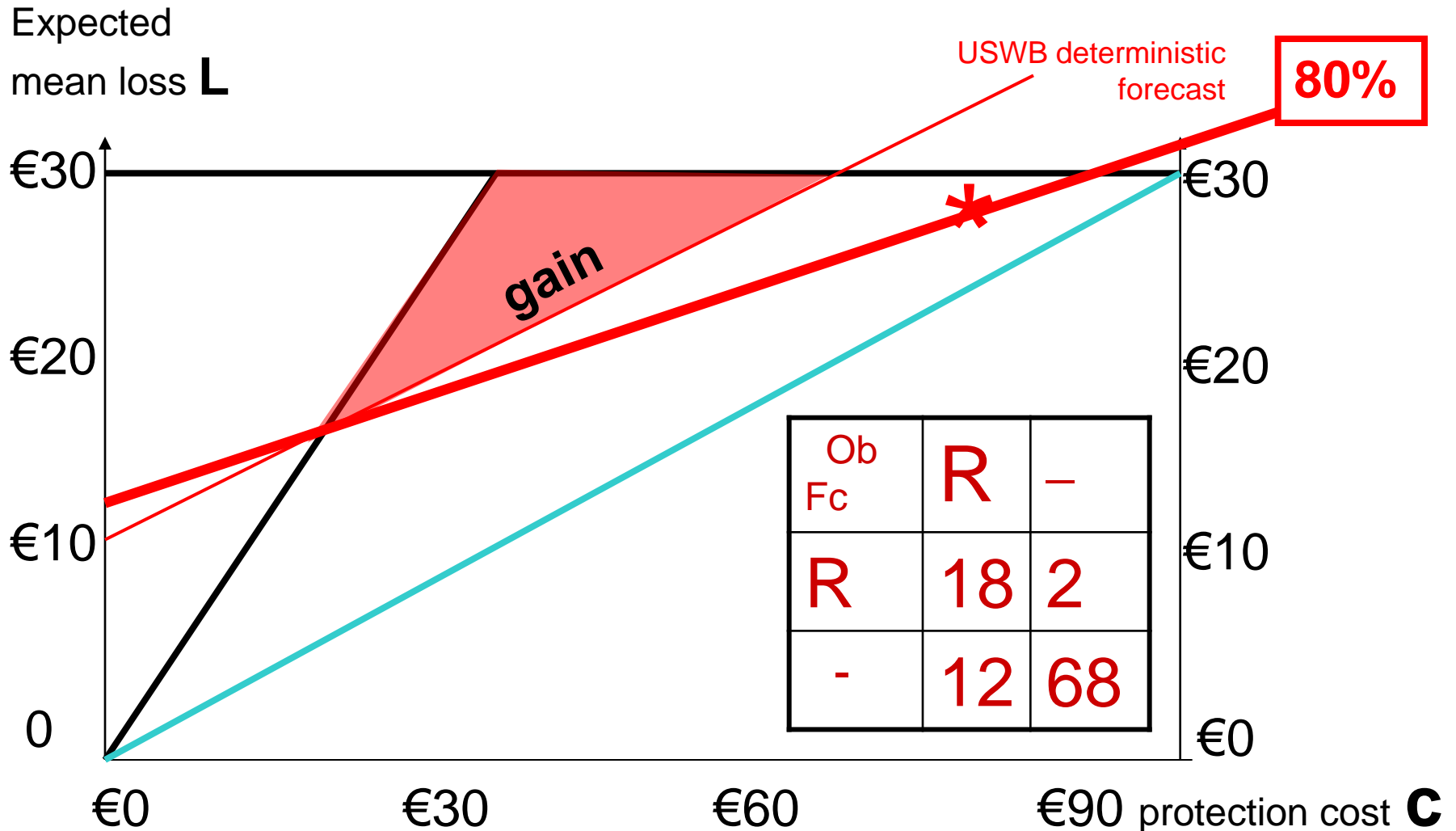
Decision matrix for people with c/L around 80%

Ob Prob	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50



Ob Fc	R	-
R	18	2
-	12	68

Gains for people with c/L around 80%



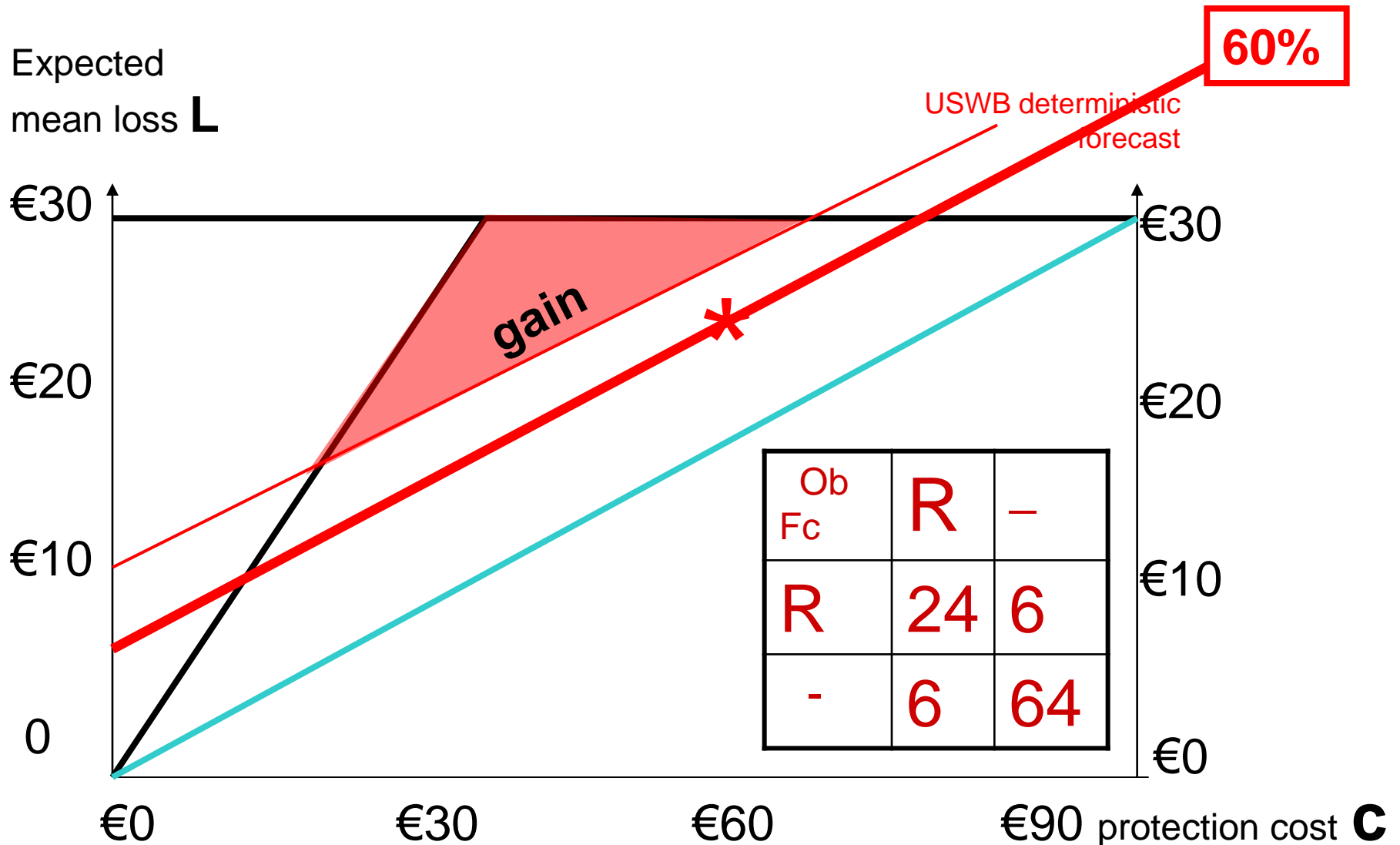
Decision matrix for people with c/L around 60%

Ob Prob	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50



Ob Fc	R	-
R	24	6
-	6	64

Gains for people with c/L around 60%



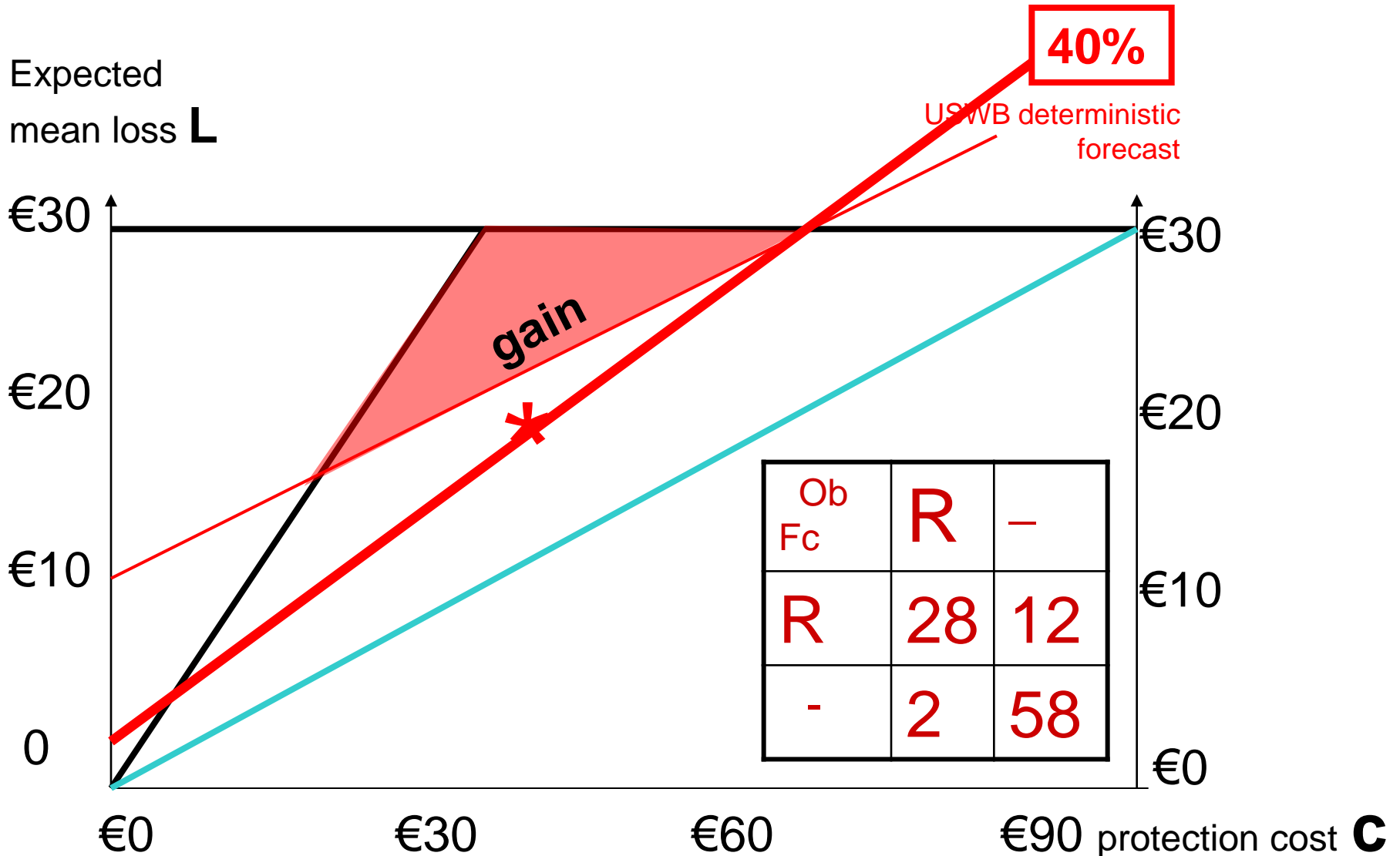
Decision matrix for people with c/L around 40%

Ob Prob	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50



Ob Fc	R	-
R	28	12
-	2	58

Gains for people with c/L around 40%



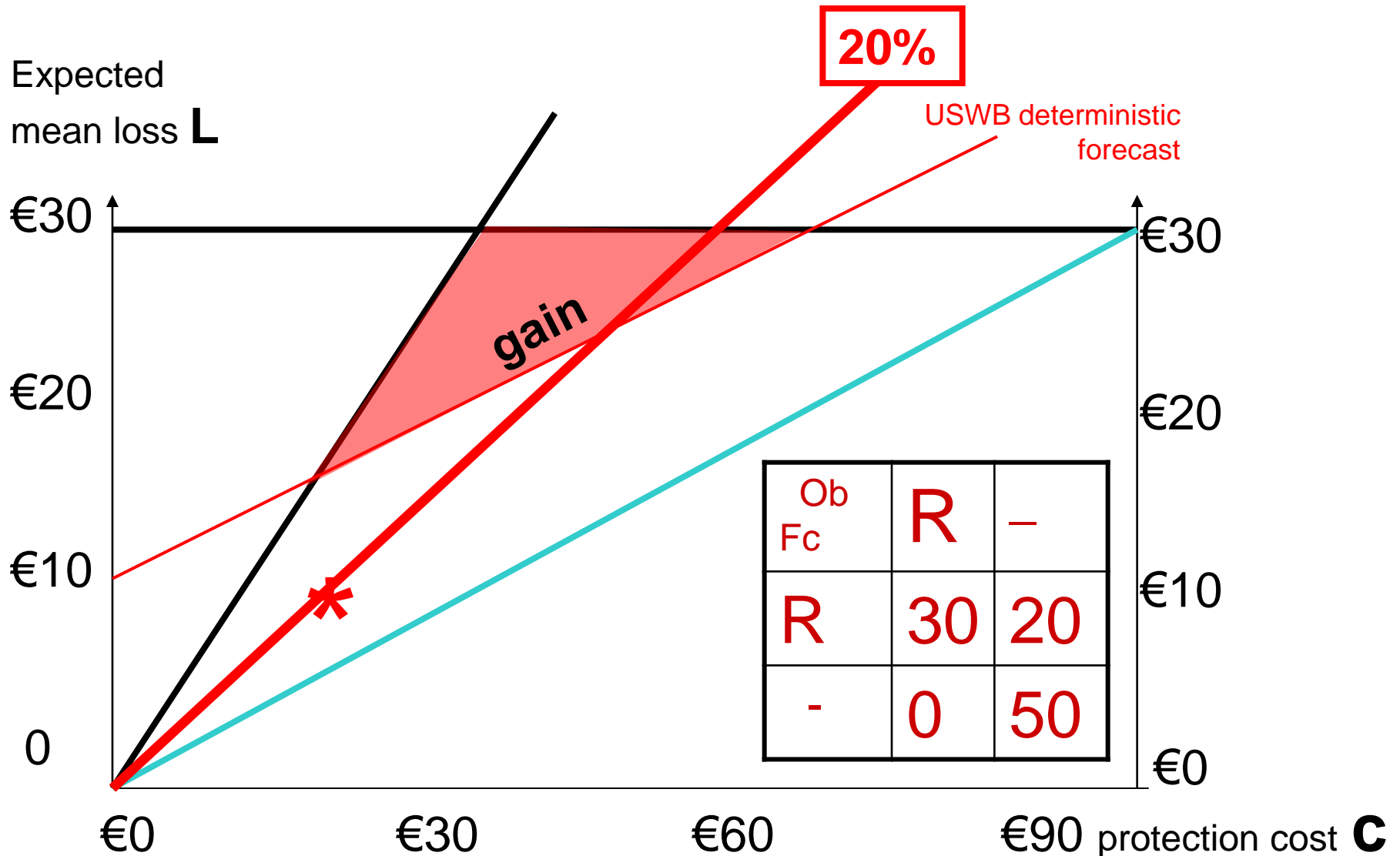
Decision matrix for people with c/L around 20%

Ob Prob	R	-
100	10	0
80	8	2
60	6	4
40	4	6
20	2	8
0	0	50

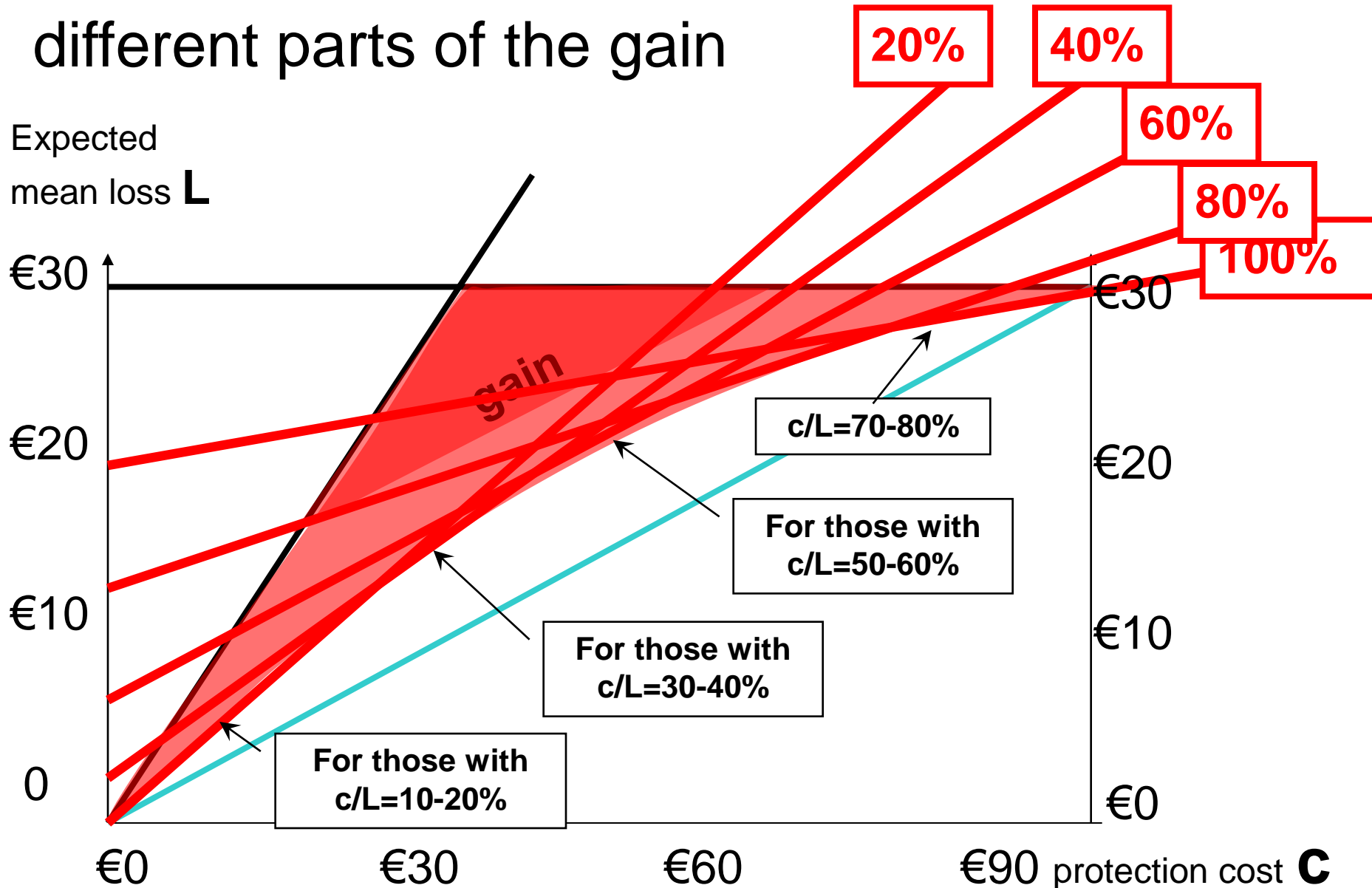


Ob Fc	R	-
R	30	20
-	0	50

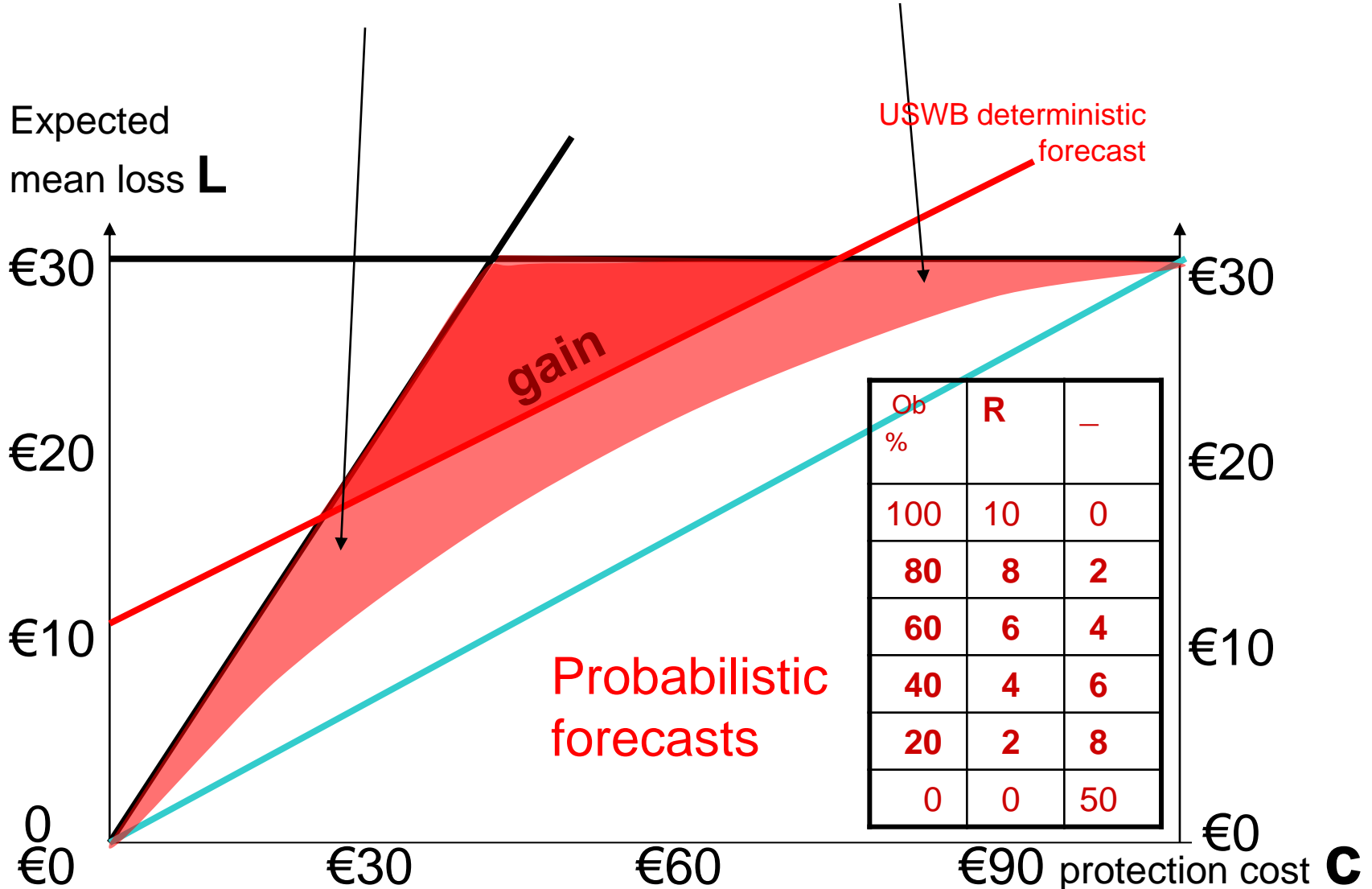
Gains for people with c/L around 20%



Different users benefit from different parts of the gain



Probabilities yield gains for all possible protection costs



END