

# Uncertainty in Cash Flows

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# Why Uncertainty and Sensitivity?

- ▶ We really don't know
- ▶ Some of our guesses may be critical and sensitivity analysis may guide us to spend more time supporting those assumptions.
- ▶ Clients may ask for risk assessments
- ▶ Clients may have other assumptions about critical parameters

# What do you mean “I don’t know”?

- ▶ We are way more certain than we should be.
- ▶ Example:
  - ▶ Make your best guess at the population of Kenya
  - ▶ Give yourself a reasonable upper and lower bound so you are 80% sure the true value is there.
  - ▶ Not zero to a billion

# Answer

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- ▶ One in five of you should be out of bounds.
- ▶ Usually, if honest, more than one in five will be out.
- ▶ If less, Hawthorn Effect

## Some parameters are more critical than others

- ▶ Cost effectiveness depends on estimating costs and benefits.
- ▶ Example: Cost effectiveness of an Energy Efficiency improvement
  - ▶ Costs, commonly install costs are easy to get.
  - ▶ Benefits, due to hours used, are critical.
- ▶ Low use, not cost effective.
- ▶ High use, cost effective.

# Client Requests

- ▶ It is not that they don't trust you.
- ▶ Risk plays into setting a MARR.
  - ▶ High risk, high MARR
  - ▶ Low Risk, low MARR
- ▶ They want to know the probability of success
- ▶ It is part of a portfolio of risks:
  - ▶ They want to know how risky
  - ▶ They want to know how correlated it is with other risks.
- ▶ They like independent risks
- ▶ They like risks that are negatively correlated with existing risk

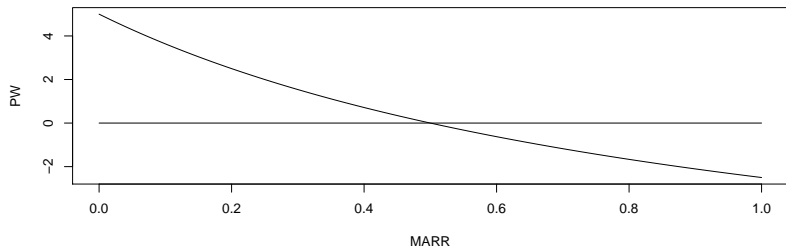
# Clients Have Different Assumptions

- ▶ They may disagree with you but you still want to show them that the results are valid
- ▶ There may be actual diversity.
  - ▶ Different costs at different locations.
  - ▶ Different experiences in different years.
- ▶ You need to do an analysis that covers all that.

## Warnings: You Have Seen Sensitivity Analysis

- ▶ Recall the PW diagrams we used in learning IRR?
- ▶ Shows how PW changes with changes in MARR.

$$PW = -10 + \frac{15}{(1+r)}$$





## Other Parameters

There are some parameters that are usually of concern:

- ▶ Initial costs
- ▶ Salvage value
- ▶ Scale

Some are harder than others:

- ▶ Planning horizon
- ▶ Uncertainty about Risks

## Why planning horizon is harder

- ▶ Mostly you assume time is discrete – not continuous.
- ▶ You need to either change to continuous discounting  $\frac{1}{e^{N(1+r)}}$  rather than  $\frac{1}{(1+r)^N}$
- ▶ Or create an equivalent MARR with smaller compounding periods but equivalent with their nominal statement.

## Example: MARR 10% per year compounded daily

Start with 10% per year MARR and find equivalent annual MARR compounded daily.

$$\left(1 + \frac{r}{365}\right)^{365} = 1.1$$

$$365 \left[1.1^{\frac{1}{365}} - 1\right] = 0.0953226$$

Allows you to vary time on a smaller time frame.

# Why uncertainty about risks is harder

- ▶ Risk is when you know the odds but don't know the outcome.
  - ▶ You know that each side of a true die has a  $1/6$  chance of showing.
- ▶ Uncertainty is when you don't know the odds.
  - ▶ Uncertain if it is your trick die that rolls a 6 half the time or the true die.
- ▶ You end up describing your beliefs about the probability as a Beta distribution.
  - ▶ Hint, you just walked into the land of Bayesian statistics.
  - ▶ We will do some Monte Carlo simulations later so you can see what to do.
- ▶ Talk about your sensitivity analysis as about the mean of the present worth rather than the present worth.

## Simple How To: PW Example

Consider a simple constant series that we can describe the present worth with factor notation:

$$P = -C + A(P|A, i = r, N = n)$$

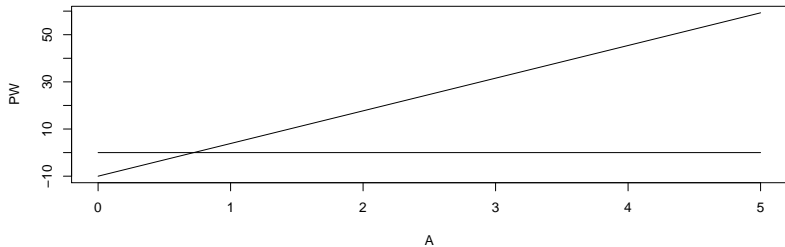
Lots of parameters to vary:

- ▶  $A$ , the size of the constant benefits.
- ▶  $r$ , the discount rate
- ▶  $n$ , how long you get payments
- ▶  $C$ , the initial cost

We warned you about  $r$  and  $n$  so let's vary just  $A$

## PW Figure

$r = .1, n = 10, C = -10, A \in [0, 5]$



BTW PW is zero when  $A = 0.7217385$

# What Do You Learn from this?

- ▶ If you are pretty sure that  $A$  is not near 0.7217385, you are done spending time getting a better estimate of  $A$ .
  - ▶ Refining your estimate of the parameter with more data, more detailed calculations, more bids, more literature review will not change the decision.
  - ▶ Looking for more precision or less uncertainty is a waste of resource.
- ▶ If not, then you need to decide if the effort is worth the refinement. That depends on your other options.

## Cost benefit analysis of doing cost benefit analysis



Figure 1



## What Do You Learn from this? (Con't)

- ▶ The client can check on the consequences of their beliefs about A.
- ▶ Note we still don't know risk, for that we need to know the distribution of our beliefs about A.
  - ▶ Is it  $N(2, .5)$ ?
  - ▶ Is it  $U(1,4)$ ?
  - ▶ We will do this later.
- ▶ Sensitivity with present worth is easiest of the three.

## AW Example

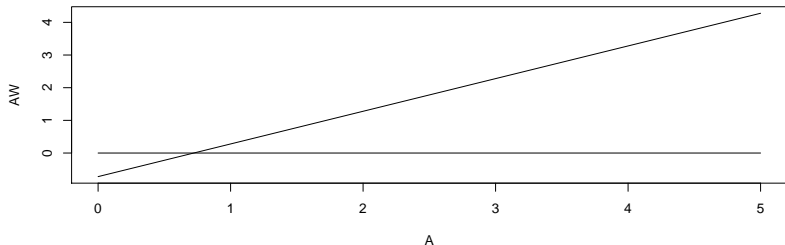
- ▶ Stick with the same cash flow as before with present worth factor notation:

$$P = -C + A(P|A, i = r, N = n)$$

- ▶ No more difficult than PW sensitivity analysis unless you want to vary the  $N$ .
- ▶ Annual worth is just

$$AW = A - C(A|P, i = r, N = n)$$

## AW Figure



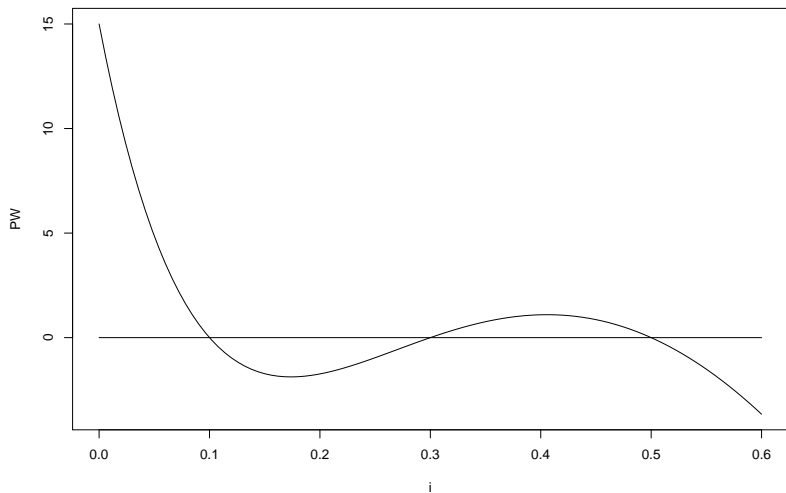
- ▶ This is the same information as the PW graph translated into AW.
- ▶ Some clients like to see it this way.

## IRR Example

- ▶ Will use a more complex model for this because IRR can get odd.
- ▶ Changing the parameters can change not just the location of the roots, i.e., the IRRS but also the number of roots.
- ▶ If you want to learn more, read up on bifurcations.
- ▶ Reuse the example for multiple IRRs we had in class:
  - ▶ -1000, 3900, -5030, 2145

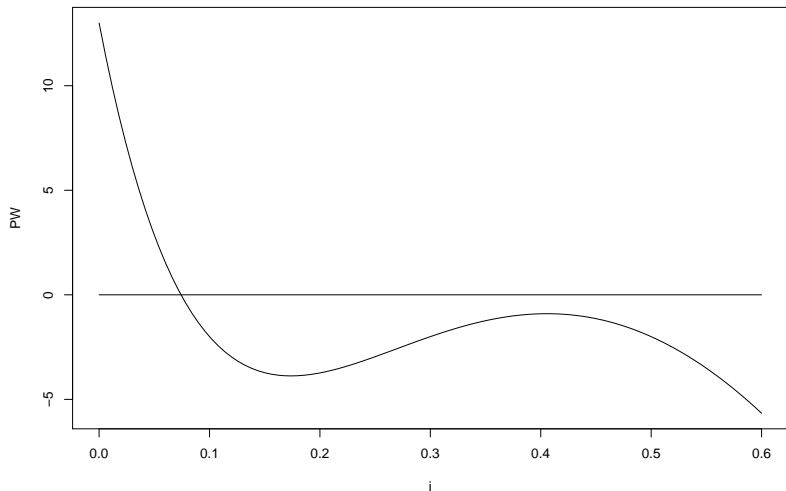
## Recall the PW function

This has IRRs at 10%, 30% and 50%



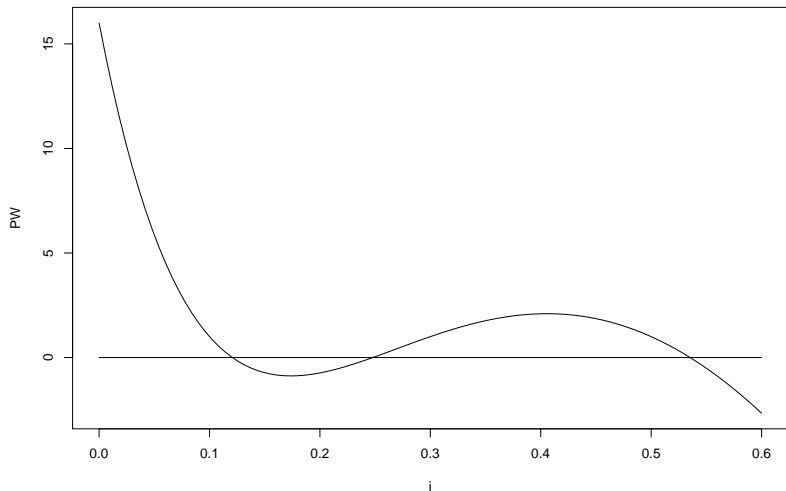
## Watch the roots change

$A_0 = -1002$  One root



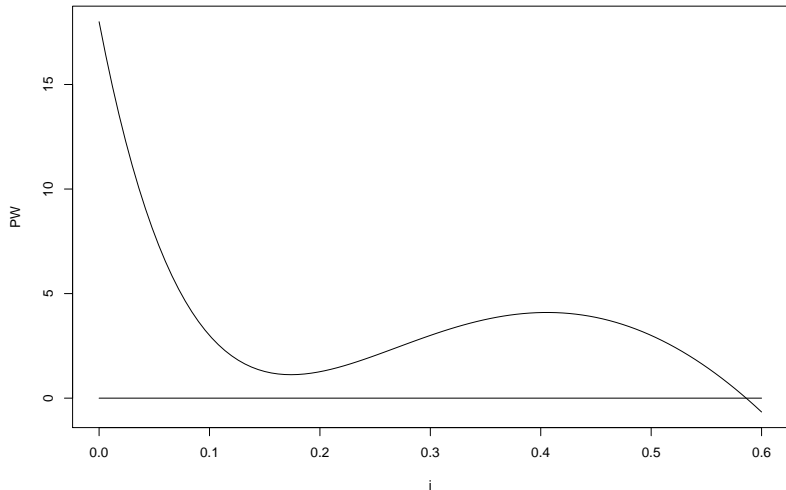
## Watch the roots change

$A_0 = -999$  Three roots



## Watch the roots change

$A_0 = -997$  One root again





# IRR Problems

- ▶ Your IRR sensitivity functions are not always continuous.
- ▶ It requires a lot of time with clients to explain this.
- ▶ Recommend talking your clients out of it.

## Priors for Sensitivity Analysis: Finding a Good Range

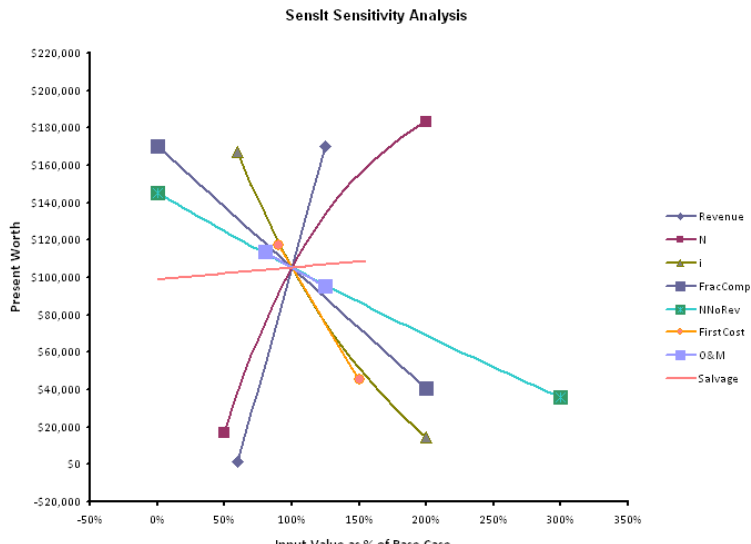
- ▶ In the examples above we just picked bounds.
- ▶ Book gives some bad advice on how to pick the design points, bounds, in a sensitivity analysis,  $\pm 20\%$ .
  - ▶ Sometimes 20% is way too big, remember the initial cost in the IRR example? \$4 on a scale of \$1000
  - ▶ Sometimes 20% is way too small. What is the smallest safety factor you have ever used? 1.25?

# Single Person

- ▶ A few warnings.
  - ▶ This does not work with everyone. They have to be reasonably numeric.
  - ▶ They have to be a real expert, with real opinions, real experience with data.
  - ▶ It is ok, if they need time, have to make some calculations or want to look a few things up.
- ▶ Three questions for each parameter
  - ▶ What is your best guess?
  - ▶ Largest value that you are 90% sure the value is below?
  - ▶ Smallest value that you are 90% sure the value is above?
  - ▶ Basically, what are your 80% bounds.

## Allows a Spider graph.

- ▶ Horizontal axis shows percent difference from best guess.
- ▶ Vertical shows the investment criteria of interest.



# What to watch for?

- ▶ Parameter that cause a decision, sign, change.
  - ▶ Can turn a good idea into a bad one without more research
- ▶ High sensitivity parameters, high slope.
  - ▶ Your expert may be more certain than they should be and should be consulted more about that parameter.
- ▶ Low sensitivity parameters, low slope.
  - ▶ No more time should be spent on these.
  - ▶ Unless, your job is devil's advocate.

# Multi-Person: Feasible

You have many choices:

- ▶ Ask everyone for their best guess for each parameter:
  - ▶ Leads to Monte Carlo Simulations.
  - ▶ Captures correlations between beliefs in parameters
- ▶ Delphi Survey:
  - ▶ Asks not just for numbers, but why you choose the number.
  - ▶ Monte Carlo Simulations

## Possible but difficult

- ▶ Ask everyone for the best guess and 80% bounds:
  - ▶ Beyond scope of course.
  - ▶ Use with small numbers of experts
  - ▶ Welcome again to Bayesian land.
- ▶ Ask everyone for joint distributions:
  - ▶ Hard to to do.
  - ▶ Goes Bayesian again.
  - ▶ Never had a client that could do it.
  - ▶ I have trouble doing it.

## Example Joint Distributions

Ask Everyone for best guess at each parameter.

- ▶ You need at least as many experts as parameters.
- ▶ If you have a lot of experts, 30+, you can simulate the objective function directly on the data or bootstrap for the uncertainty.
- ▶ If you have few, you can approximate with a multivariate normal distribution.



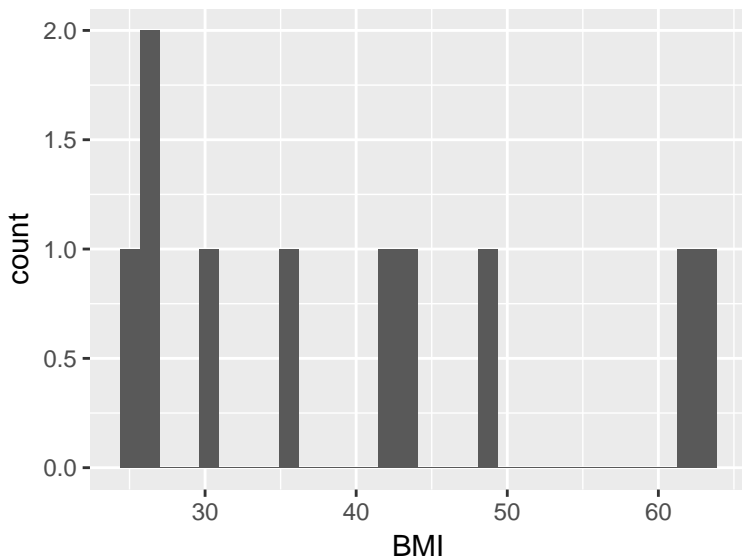
## What is Jamie's BMI?

- ▶ Figure out Jamie's Body Mass Index,  $\frac{kg}{M^2}$ .
- ▶ Collected these 10 guesses of height and weight in M and kg.

Ht	Wt
1.821870	83.46805
1.386833	83.70864
1.827694	88.13024
1.680432	87.01051
1.419243	84.47353
1.178424	85.92407
1.815408	86.93528
1.570278	87.64317
1.345283	87.76748
1.152229	84.07292

## Direct

- ▶ Experts say the median BMI is 38.7409177.
- ▶ 80% are sure it is between 26.2551993 and 62.0195641



## Or Estimate the Joint Normal Distribution

- ▶ Means of 1.5197694, 85.9133883
- ▶ and var/cov of

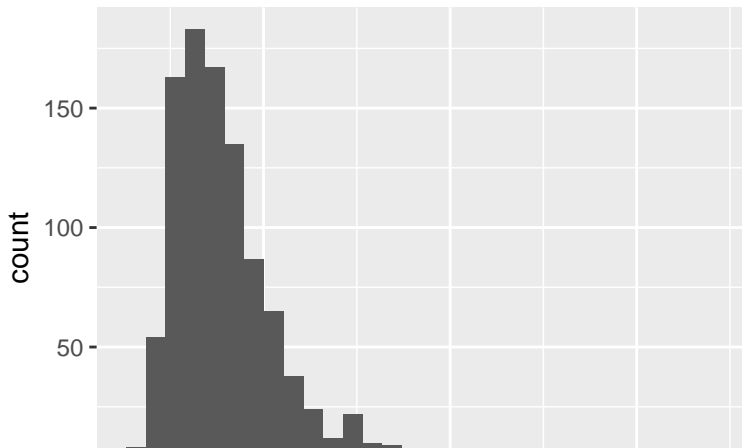
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0.0679637	0.1440837
0.1440837	3.3214899

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## And Simulate for 1000 replications

- ▶ You can get exact results for Normals added to Normals but the minute you divide, stick to reporting confidence intervals, medians and modes.
- ▶ Normal divided by Normal is Cauchy, which has no defined mean and no defined variance.



# Problems with “Just ask experts”

- ▶ Responses are not always independent
  - ▶ The boss is in the room and you don't want to disagree.
  - ▶ Everyone uses the same reference document
  - ▶ Example, you can tell which field by the number people give for the value of a statistical life.
    - ▶ \$9.1 million, Environmental Protection Agency
    - ▶ \$7.9 million, Food and Drug Administration
    - ▶ \$9.4 million, Transportation Department
- ▶ Little thought is put into it

# Fixes

- ▶ Use incentivized polls like Intel.
- ▶ Parimutuel betting, like some electrical utilities.
- ▶ Delphi Surveys

# Incentivized Polls

- ▶ Guess Jamie's BMI. Closest to the actual gets \$10.
- ▶ Skin in the game gives incentive for more thought.
- ▶ Only works when you will eventually find out actual.
- ▶ Oddly works better if you don't know how people are betting.
  - ▶ Can produce group-think bubbles, if you see how others are betting.
  - ▶ Easy to find the mode, but harder to tell if it is real.
- ▶ Can get expensive with many winners:
  - ▶ everyone guesses the same correct answer.
  - ▶ More expensive when there is less certainty.

# Parimutuel Betting

- ▶ Guess Jamie's BMI. Correct answers split the pot with the others that got it right.
- ▶ Racetrack method.
- ▶ Less likely to get bubbles, but
  - ▶ Harder to see if there is a mode in the guess.
- ▶ Cheaper than incentivized poll if high probability of multiple winners.
- ▶ Cost does not increase with a priori certainty.



# Delphi Surveys

## General Pattern:

- ▶ Ask each person in private for best guess.
- ▶ Compile results.
- ▶ Ask outliers why they said what they did.
- ▶ Give everyone:
  - ▶ The distribution of guesses for each parameter
  - ▶ The reasons the outliers gave for the answer they gave.
- ▶ Ask for another guess.
- ▶ Report the new distribution or repeat if desired.

Interested in More?