

### **Determining the Value of a Prototype**

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### What are prototypes worth?





### Setting the Stage – A program is proposed

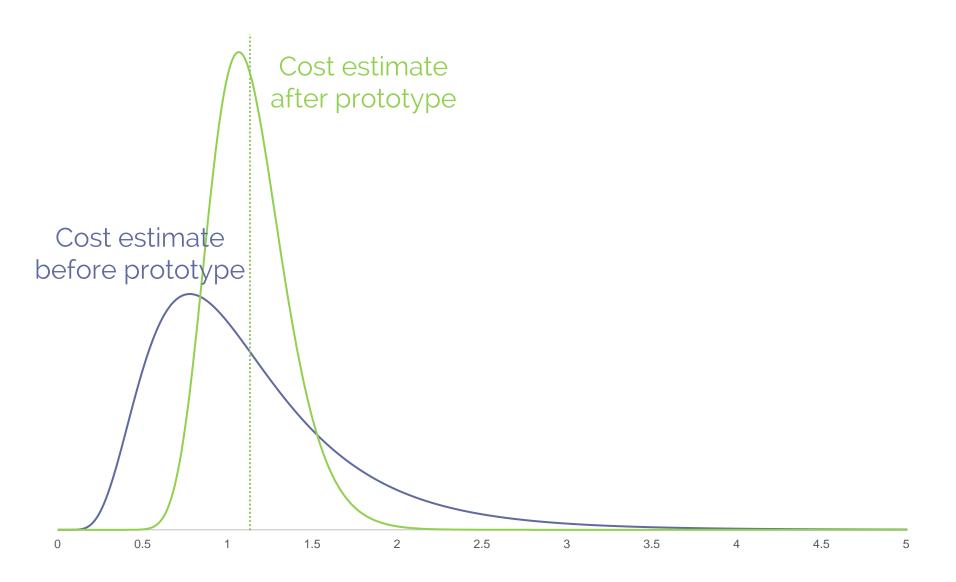
A decisionmaker has three options:

- 1. Start the program
- 2. Do not start the program
- 3. Make a prototype, and then start or do not start the program

An alternative program option exists with a known cost Cost is the discriminator

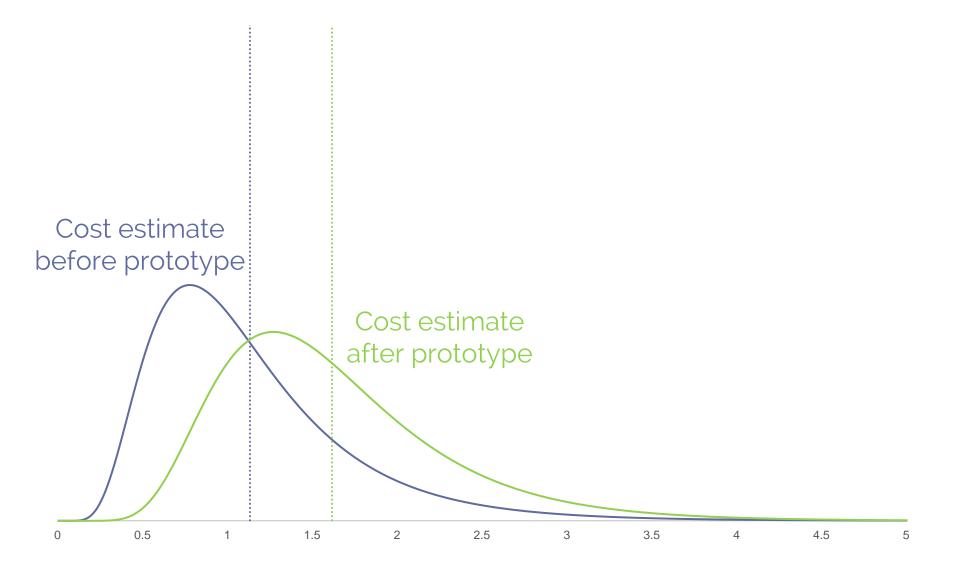


### **Prototypes give us information**



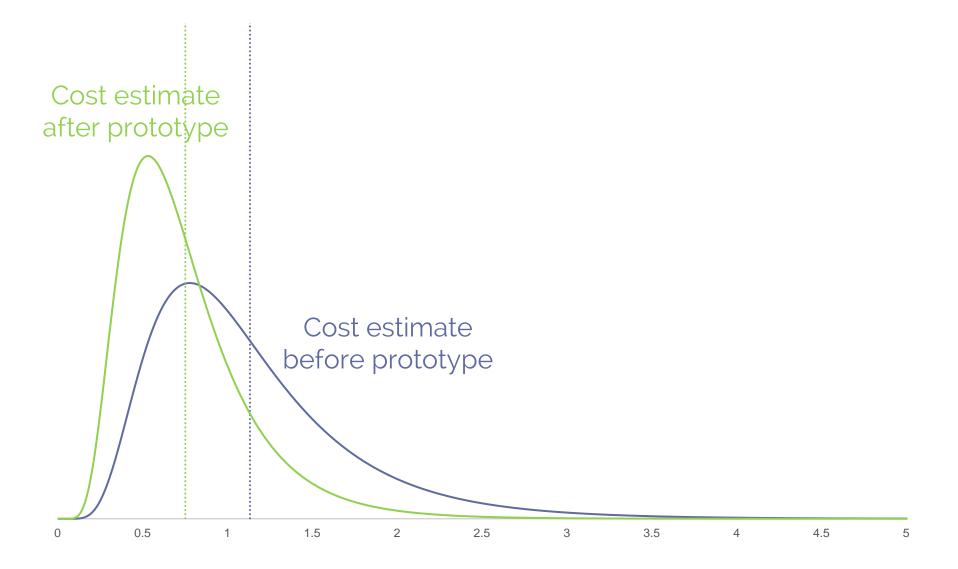


### **Prototypes give us information**



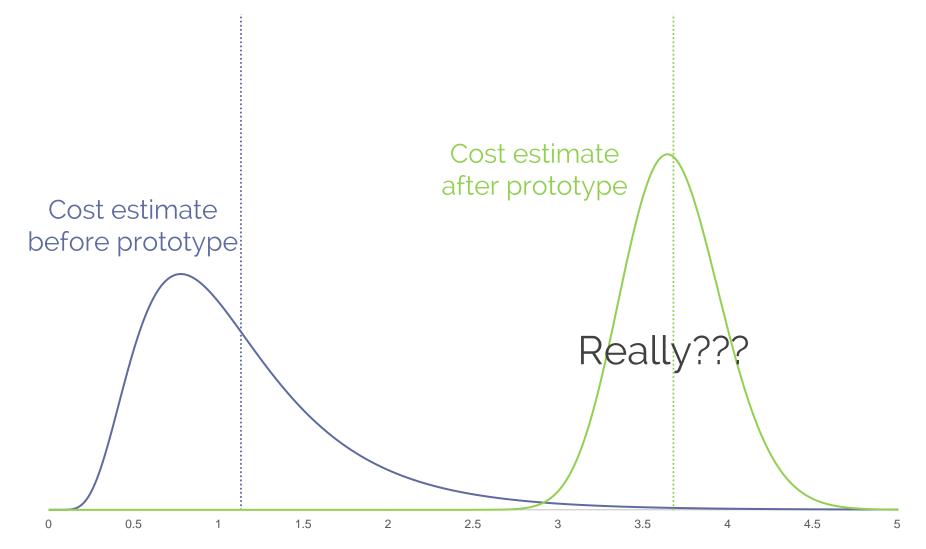


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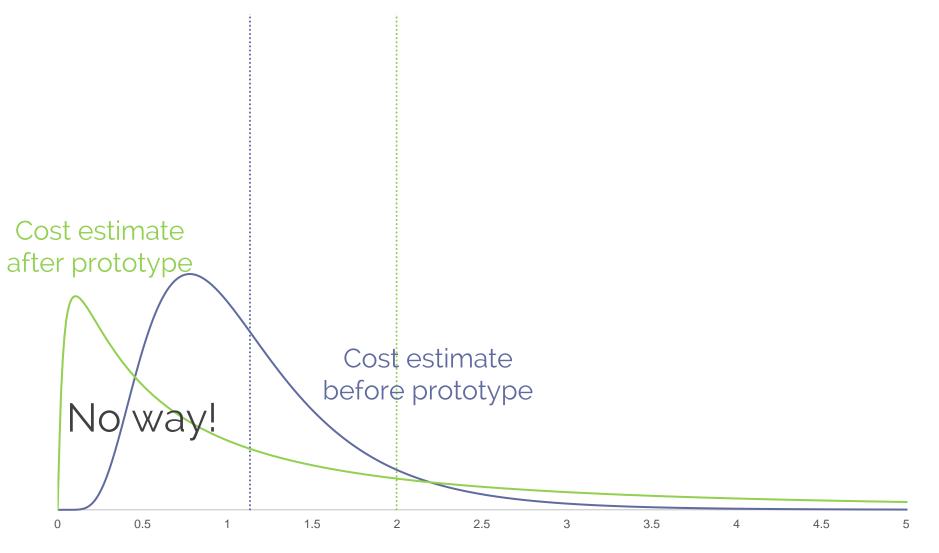


## Prototypes give us information...but we know a little of what to expect





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#### **The Preposterior Distribution**

## The distribution of the means of all possible cost distributions after a prototype outcome



### Properties of the Preposterior Distribution

The mean (expected value):

E[preposterior] = E[prior]

## The mean of the preposterous is equal to the mean of the prior



Properties of the Preposterior Distribution The mean (expected value): E[preposterior] = E[prior]

The variance:

 $Var[preposterior] = Var[prior] - E_{d \in posteriors}[Var[d]]$ 

The variance of the preposterous is equal to the variance of the prior less the mean of the posterior variances



Properties of the Preposterior Distribution The mean (expected value): E[preposterior] = E[prior]

The variance:  $Var[preposterior] = Var[prior] - E_{d \in posteriors}[Var[d]]$ 

The shape:

 $Dist[preposterior] \rightarrow Dist[prior]$  as  $experiments \rightarrow \infty$ 

The shape of the preposterous approaches that of the prior with increased experimental data



Properties of the Preposterior Distribution The mean (expected value): E[preposterior] = E[prior]

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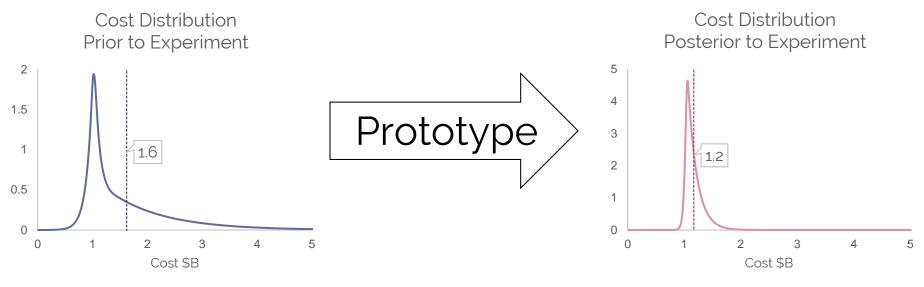
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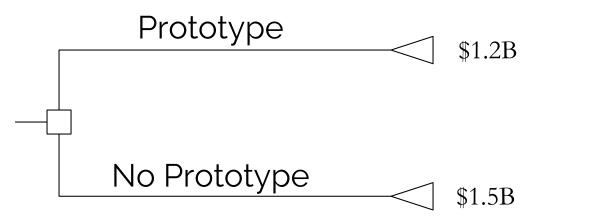
In other words, the preposterior distribution looks a lot like the prior distribution



### How to use the preposterior distribution



Suppose the alternative costs \$1.5B

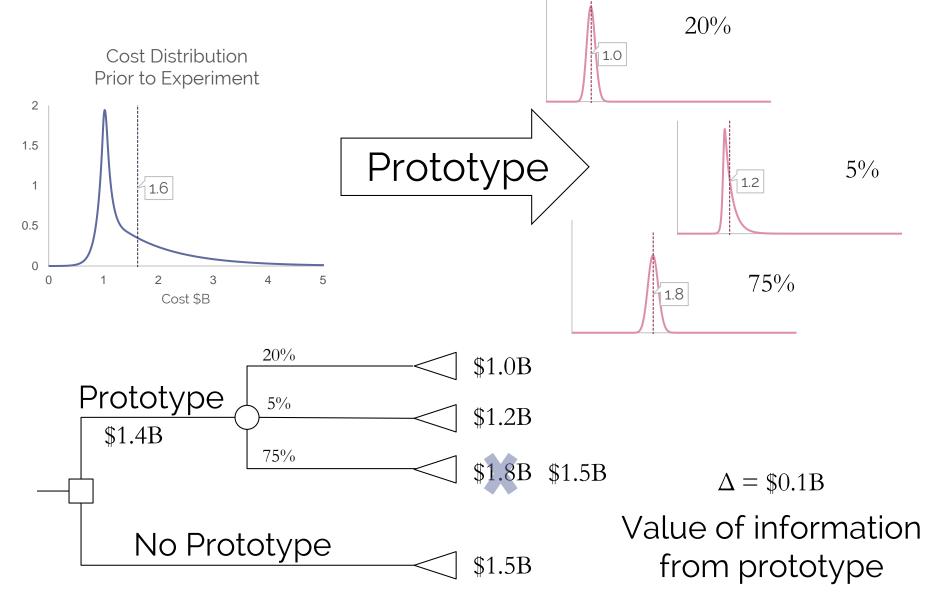


 $\Delta = \$0.3B$ 

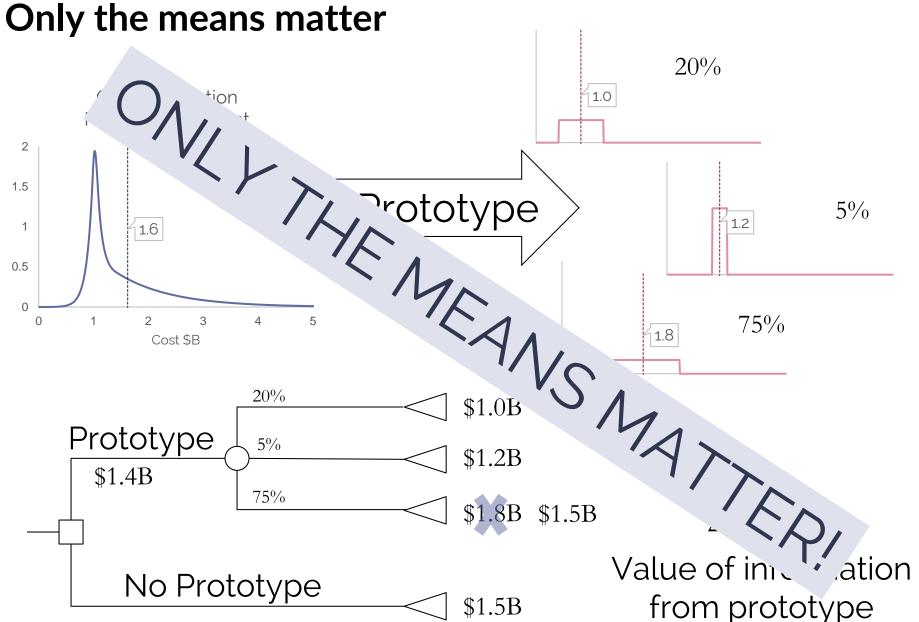
Value of information from prototype



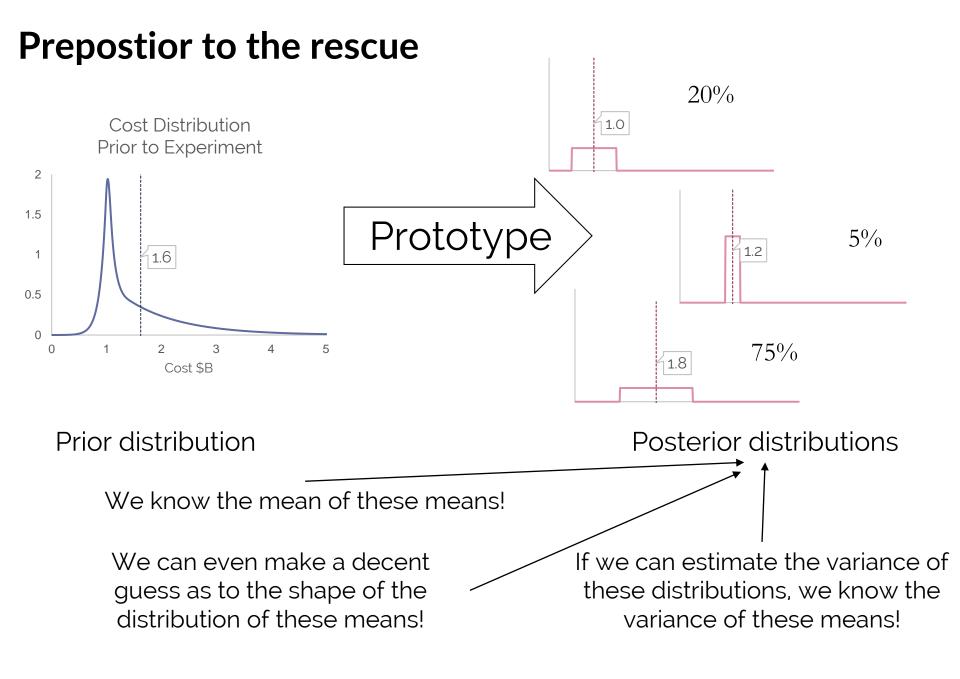
### How to use the preposterior distribution













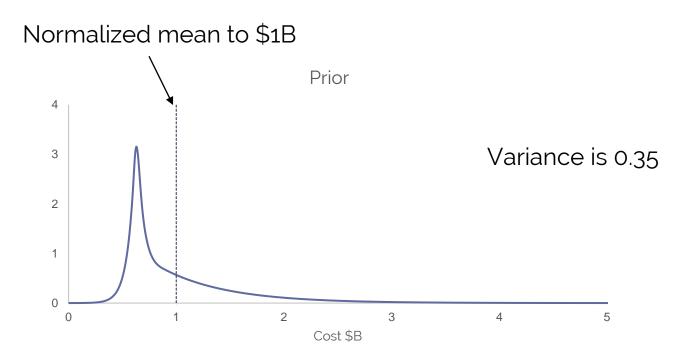
### Steps to estimate Value of Information for prototype

- 1. Estimate distribution of costs (prior)
- 2. Estimate reduction in variance of cost due to prototype (posteriors)
- Model preposterior distribution as prior distribution with mean held constant (from 1) and variance equal to reduction in variance (from 2)
- 4. Perform decision tree on preposterior distribution with some given alternative cost



From Prototyping Defense Systems

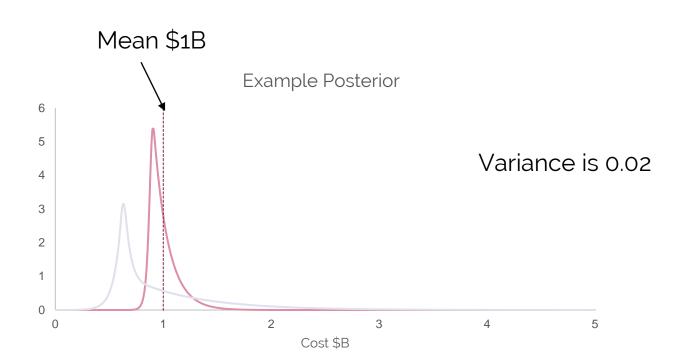
## Derived prior distribution from percent cost overruns of systems without prototypes





From Prototyping Defense Systems

# Derived a posterior variance from percent cost overruns of systems with prototypes





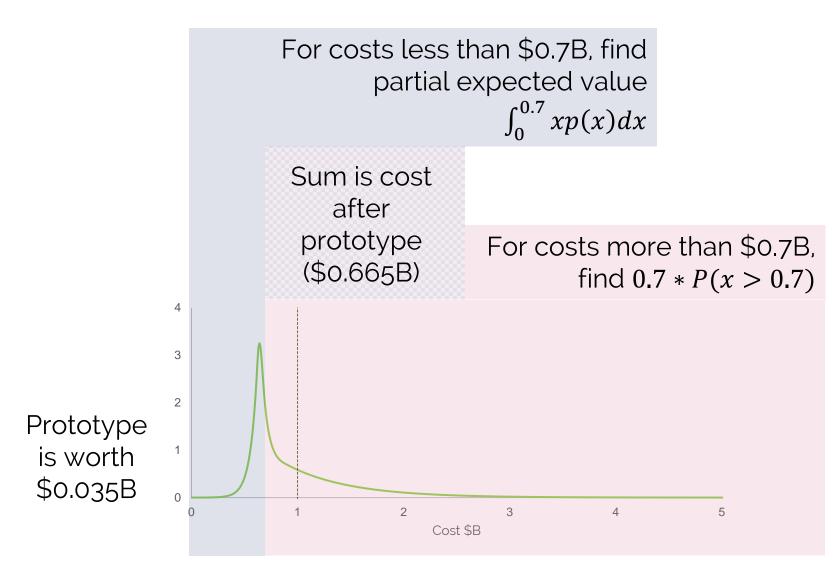
From Prototyping Defense Systems

### Derived a preposterior distribution as prior with variance decreased by 0.02



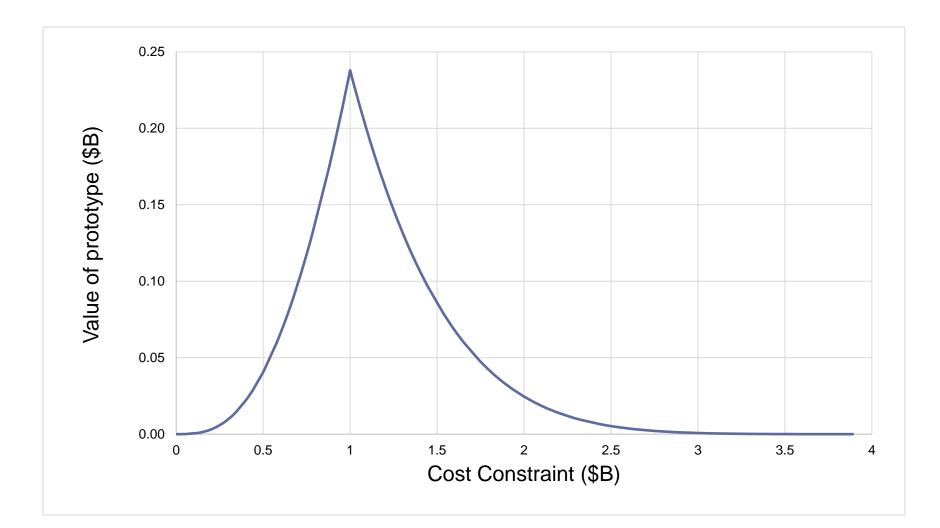


### Suppose alternate cost is \$0.7B



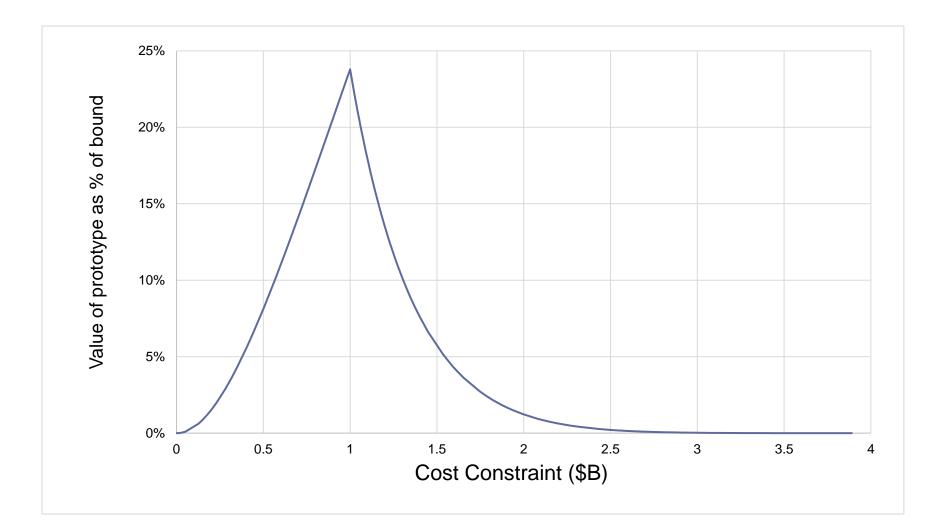


### Letting cost of alternative vary





### Letting cost of alternative vary





### Thanks to the Air Force Research Laboratory for their inspiration and help in this project. And thanks to you!



### Backup



#### Bayes tells us something about what we can expect

P(A|B)

$$\frac{P(B|A)P(A)}{P(B)}$$

We want the distribution of

the cost *given* some prototype outcome

We need the distribution of

- the prototype *given* cost ??????
- the cost
- the prototypes Just a scaler!

So, not great. But what if we look at attributes that incorporate all possible prototype outcomes?

