

Bayesian Dirichlet Cost Rules of Thumb

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Cost Rules of Thumb (ROT)

- Cost rules of thumb are critical components of early project formulation, specifically for low CML (Concept Maturity Level)
- Rough costs of spacecraft subsystems must be generated before more detailed estimates can be made
- It is generally thought that there is a "typical" percent allocation of total mission cost to the Level 2 WBS (Work Breakdown Structure) elements

WBS 01	Project Management	WBS 06	Flight System / Spacecraft
WBS 02	Systems Engineering	WBS 07	Mission Operations System (MOS)
WBS 03	Safety and Mission Assurance	WBS 08	Launch Vehicle / Services
WBS 04	Science / Technology	WBS 09	Ground Data System (GDS)
WBS 05	Payload(s)	WBS 10	System Integration, Assembly, Test & Check

Mission A





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Traditional ROT Approach

Mission	Category	Fixed FY WBS 01 Cost %	Fixed FY WBS 02 Cost %	Fixed FY WBS 03 Cost %	Fixed FY WBS 04 Cost %	Fixed FY WBS 05 Cost %	Fixed FY WBS 06 Cost %	Fixed FY WBS 07 Cost %	Fixed FY WBS 09 Cost %	Fixed FY WBS 10 Cost %
x	MIDEX	4.5%	3.6%	3.4%	18.1 <mark>%</mark>	25.6 <mark>%</mark>	39.5%	4.0%	1.3%	
x	MIDEX	8.5%	1.0%	1.5%	1.7%	45.3%	34.1%	3.9%	3.9%	
x	MIDEX	5.1%	3.2%	1.7%	8.3%	34.4%	30.9%	1.6%	12.0%	2.7%
x	MIDEX	3.6%	2.1%	2.2%	4.7%	30.9%	39.8%	10.9'%	2.1%	3.7%
x	MIDEX	6.5%	1.8%	7.9%		40.7%	35.4%		4.1%	3.5%
x	MIDEX	6.7%	4.1%	2.0%	4.4%	43.1%	34.5%		2.6%	2.6%
x	MIDEX	1.7%				45.0%	33.5%	8.5%	11.3%	
x	MIDEX	6.0%	10.0%		2.2%	18.7%	53.1%	4.9%	5.1%	
Average Percent (ignoring missing data):		5.3%	3 .7%	3 .1%	6.6%	35.5%	37.6%	5.6%	5 .3%	3.1%
Rescaled Rule of Thumb:		5.0%	3.5%	3.0%	6.2%	33.5%	35.5%	5.3%	5.0%	2.9%

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Dirichlet ROT Approach

- The <u>Dirichlet distribution</u> describes the uncertainty of multivariate distributions
- Can be understood as the multivariate extension of the Beta distribution
 - Binomial = model for # successes of n trials with probability p
 - Beta = model for uncertainty around p
 - Multinomial = multivariate extension to binomial distribution, defined by vector of probabilities which sum to 1
- The Dirichlet distribution is the simplest distribution on the <u>simplex</u> (vectors whose components sum to one)



Figure: Full uncertainty and correlation between WBS elements of posterior predictions

Aleatoric and Epistemic Uncertainty

- <u>Aleatoric Uncertainty</u>: uncertainty which arises from inherent randomness
 - e.g. normal distribution standard deviation
- <u>Epistemic Uncertainty</u>: uncertainty in the model form or in the uncertainty in the parameters of the model form(s)
 - e.g. uncertainty in the value of the std. deviation of the normal distribution
 - e.g. the uncertainty could be described by a normal distribution <u>or</u> a t distribution
 - For our model we only consider uncertainty in the parameters
 - The Dirichlet distribution is parameterized by a mean vector \vec{a} and a onedimensional precision parameter ϕ



Figure: Schematic of aleatoric and epistemic uncertainty for a Dirichlet model on Δ^2 (the simplex on \mathbb{R}^3 ; the points [x, y, z] such that x + y + z = 1 and $x, y, z \ge 0$)

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Mission Classes and Model Form

- Each mission class has a different set of cost categories, i.e. different WBS and mission phases included based on the AO
- We split the data based on mission AO and form separate models for each

Mission Class	Mission Cost Range	Number of Records	Cost Categories
Flagship	> \$2B	3	Phase B-D WBS 1-10 except 8
New Frontiers	\$1B	4	Phase A-D WBS 1-10 except 8
Discovery	\$500M	13	Phase B-D WBS 1-10 except 8, and Phase E-F WBS 1-4, 7, and 9
MIDEX	\$300M	8	Phase A-F WBS 1-10 except 8

For Flagship, New Frontiers, and MIDEX: $\vec{c} \sim \text{Dirichlet}(\vec{\alpha}, \phi)$

 \vec{c} is a nine-dimensional vector on the simplex representing WBS 1, 2, 3, 4, 5, 6, 7, 9, and 10 for the Phases in the table

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For Discovery:
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$$\vec{c}_{\text{phase}} \sim \text{Dirichlet}(\vec{\alpha}_{\text{phase}}, \phi_{\text{phase}})$$

$$\vec{c}_{BCD} \sim \text{Dirichlet}(\vec{\alpha}_{BCD}, \phi_{BCD})$$

$$\vec{c}_{EF} \sim \text{Dirichlet}(\vec{\alpha}_{EF}, \phi_{EF})$$

$$\vec{c} = \begin{bmatrix} c_{\text{phase},1} \vec{c}_{BCD} & c_{\text{phase},2} \vec{c}_{EF} \end{bmatrix}$$

** We also have priors on the Dirichlet parameters (outside scope of paper & presentation)

Missing Data Imputation



Figure: Example of imputation where black points are data from other missions, WBS 4 % is known and blue points show uncertainty on split between WBS 7 & 9, red are posterior predictive for future missions

- Some missions have incomplete WBS breakdowns
 - WBS 7 is sometimes bookkept in WBS 9
 - WBS 10 is sometimes bookkept in WBS 6
 - WBS 1, 2, 3 often not split rigorously
- Costs imputed using <u>multiple imputations</u> with a Dirichlet distribution on elements that are often bookkept together
- After imputation, the larger Dirichlet model across all WBS elements is fitted to imputed datasets and the posteriors are combined to account for imputation uncertainty

Cost Allocation Rule of Thumb Tool (CARoTT)



CARoTT Capabilities:

- 1. Predict unknown individual WBS costs
- 2. Predict total mission cost
- 3. Estimate probabilities of meeting cost targets
- 4. Visualize WBS correlations
- 5. Extract posterior samples
- 6. Choose what types of uncertainty to include

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	Bayesiar	n Rule of Thumb					Clas	ssical Rule of	Thumb		
lagship Ne	w Frontiers Discov	very MIDEX SMEX	Cube	eSat / Sn	nalisat	Earth Venture In	strument				
Da	awn Genesis GRAII	IL Insight Kepler LUC	Y Ma	rs Odys:	sey N	fars Pathfinder Mi	ESSENGER N	IEAR Phoer	nix Psyche Sta	rdust	
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Basic Input	S					Add Assumpt	ions				
Uncertainty T	/pe ? Epistemic C	Only			~	BCD WBS 01	BCD WBS 02	BCD W	BS 03 🗌 BCD W	BS 04	
	BCD WBS 01 BCD WBS 02 BCD WBS 03 BCD WBS 04	Input Uncertainty Form Point Uniform Input Value(s)	1 1 O Ti	riangular		BCD WBS 05 BCD WBS 10 EF WBS 07 Subtotal (E-F)	BCD WBS 06 EF WBS 01 EF WBS 09 Total (E-F)	BCD W EF WBS 0 Subtotal (B Total (A-	IBS 07 BCD W 12 EF WBS 03 -D) Total (B-D F)	BS 09 EF WBS 04 0)	
	BCD WBS 05	220		<>		Minimum	\bigcirc	Maximum	0		
	BCD WBS 07	235		\bigcirc			WBS Comb	ination	Minimum	Maximur	
	BCD WBS 09 BCD WBS 10 EF WBS 01 EF WBS 02 EF WBS 03 EF WBS 04 EF WBS 07 EF WBS 09	250		\sim		Add Targets					
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	 Subtotal (B-D) Total (B-D) 					Minimum	0	500		C Add Targe	
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Reserves (B-I) 30			0	%						
Reserves (E-F) 15			0	%						
			0	\$M (FY2	0)						
Launch Vehic	e										

1. Predict unknown individual WBS costs from one or more known WBS costs or a known total cost target

	sitory										
	Bayesian	Rule of Thumb			Classical Rule of Thumb						
lagship Nei	w Frontiers Discove	ary MIDEX	SMEX CubeSat / Si	nallsat Earth Ventur	e Instrument						
	Astrophysics		н	eliophysics		All					
Jncertainty Ty	Vpe ? Aleatoric an WBS 01 WBS 02	d Epistemic Input Uncertai	inty Form Uniform O Triangular	WBS 01 WBS 07 WBS	VBS 02 WBS 03 WB VBS 09 WBS 10 Sub	S 04 🗌 WBS (ototal (A-F) 📄	05 🗌 WBS 06 Total (A-F)				
Uncertainty Ty	vpe ? Aleatoric an	d Epistemic Input Uncerta	inty Form	WBS 01 WBS 07 W	/BS 02 WBS 03 WB	S 04 🗌 WBS (ototal (A-F) 📄	05 🗌 WBS 06 Total (A-F)				
	WBS 03			Minimum	Maximum	C.	Add Assumption				
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Input Variable Reserves (A-F)	 WBS 04 WBS 05 WBS 06 WBS 07 WBS 09 WBS 10 Subtotal (A-F) Total (A-F) 	Minimum Most Like Maximum *Selecting multi inputting a value elements*	aly 0 h 0 ple WBS elements represent a for the sum of the selected 0 9	Add Targets Select multiple for WBS 01 W WBS 07 W Minimum	WBS Combination r sum /BS 02 WBS 03 WB /BS 09 WBS 10 Sub @ Maximum WBS Combination	Minimum	05 UWBS 06 Total (A-F) Add Target Maximun				

Run CARoTT

2. Predict total cost from one or more known WBS element costs

0	100 200	300 Cost (\$	400 SM)	500	600
vier Dradiative Distribut	ion Cummon.				
rior Predictive Distribut	ion Summary				
		•			
Line Item	2.5th Percentile	30th Percentile	50th Percentile	70th Percentile	97.5th Percentil
Phase B-D WBS 01	0.478	5.366	9.418	15.611	45.60
Phase B-D WBS 02	1.089	7.372	12.363	19.376	53.69
Phase B-D WBS 03	0.109	2.773	5.93	10.682	36.00
Phase B-D WBS 04	0.355	4.266	7.95	13.6	42.63
Phase B-D WBS 05	17.533	37.387	47.377	58.225	91.37
Phase B-D WBS 06	133.26	166.769	177.49	187.685	207.71
Phase B-D WBS 07	0.769	6.471	10.986	17.631	47.86
Launch Vehicle	0	0	0	0	
Phase B-D WBS 09	1.197	7.758	12.843	19.747	52.08
Phase B-D WBS 10	0.995	7.202	12.475	19.387	53.90
Phase B-D Subtotal	265.023	297.927	314.971	336.201	419.14
Phase B-D Reserves	79.507	89.378	94.491	100.86	125.74
Phase B-D Total	344.529	387.305	409.462	437.062	544.88
Phase E-F WBS 01	0.237	2.081	3.501	5.441	15.37
Phase E-F WBS 02	0.002	0.343	0.925	1.911	8.4
Phase E-F WBS 03	0	0.19	0.604	1.454	7.33
Phase E-F WBS 04	6.136	13.821	17.947	22.862	41.97
Phase E-F WBS 07	8.148	17.225	22.055	27.802	49.2
Phase E-F WBS 09	0.38	2.392	3.894	6.186	16.13
Phase E-F Subtotal	25.705	44.132	53.458	64.228	103.1
Phase E-F Reserves	3.856	6.62	8.019	9.634	15.46
Phase E-F Total	29.561	50.751	61.477	73.862	118.58
Phase A	0	0	0	0	
Phase A-F Total	388.345	445.209	473.205	507.28	628.63

Posterior Exploration

3. Estimate probabilities of simultaneously meeting one or more cost targets



4. Visualize marginal or two-dimensional joint predictive densities



Last Update: February 1, 2023 Questions & Comments: Patrick Bjornstad - patrick.t.bjornstad@jpl.nasa.gov JPL is a federally funded research and development center staffed and managed for NASA by Caltech

29,906

6.152

39.92

2.051 76.906

147.465

29,708 0 22,865

5. Extract a large sample of the full posterior predictive distribution



5.22

192

108.058

5.257 3.356 0.582 41.383

468.25

23.779

78.

4.114

6. Choose what types of uncertainty they want to include (aleatoric only, epistemic only, aleatoric and epistemic, neither aleatoric nor epistemic)



Posterior Predictive Distribution

Export

BCD WBS 01	BCD WBS 02	BCD WBS 03	BCD WBS 04	BCD WBS 05	BCD WBS 06	BCD WBS 07	LV	BCD WBS 09	BCD WBS 10	BD Subtotal	BD Reserves	BD Total	EF WBS 01	EF WBS 02	EF WBS 03	EF WBS 04	EF WBS 07	EF WBS 09	, Subto
29.906	6.152	39.92	2.051	76.906	147.465	29.708	0	22.865	5.22	360.192	108.058	468.25	5.257	3.356	0.582	41.383	23.779	4.114	78.4
39.125	15.155	7.774	1.449	30.466	195.808	24.155	0	13.573	9.793	337.3	101.19	438.49	16.185	0.52	2.272	26.723	16.312	18.897	80.
25.756	3.124	0.353	2.536	67.992	158.567	25.67	0	2.162	25.295	311.454	93.436	404.89	0.942	0.186	0.049	7.872	13.419	11.125	33.5
13.857	3.319	14.43	4.828	46.48	180.022	12.275	0	2.958	25.675	303.842	91.153	394.995	18.377	15.6	0	32.496	18.114	12.295	96.8
19.469	5.885	7.275	12.159	90.228	134.548	11.993	0	3.66	8.038	293.255	87.976	381.231	0.93	0.075	0.035	9.506	15.123	3.279	28.9
1.642	22.59	8.875	6.788	30.234	191.02	14.39	0	7.249	11.742	294.53	88.359	382.889	2.006	0.045	0.542	20.854	29.848	2.569	55.8
1.963	8.521	0.104	4.278	58.558	168.181	6.838	0	3.772	9.534	261.749	78.525	340.274	4.475	3.683	2.471	12.835	26.475	1.406	51.3
0.453	7.554	0.229	1.416	47.995	181.313	4.215	0	20.556	3.74	267.472	80.242	347.713	8.585	0.208	6.851	11.907	24.196	7.068	58.8
20.471	9.967	10.505	14.258	44.2	181.406	25.332	0	20.112	11.364	337.615	101.285	438.9	6.284	0.996	0.067	30.82	34.575	4.304	77.0
16.971	7.513	4.68	0.623	64.032	159.544	3.737	0	23.397	6.753	287.25	86.175	373.425	4.121	2.161	0.194	28.164	14.054	7.902	56.5
																~	< 1	/ 400	>

Using Aleatoric vs. Epistemic mode?

Neither

- Traditional ROT validation
- Want to show example allocation for a given mission class and cost cap

Aleatoric Only

- Assuming we have the right mean, how much do projects vary around that mean?
- Probably not very useful in practice



Epistemic Only

- Estimate of WBS means based on data
- Model what future cost models may predict

Aleatoric & Epistemic

- Quantify as much uncertainty as possible to assess probability of achieving a certain cost
- Correlation between WBS
 elements

Conclusion & Future Work

Conclusion

- · We can and should provide more robust cost estimates with an understanding of uncertainty
- This is important at all stages of project formulation, including very early project formulation
- Back-of-the-envelope calculations which use average ratios between WBS elements may provide engineers a rough sense of whether a concept is feasible or not, but comes with no insight into the risk of overrunning cost
- Here we have shown we can ingest high level cost allocation data with missing or incomplete information and generate probabilistic predictions of allocations

Future Work

- Future work should continue to improve upon the selection of an informative prior
- Other applications of the Dirichlet model: mass budgets, workforce allocation models for center management



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Questions? Contact Melissa Hooke (Melissa.A.Hooke@jpl.nasa.gov)

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