



Rolling the Dice

Using the Monte Carlo Method
to Optimize Solids Management

Kevin Campanella, Burgess & Niple

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Acknowledgments

Project Team

- City of Columbus:
 - Patrick Eiden
 - Josh Lutz
 - Todd Krenelka
 - Heather Curtis
 - Brandon Fox
- Black and Veatch: Bob O'Bryan
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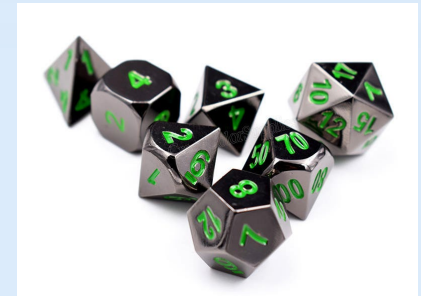
What if you want to...



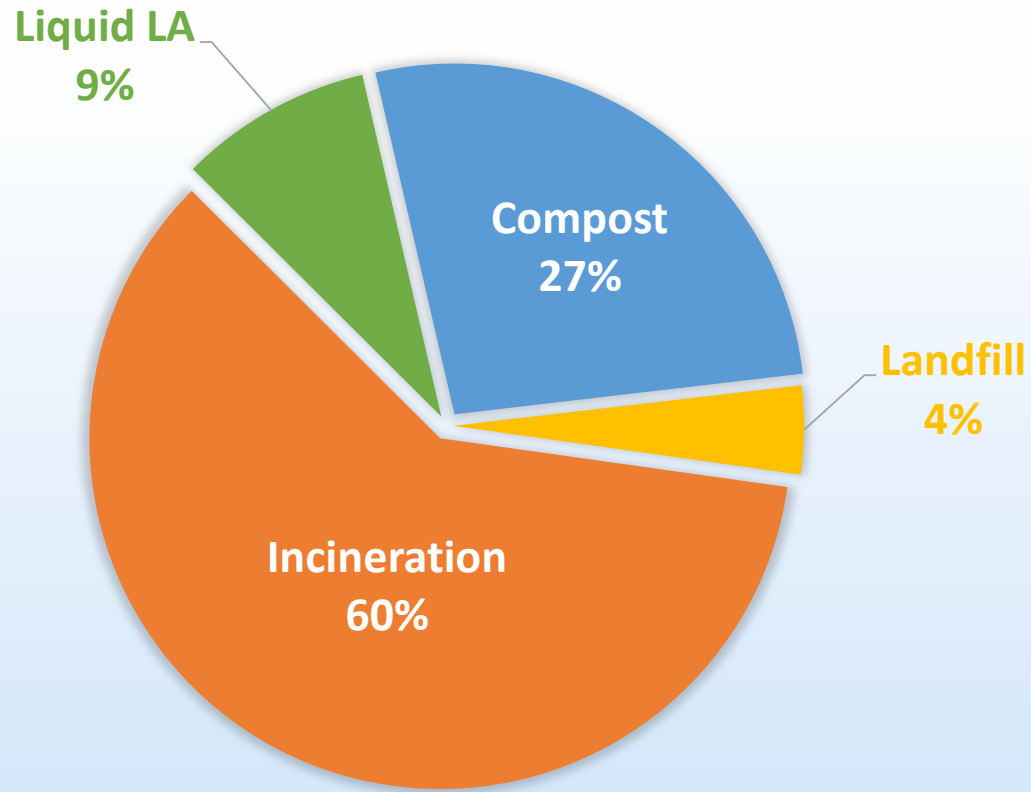
- Model risks of a project or program?
 - Multiple inputs
 - Inputs are complex and variable
 - Potential outcomes are broad ranging

Monte Carlo Analysis

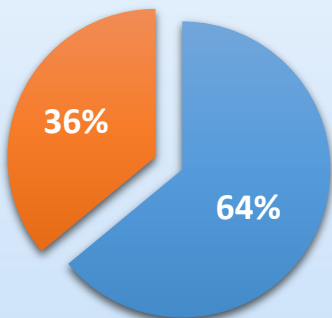
- Monte Carlo Analysis Method
 - Can model many complex input variables
 - Can model “what if” scenarios quickly
 - Typical Monte Carlo analysis involve 5,000+ simulations
 - Produces an understanding of each possible outcome and its likelihood
 - Results help optimize investments and risks



Columbus WW Solids Distribution 2010

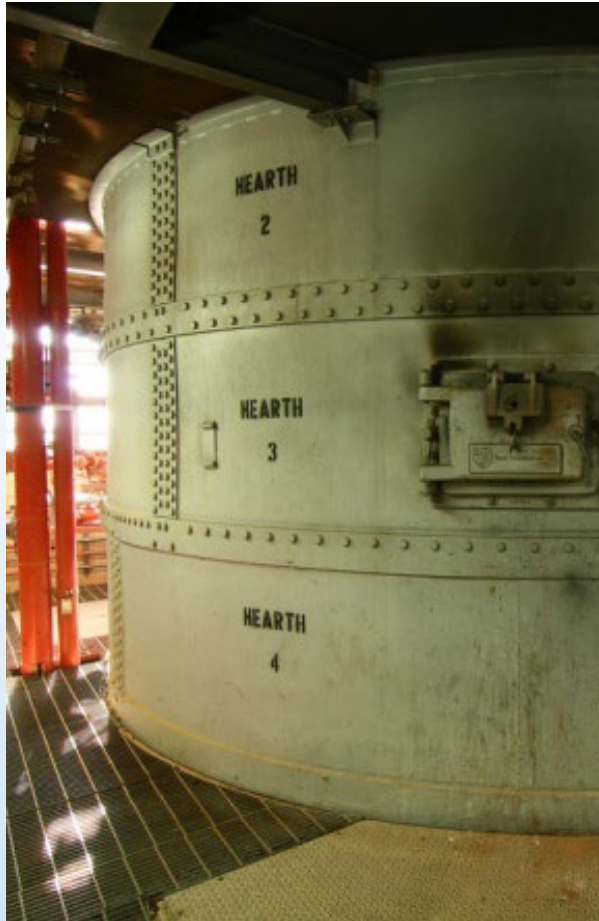


Average Solids
Production, 25K DT



■ Class B ■ Unclassified

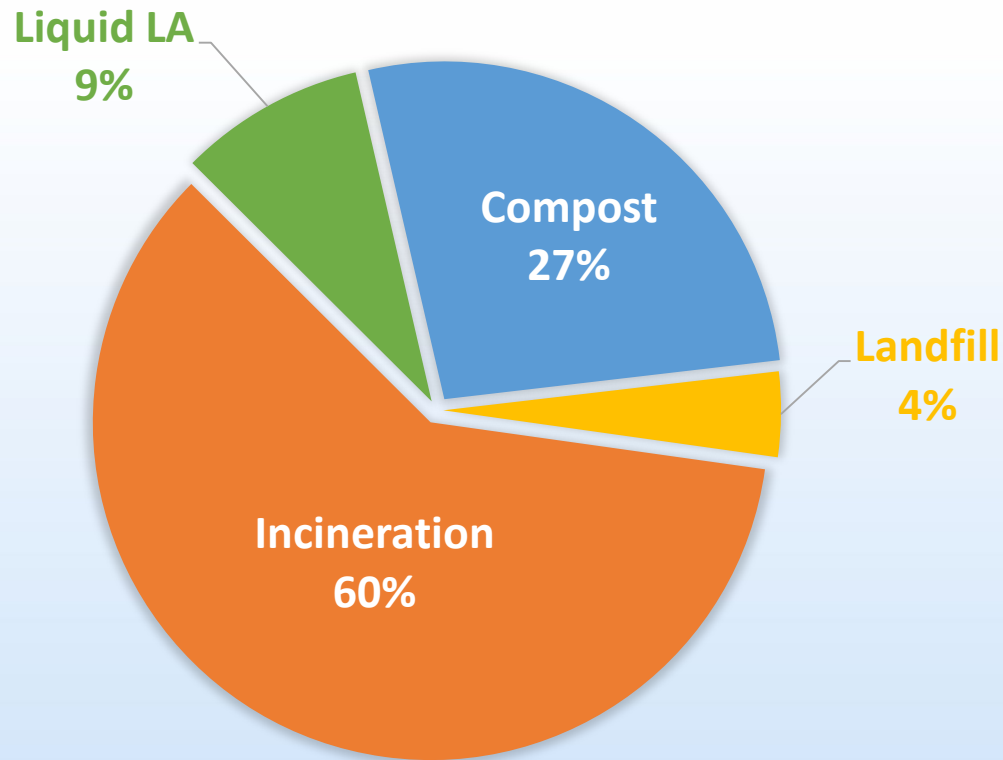
MACT Compliance



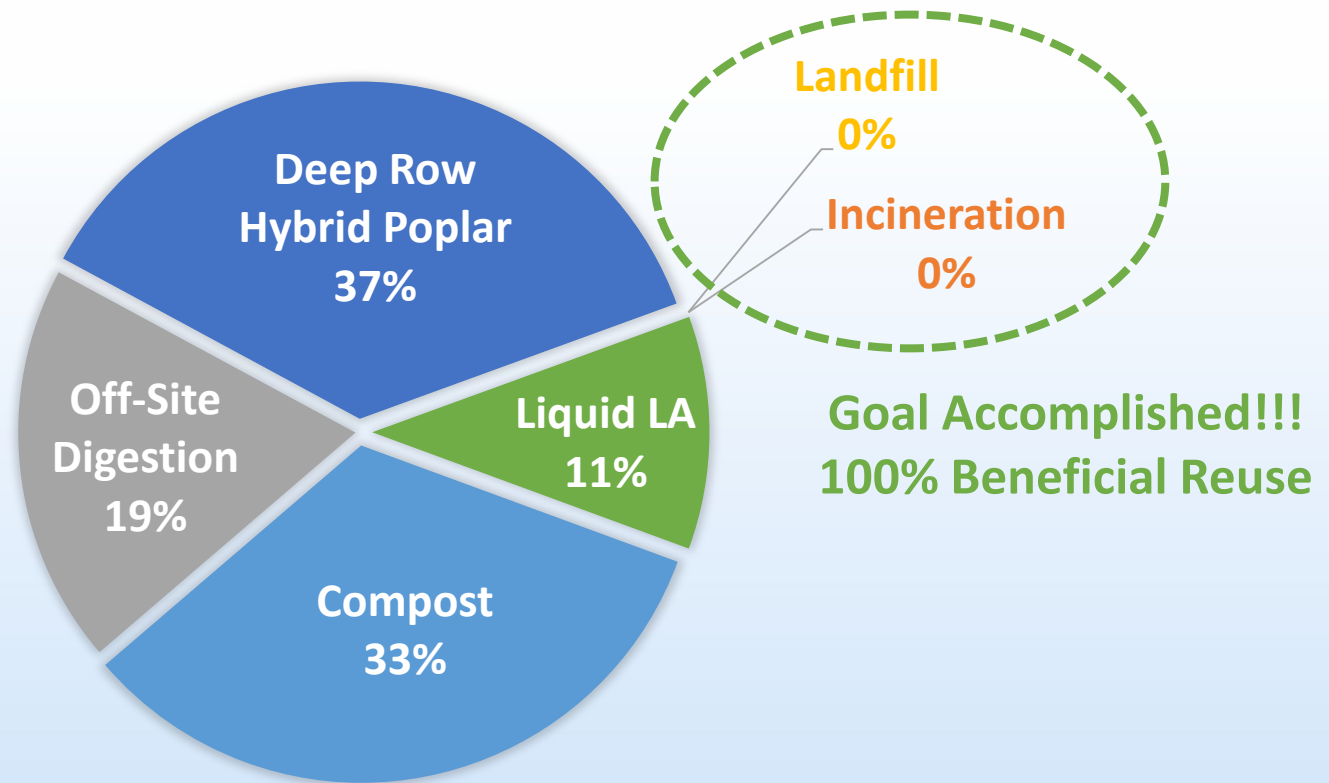
Southerly WWTP Multiple Hearth Incinerator (MHI)

- 2010 Title 5 Maximum Achievable Control Technology Standards (MACT)
- MHI Condition Assessment and BCE
 - Initial Goal: Determine what incinerator repairs are necessary at both plants.
 - Revised Goal: Determine the optimal number of incinerators to improve based on available capacity of ALL management outlets.
 - Comprehensive system approach.
- Findings
 - *Incineration improvements not necessary with an expanded beneficial reuse program.*

Columbus WW Solids Distribution 2010

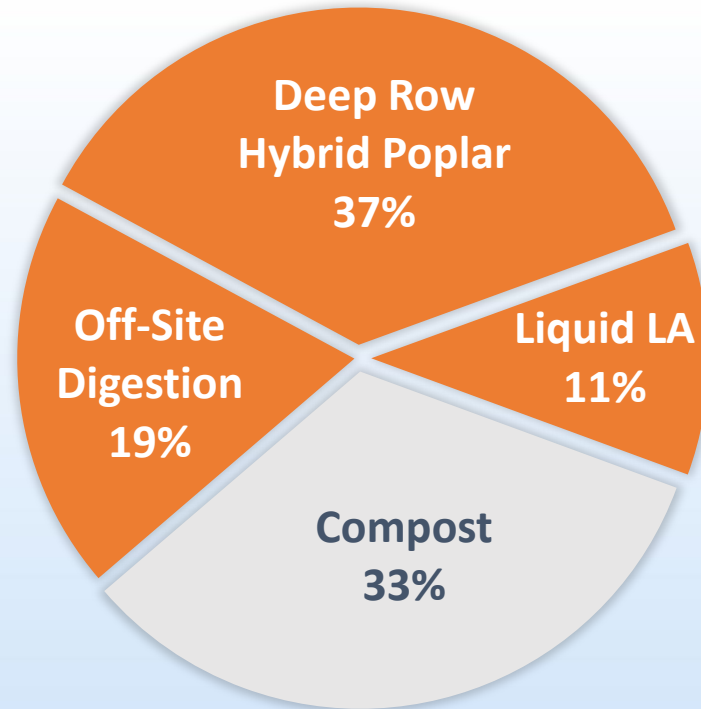


Columbus WW Solids Distribution 2017



Columbus WW Solids Distribution 2017

**Heavy Reliance on Private Contractors
for Solids Management.**

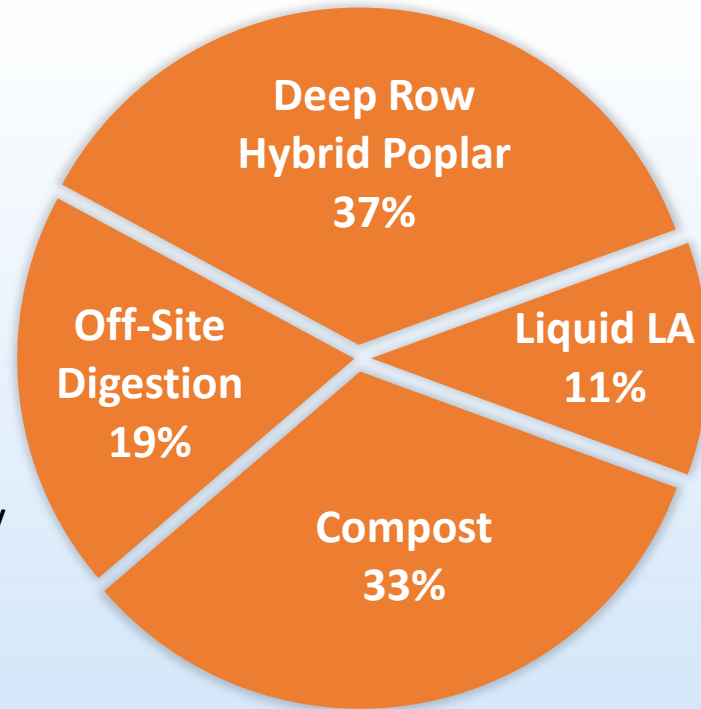


Columbus WW Solids Distribution 2017

Heavy Reliance on Private Contractors for Solids Management.

... a management outlet goes out of business?

... a regulatory change affects the City's ability to direct solids to a given outlet?



What if...

... an unexpected digestion outage creates a large short-term increase in unclassified solids production?

... an economic driver changes the reliability or capacity of a given management outlet?

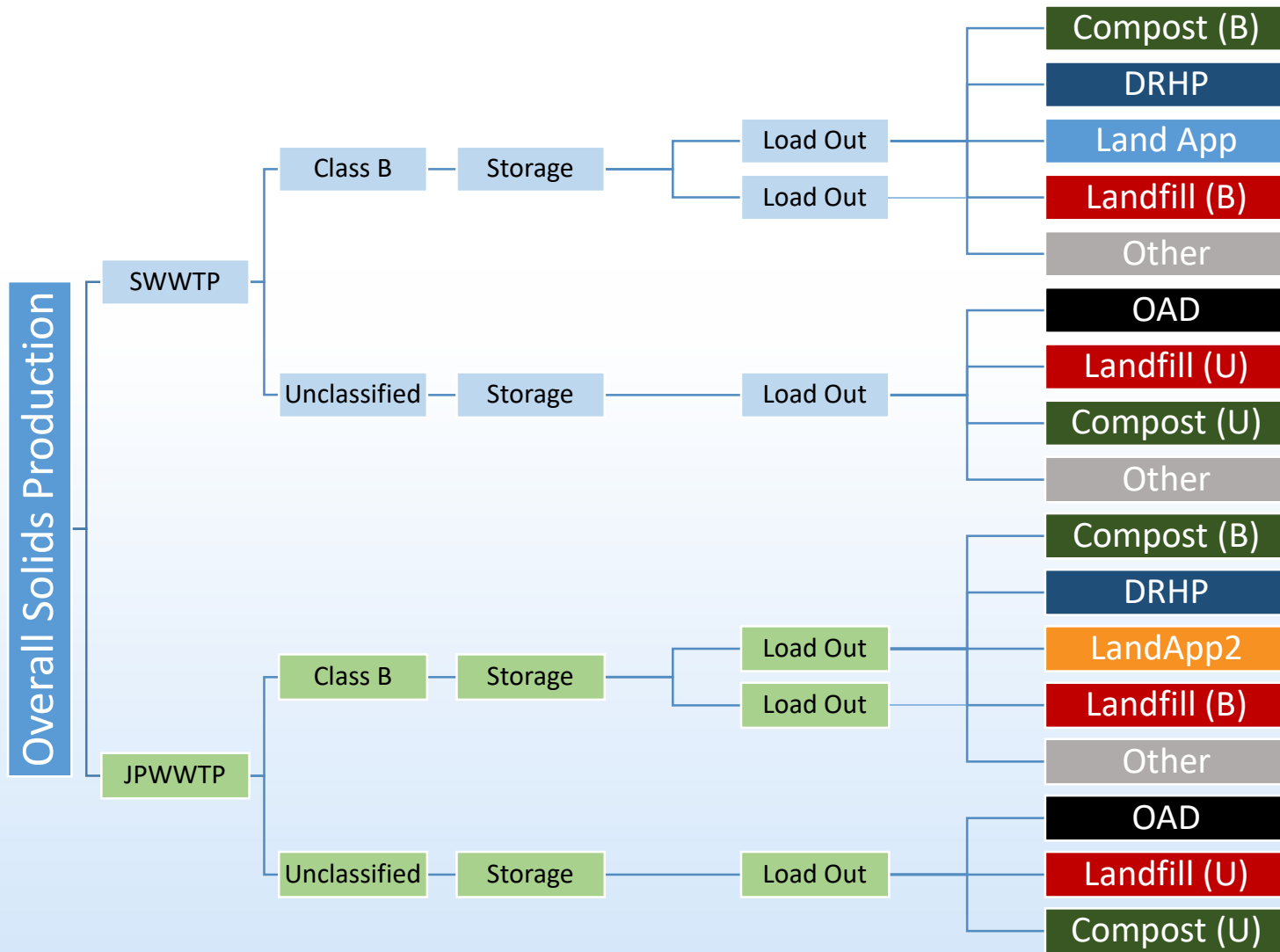
What are the range of possible outcomes for that scenario?

Can the management system "weather the storm"?

Modeling Solids Handling

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Key:
 DRHP = Deep Row Hybrid Poplar Mine Reclamation
 OAD = Offsite Anaerobic Digestion
 Land App = Land Application

Monte Carlo Model Overview

Models

Daily

- Production of solids (2 plants, 2 types), based on empirical data
- Solids outlet capacity based on empirical data
- Distribution of solids to six outlets
- BLAF storage utilization at both plants

Monthly

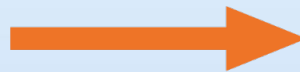
- Digester Reliability/ Catastrophic Failures

Annually

- Costs

Allows for **User-Defined** scenarios

Simulates a full calendar year



5,000
times

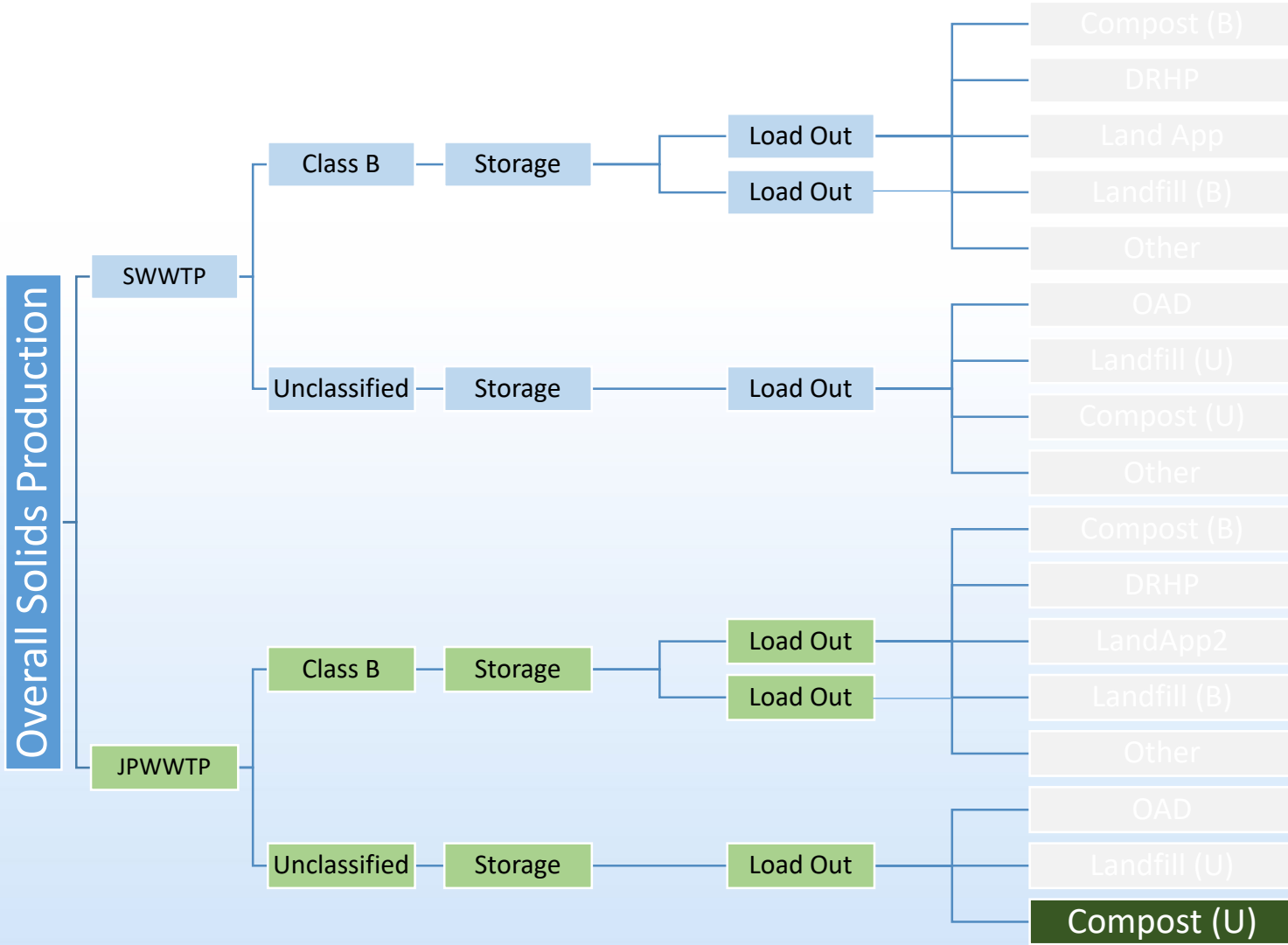
Solids Planning and Risk Evaluation (SPARE) Tool

	A	B	C	D	E	F	G	H	I	J	K	L	M
1													
2						Baseline Solids Production Data (assuming Digesters operational)							Adjusted Daily
3	Day of the week	Count of Days	Month	Week	Date	Overall Solids Production (DT)	JP Class B Solids Production (DT)	SWWTP Class B Solids Production (DT)	JP Unclassified Solids Production (DT)	SWWTP Unclassified Solids Production (DT)	JP Digesters Out of Service for (#) More Days	SWWTP Digesters Out of Service for (#) More Days	JP Class B Production (DT) adjusted for Digester Outage
4	Sunday	1	1	1	1/1/2017	65.95	25.06	23.96	0	16.93	0	0	25.06
5	Monday	2	1	1	1/2/2017	70.98	18.25	22.18	0	30.55	0	0	18.25
6	Tuesday	3	1	1	1/3/2017	56.99	14.13	22.68	0	20.18	0	0	14.13
7	Wednesday	4	1	1	1/4/2017	56.55	16.70	20.79	0	19.06	0	0	16.70
8	Thursday	5	1	1	1/5/2017	77.60	30.22	23.36	0	24.02	0	0	30.22
9	Friday	6	1	1	1/6/2017	61.88	31.46	11.18	0	19.23	0	0	31.46
10	Saturday	7	1	1	1/7/2017	75.56	24.22	33.86	0	17.49	0	0	24.22
11	Sunday	8	1	2	1/8/2017	74.07	24.28	28.32	0	21.47	0	0	24.28
12	Monday	9	1	2	1/9/2017	68.70	30.75	20.46	0	17.49	0	0	30.75
13	Tuesday	10	1	2	1/10/2017	61.65	18.45	22.96	0	20.23	0	0	18.45
14	Wednesday	11	1	2	1/11/2017	76.43	31.92	20.79	0	23.73	0	0	31.92
15	Thursday	12	1	2	1/12/2017	70.73	25.44	23.82	0	21.47	0	0	25.44
16	Friday	13	1	2	1/13/2017	72.83	27.63	24.98	0	20.23	0	0	27.63
17	Saturday	14	1	2	1/14/2017	57.34	19.86	20.00	0	17.49	0	0	19.86
18	Sunday	15	1	3	1/15/2017	74.95	31.17	28.68	0	15.11	0	0	31.17
19	Monday	16	1	3	1/16/2017	79.14	25.02	20.46	0	33.66	0	0	25.02
20	Tuesday	17	1	3	1/17/2017	69.63	19.93	17.72	0	31.98	0	0	19.93
21	Wednesday	18	1	3	1/18/2017	59.45	14.74	24.21	0	20.50	0	0	14.74
22	Thursday	19	1	3	1/19/2017	59.52	18.03	21.63	0	19.87	0	0	18.03

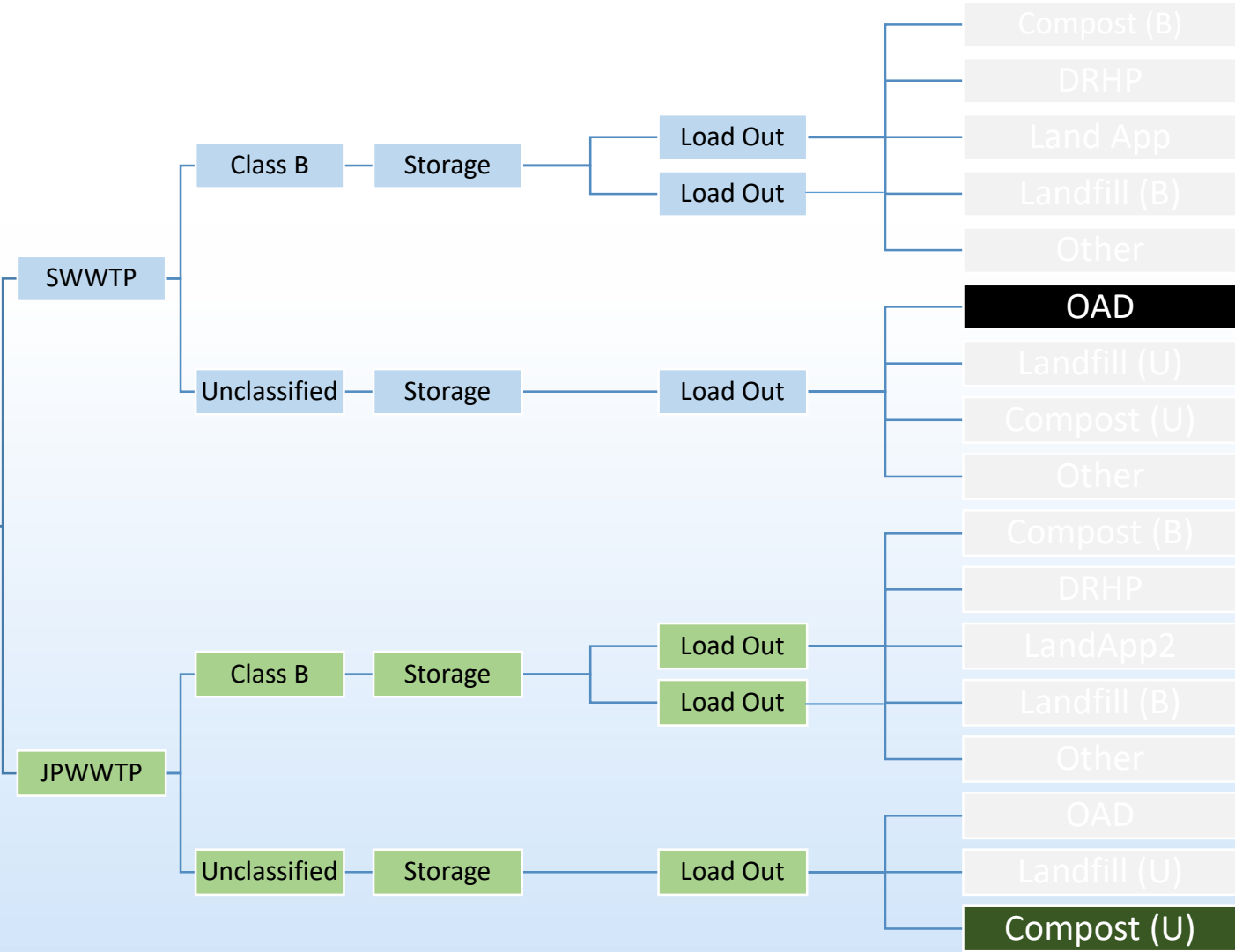
Modeling Undigested (Unclassified) Solids Handling

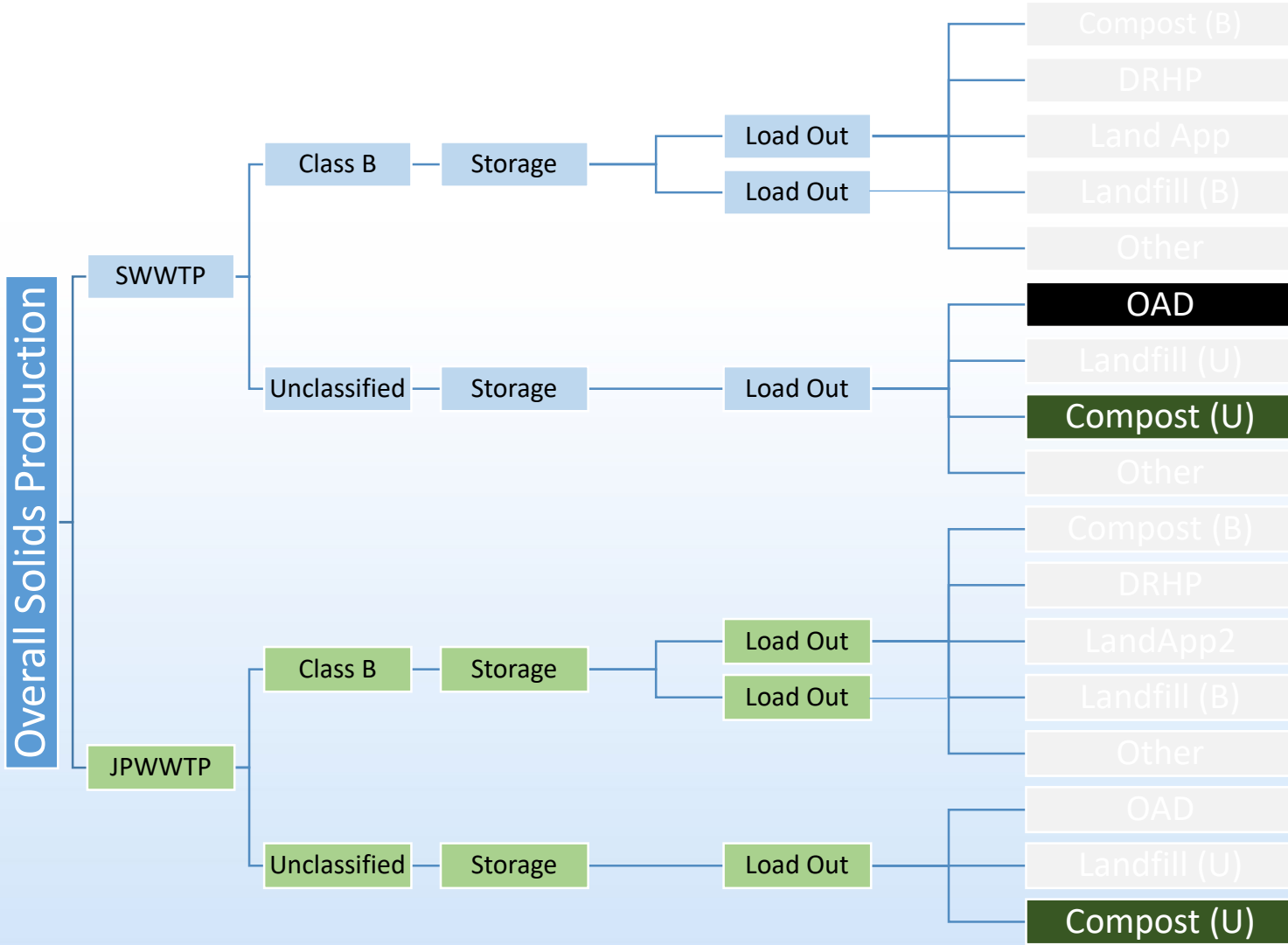
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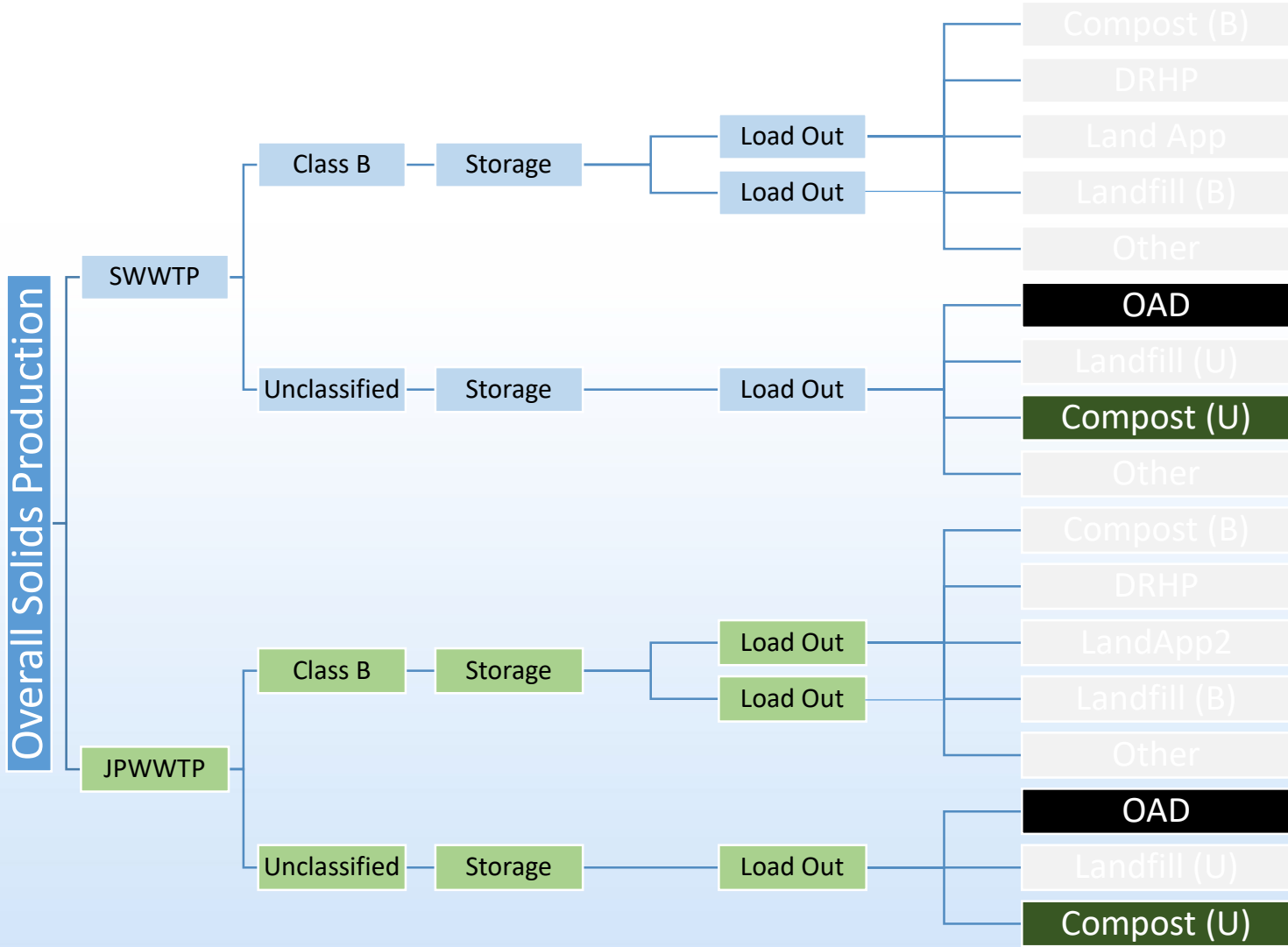


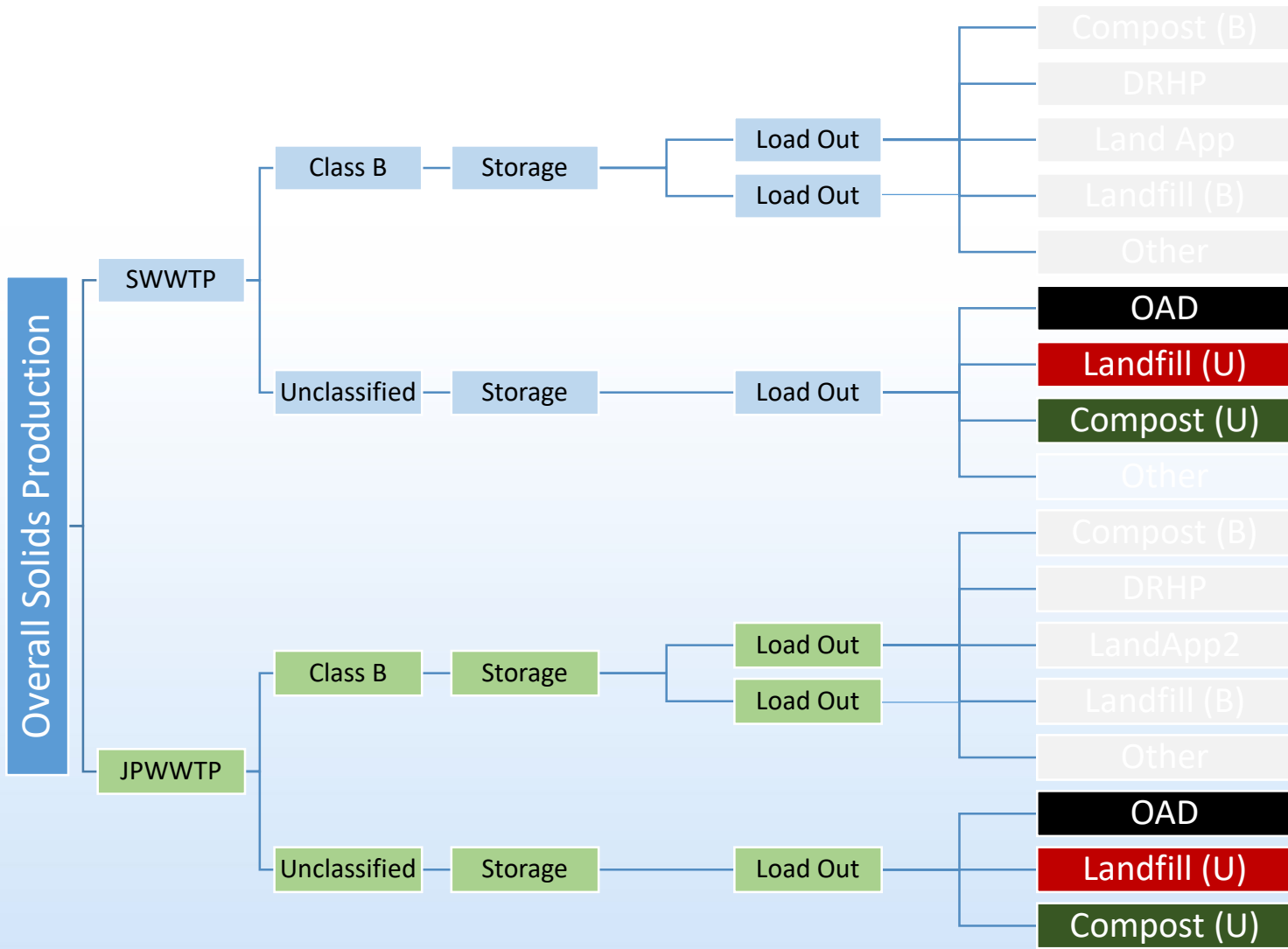


Overall Solids Production

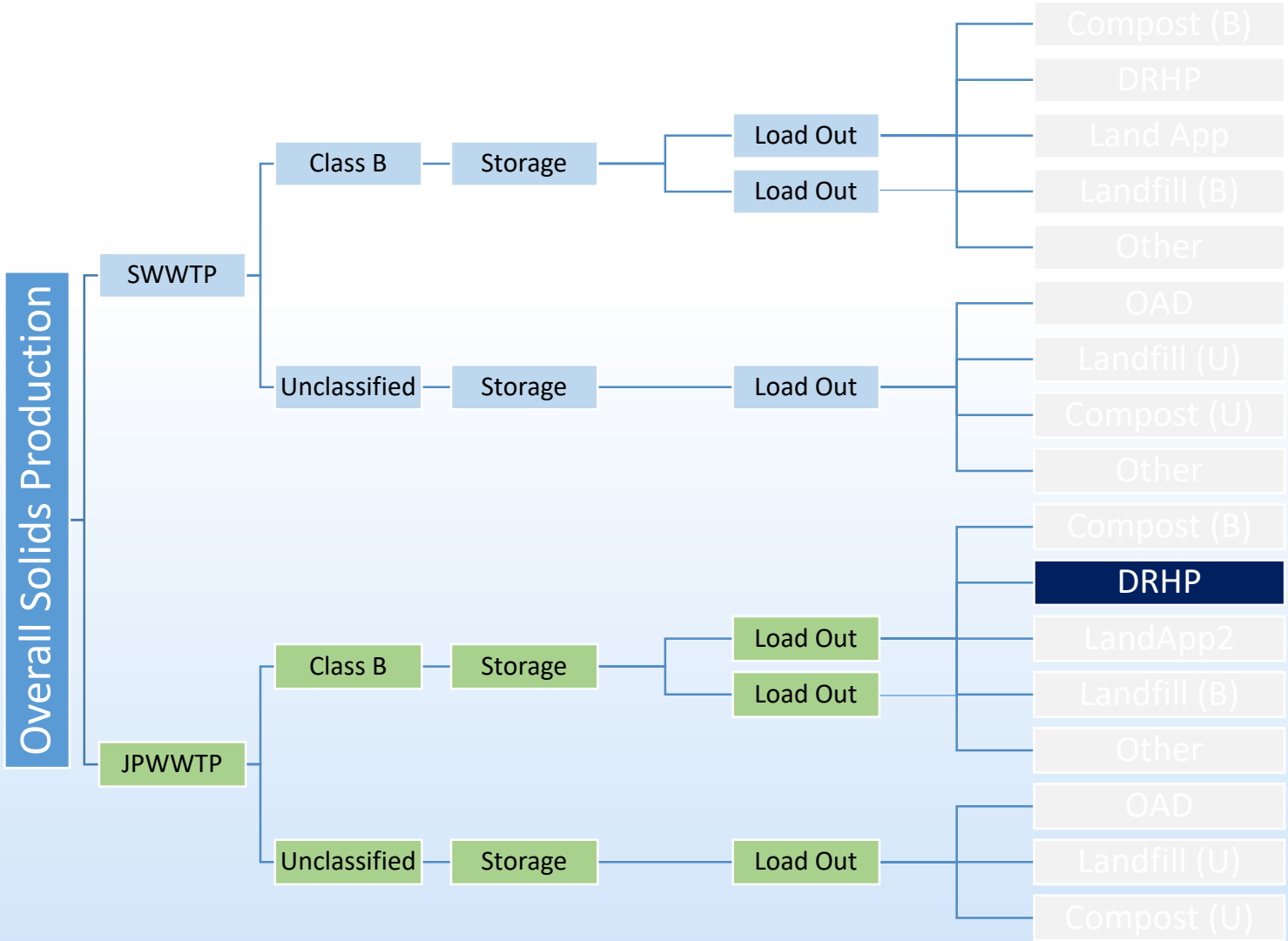


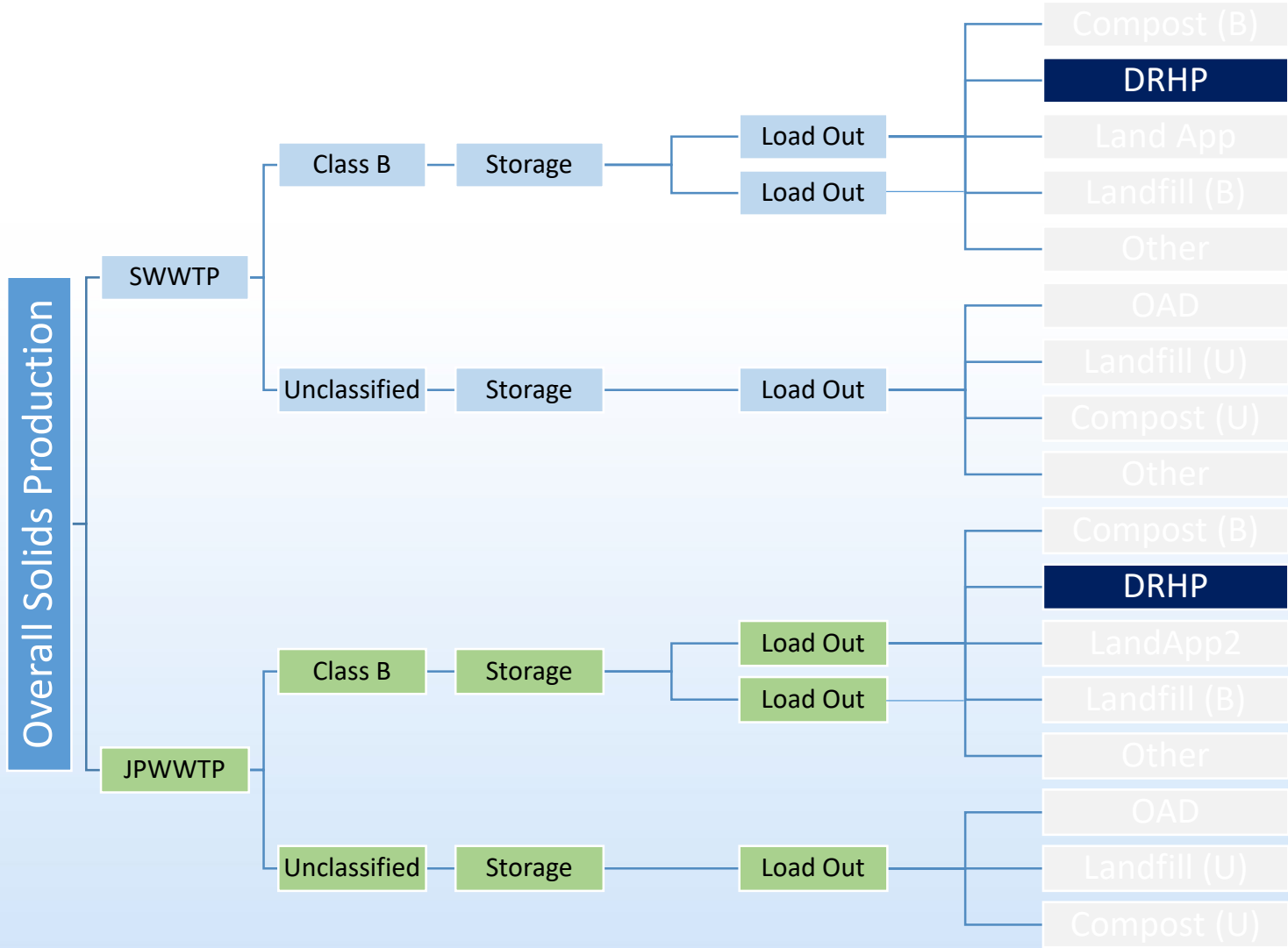






Modeling Digested (Class B) Solids Handling





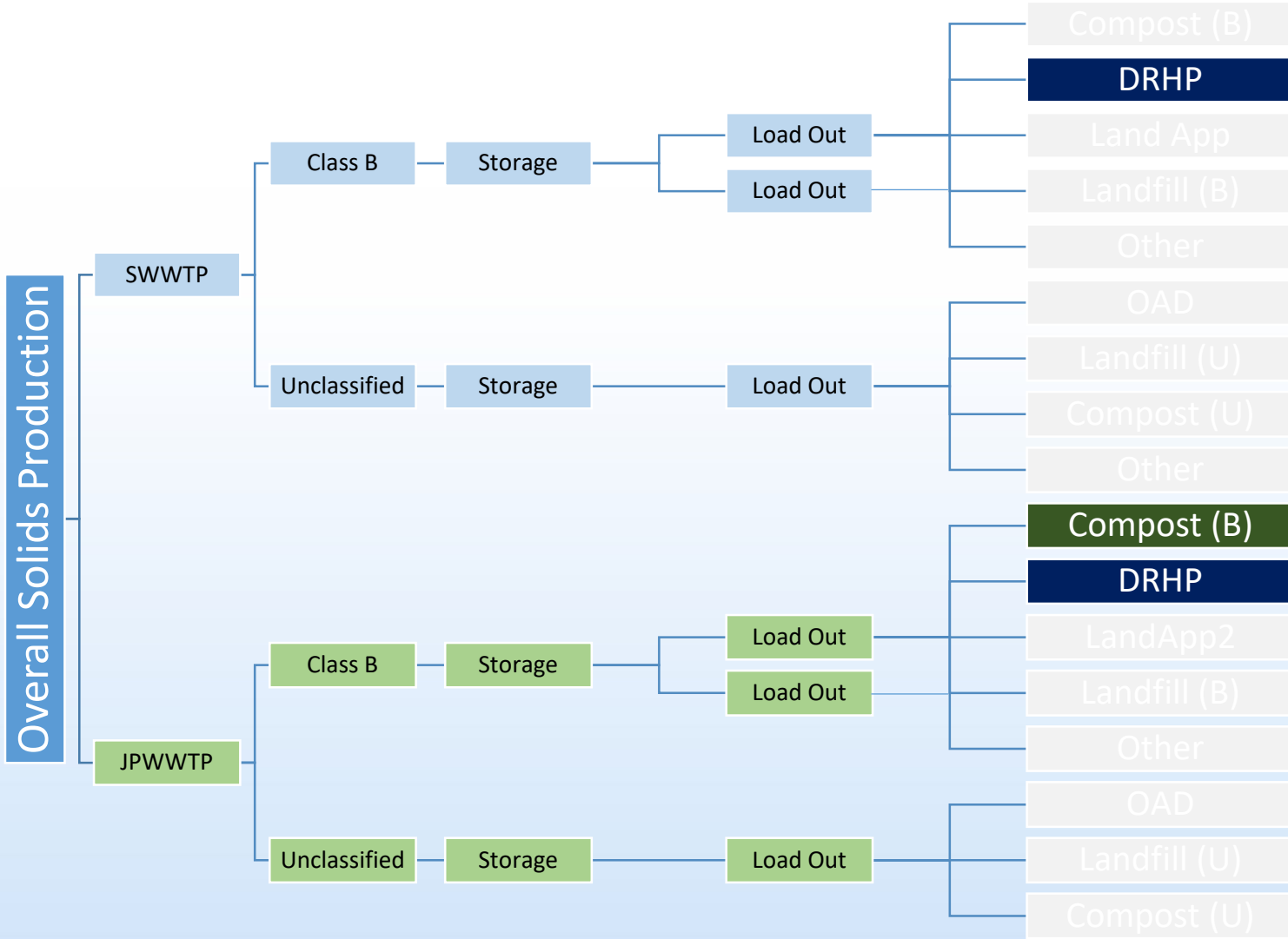
Class B Solids to DRHP

Baseline Circumstances

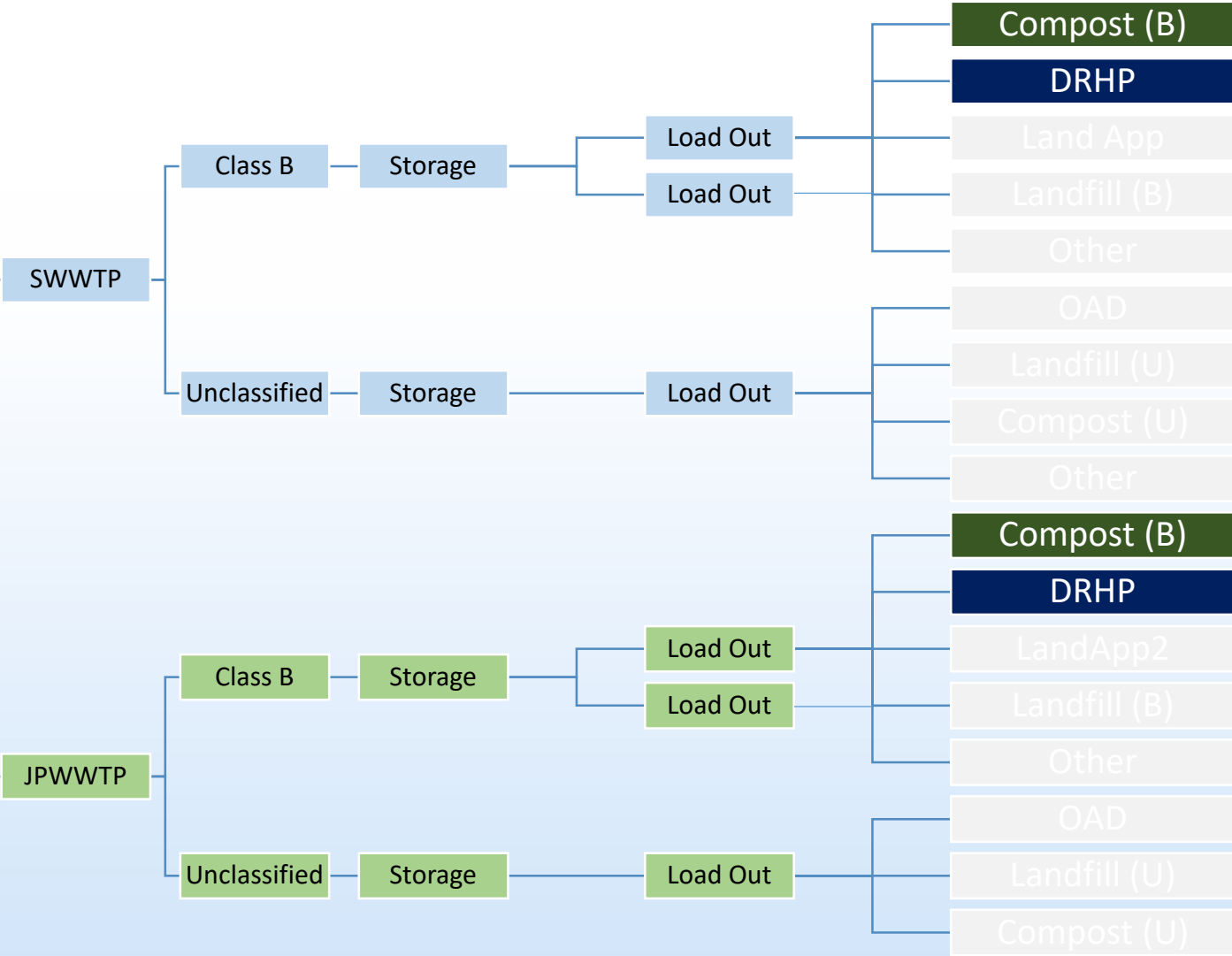
DRHP capacity
dedicated to JP
Class B first

If available storage at SWWTP < JP

DRHP capacity is
shared equally



Overall Solids Production



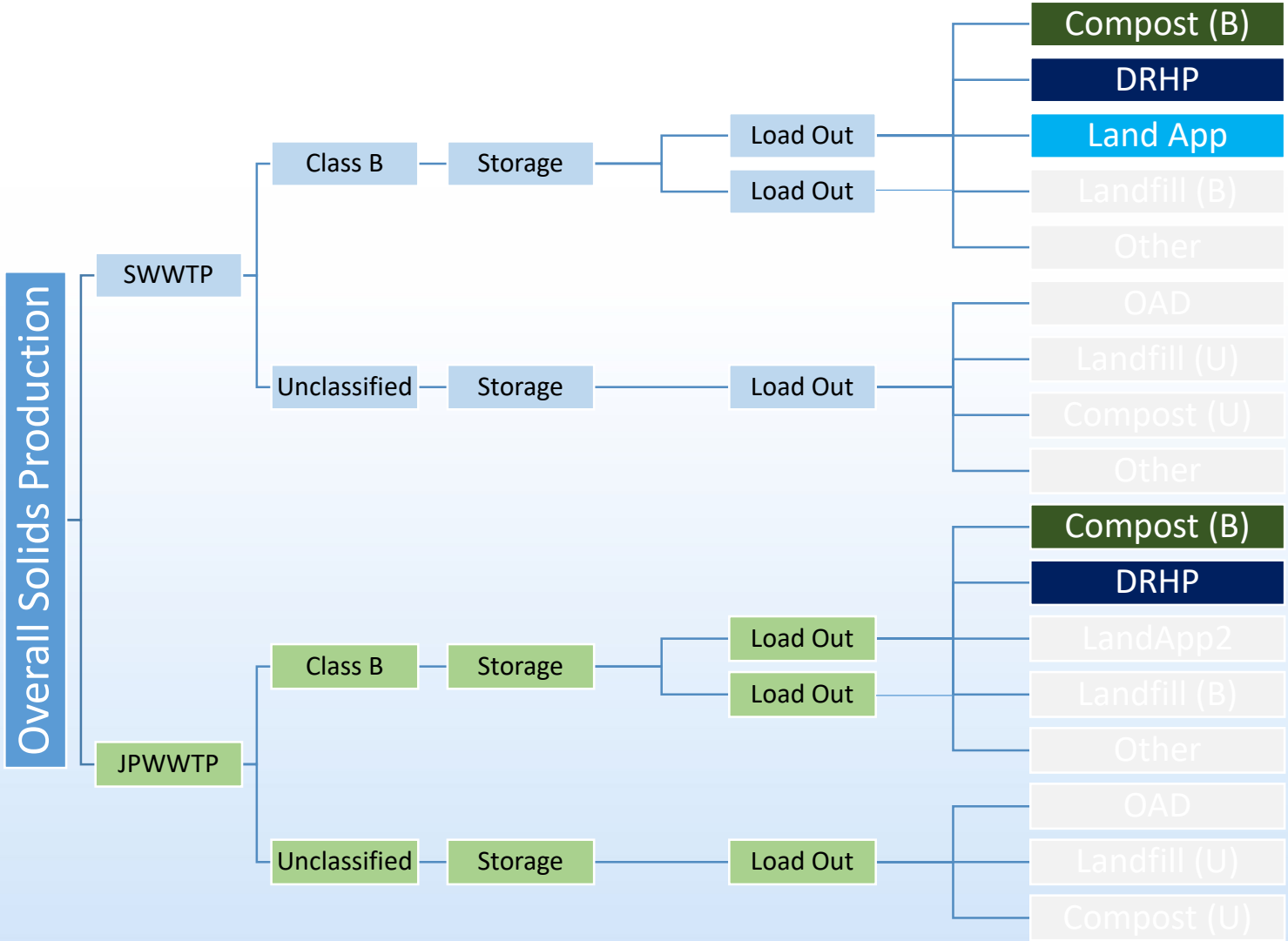
Class B Solids to Compost

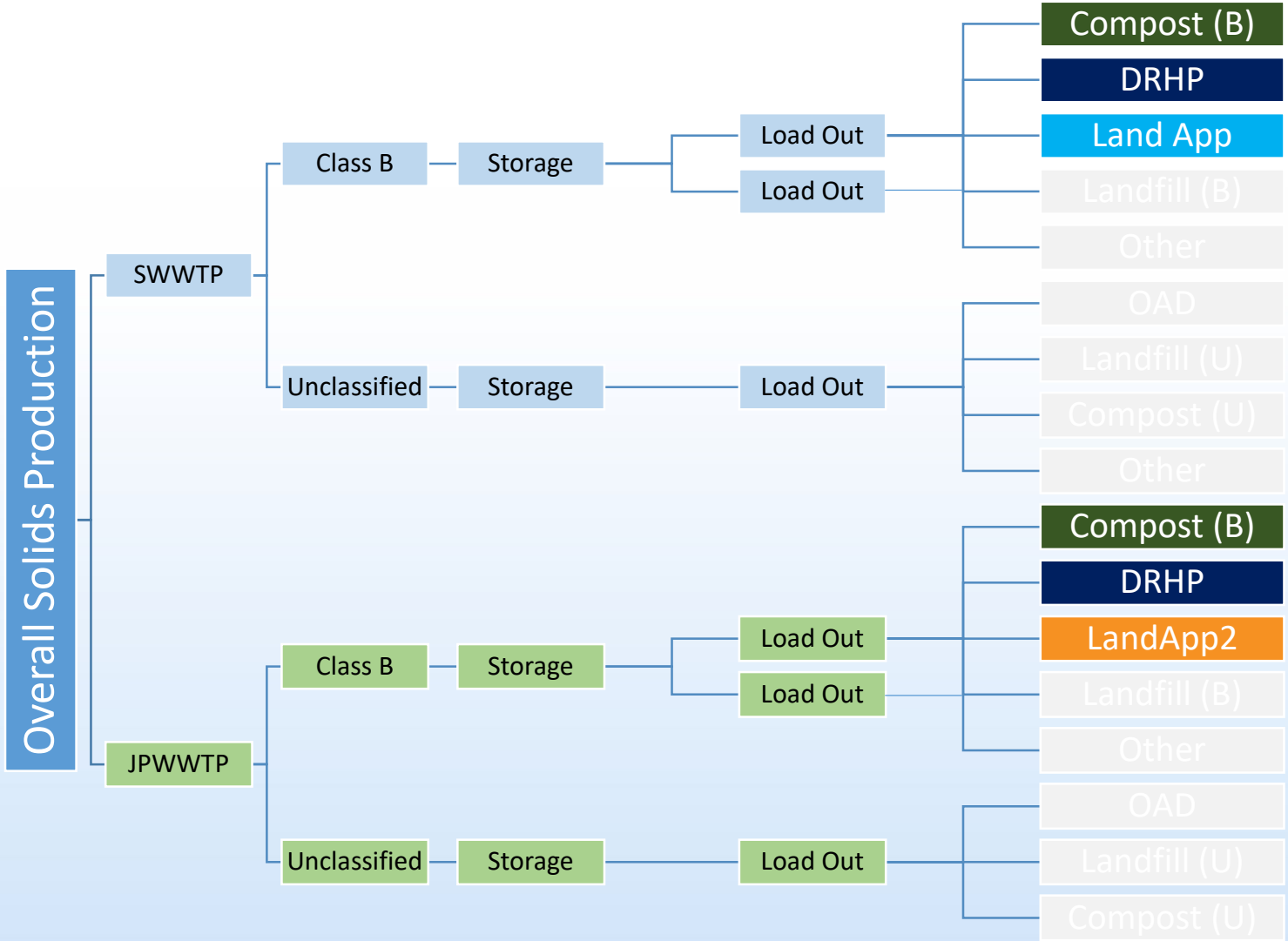
Baseline Circumstances

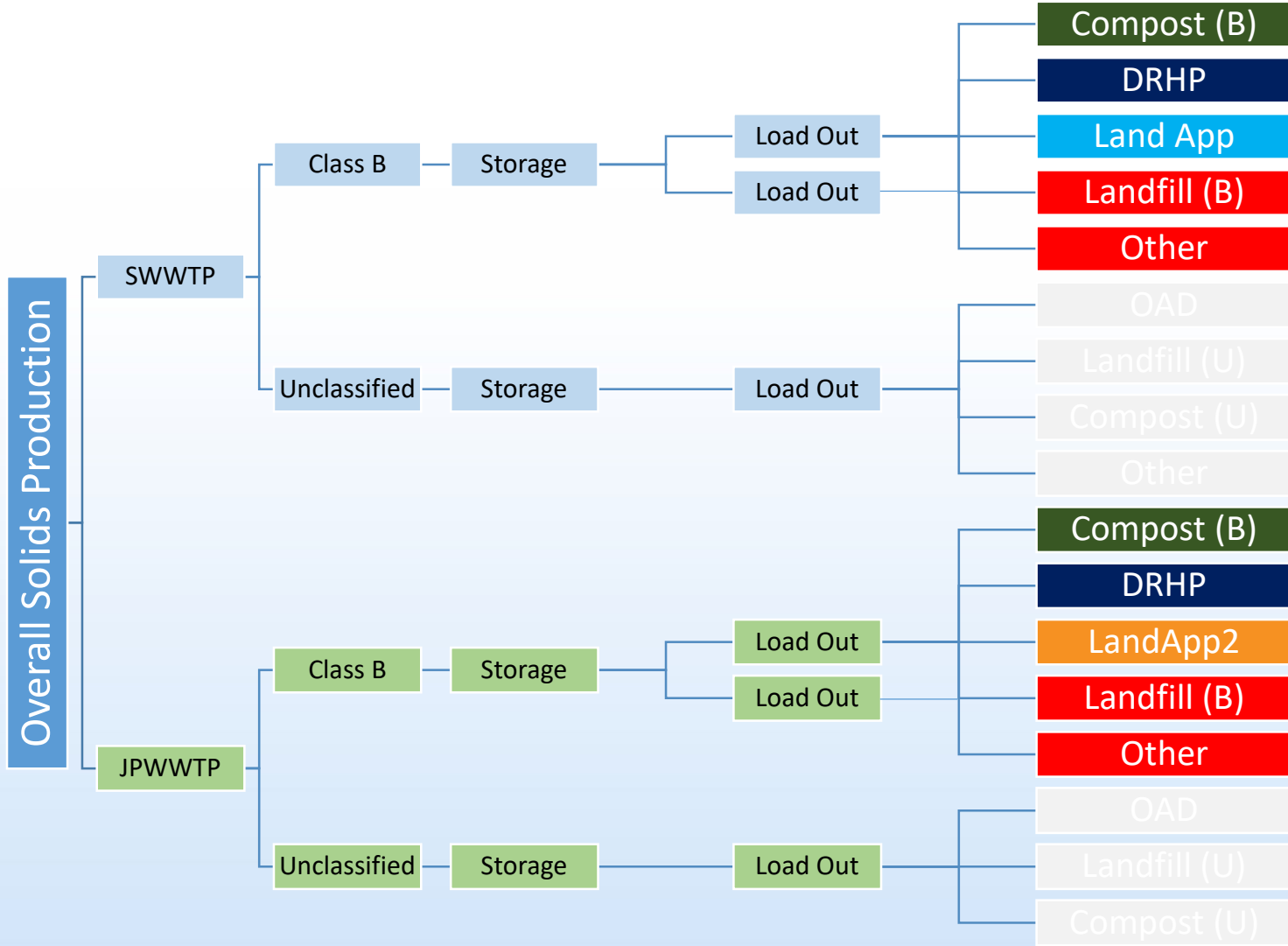
Compost capacity
dedicated to JP
Class B first

If available storage
at SWWTP < JP

Compost capacity
dedicated to
SWWTP Class B
first







Modeling Methods

Conventional Static Design Scenarios vs. Monte Carlo

Static Design v. Monte Carlo?



- Static Modeling involves the user to define discrete scenarios
 - Average (Solids Production, Digester Reliability, Outlet Availability)
 - Worst Case / Worst Year

But what is a hypothetical worst year?

Defining the Worst Case

Compost production is down due to construction

It won't stop raining and I can't land apply

Solids production is high at both plants

Digesters reliability is down

My primary outlet went out of business



Defining the Worst Case



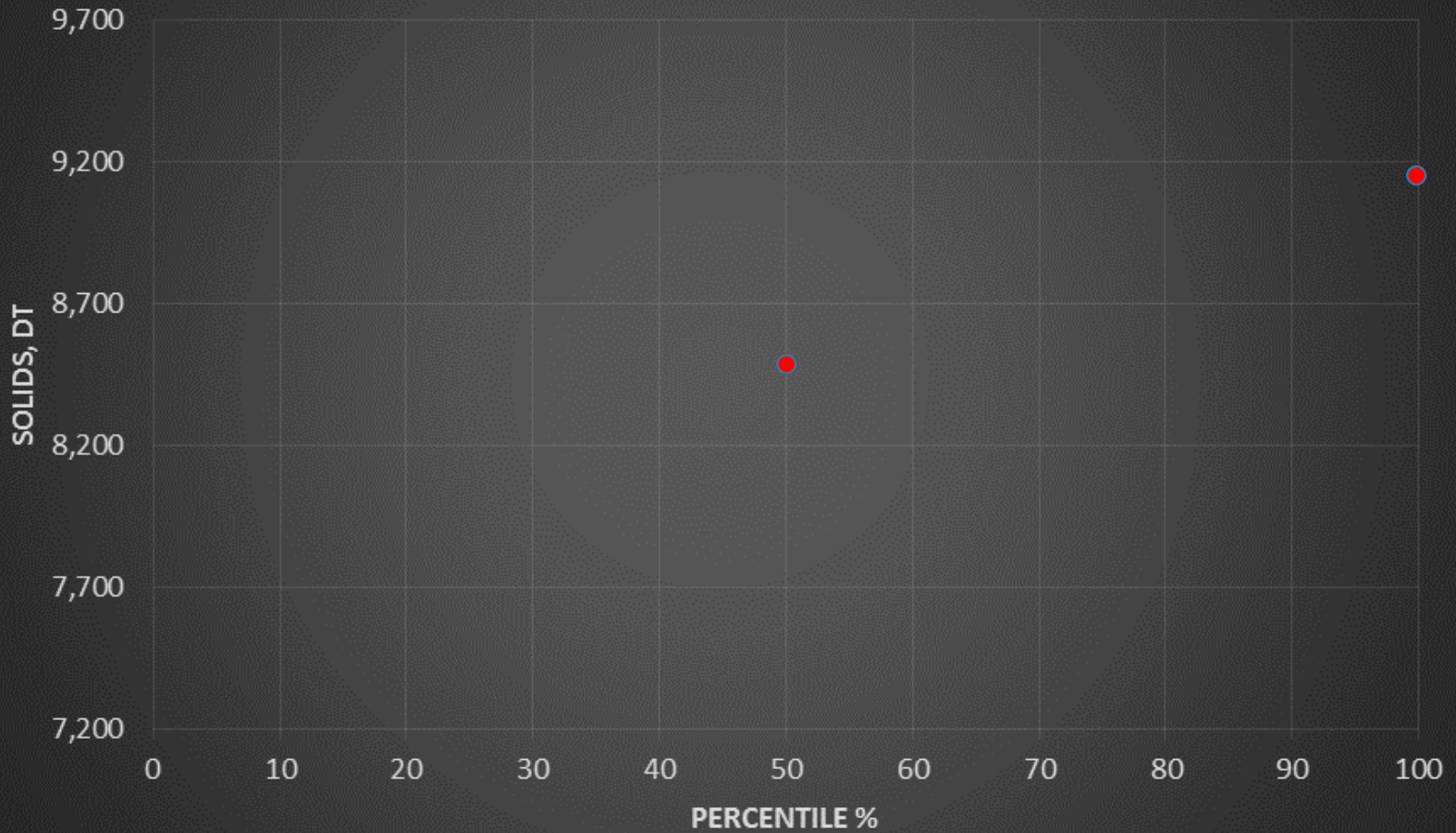
Why Apply Monte Carlo?

Some benefits of Monte Carlo

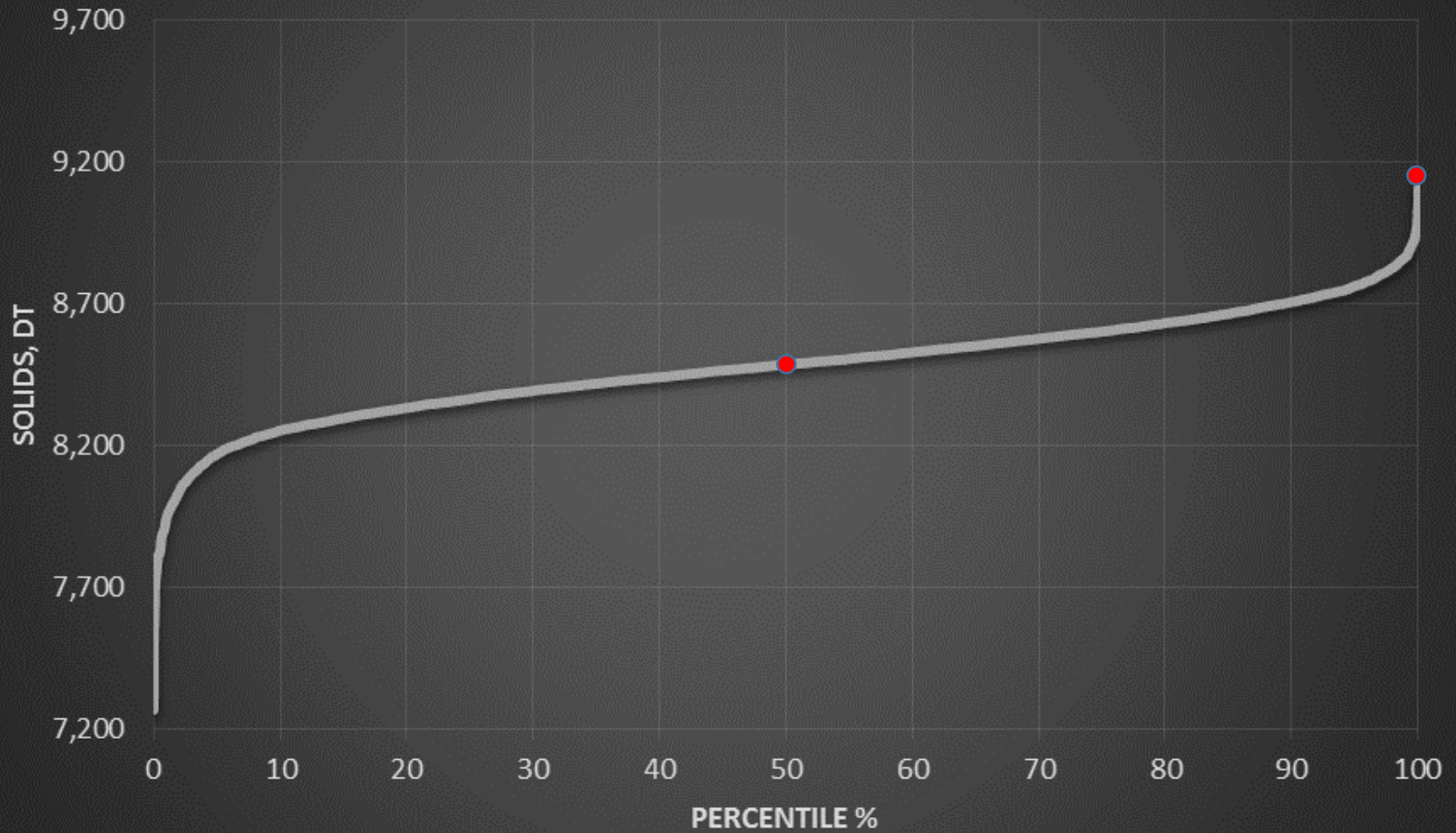


Monte Carlo not only answers these questions, it eliminates the need to ask the questions in the first place, saving time and money.

Without MC: JP Class B Solids Production



With MC: JP Class B Solids Production

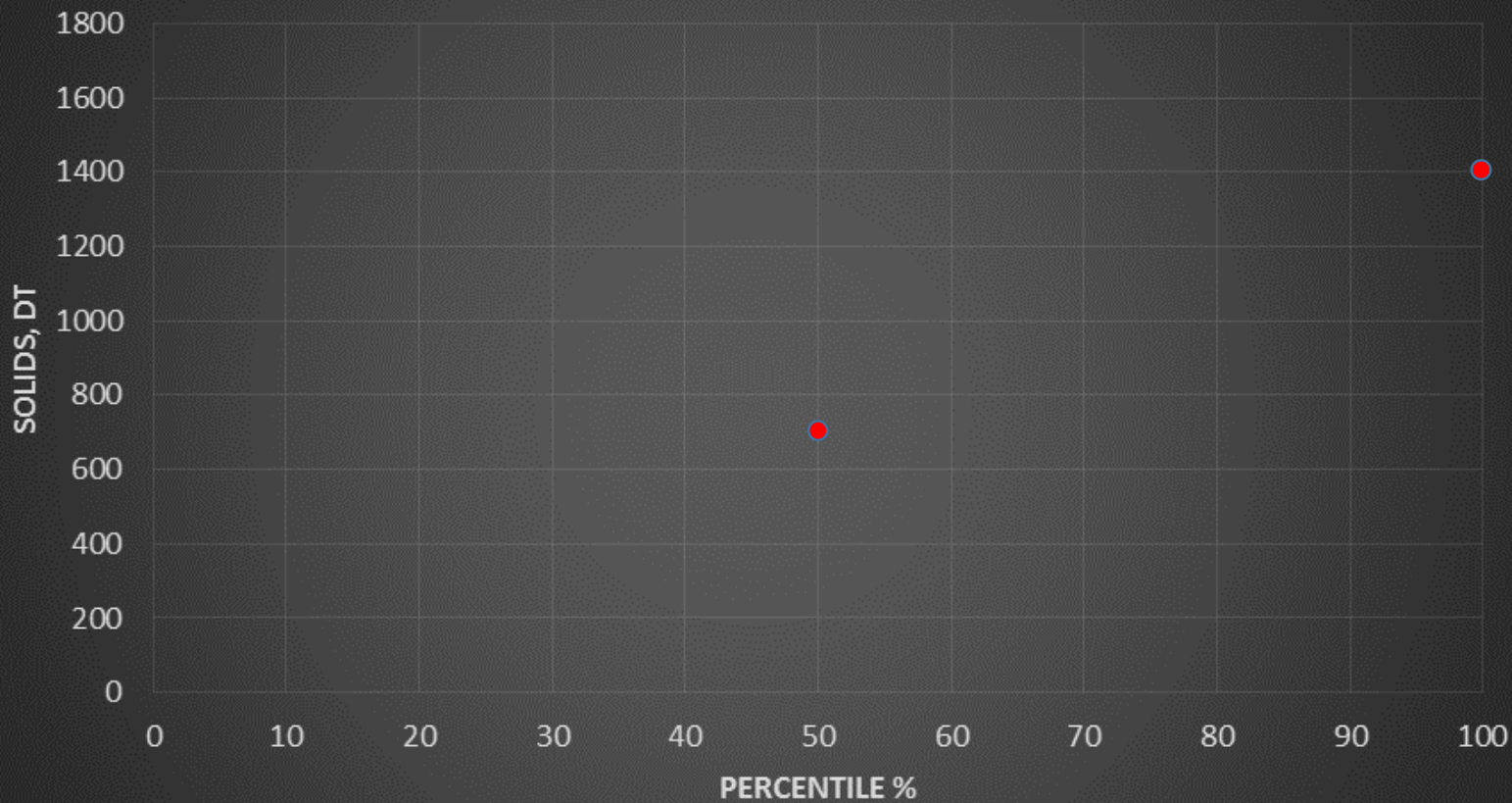


What About the Output?

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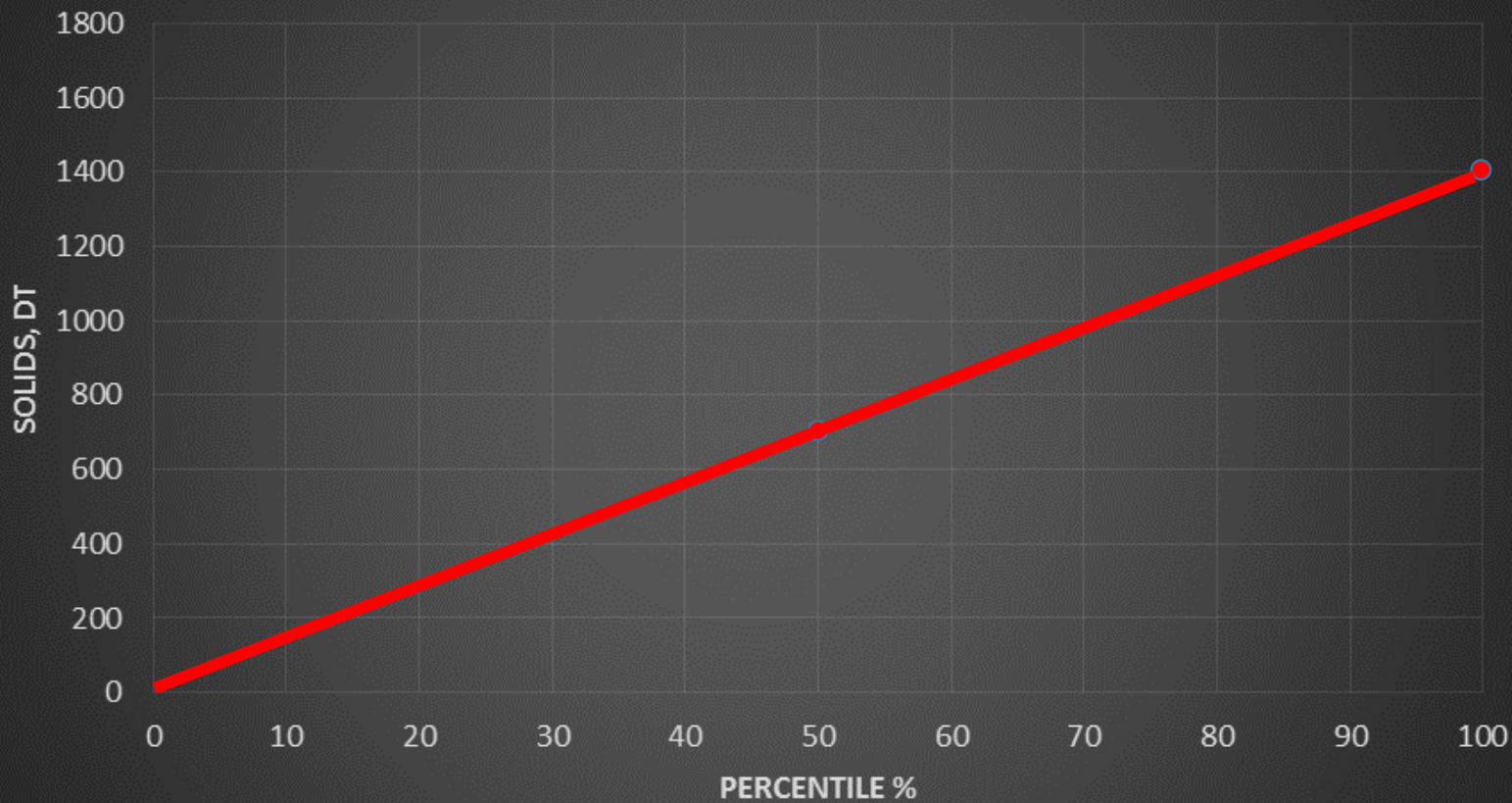
w/o MC: SWWTP Class B Solids to Compost



Baseline

Alternative

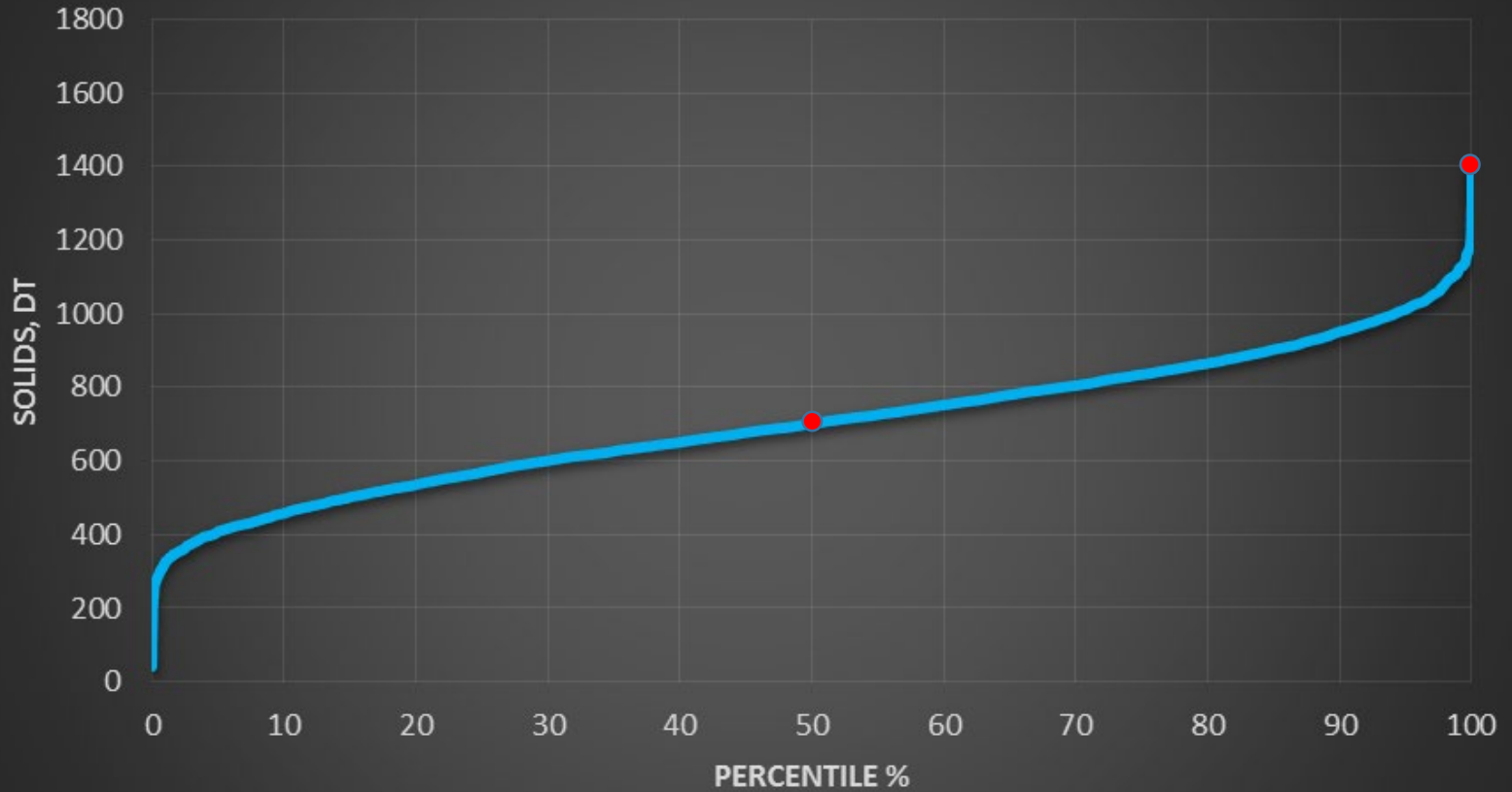
w/o MC: SWWTP Class B Solids to Compost



Baseline

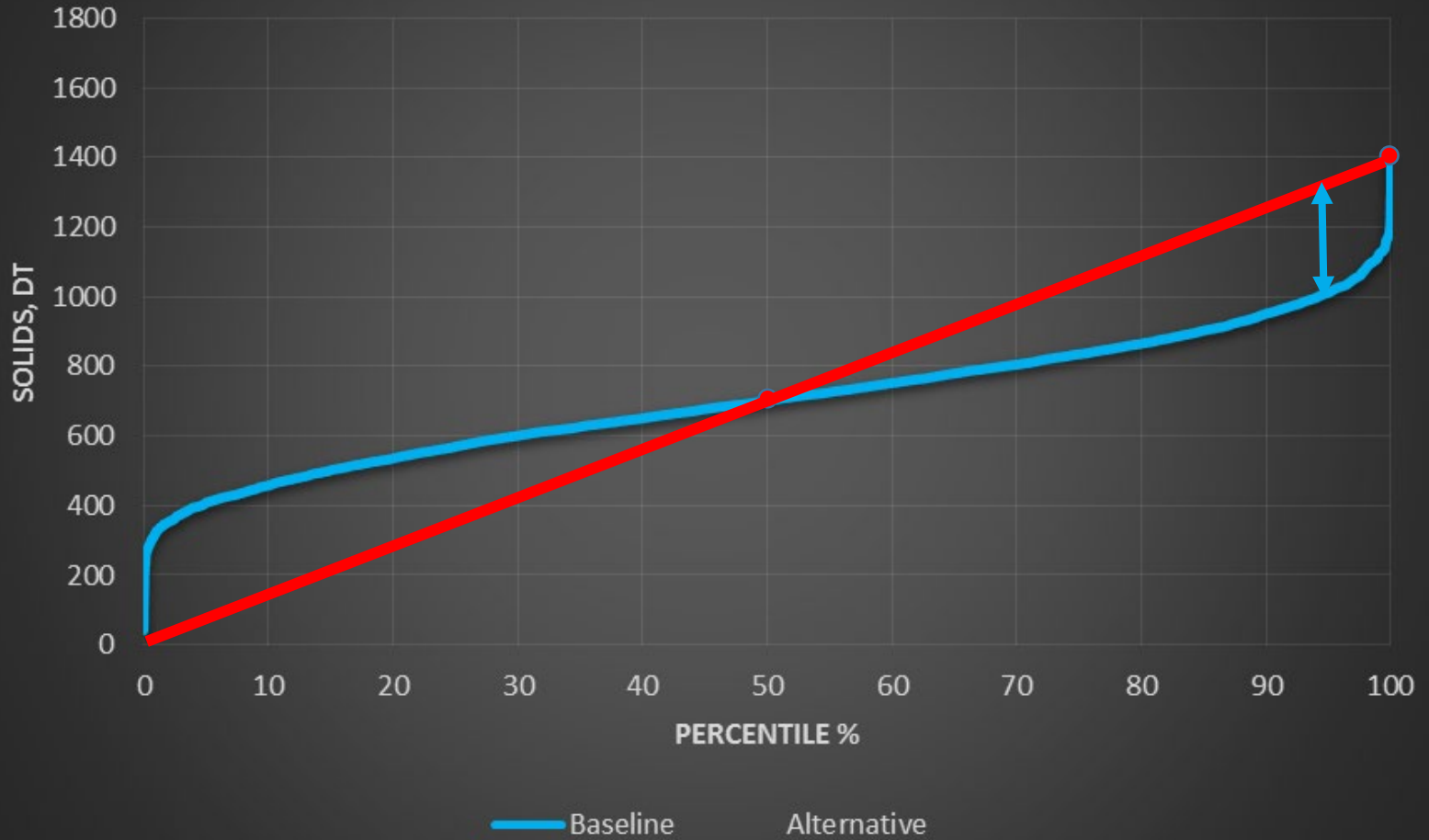
Alternative

SWWTP Class B Solids to Compost

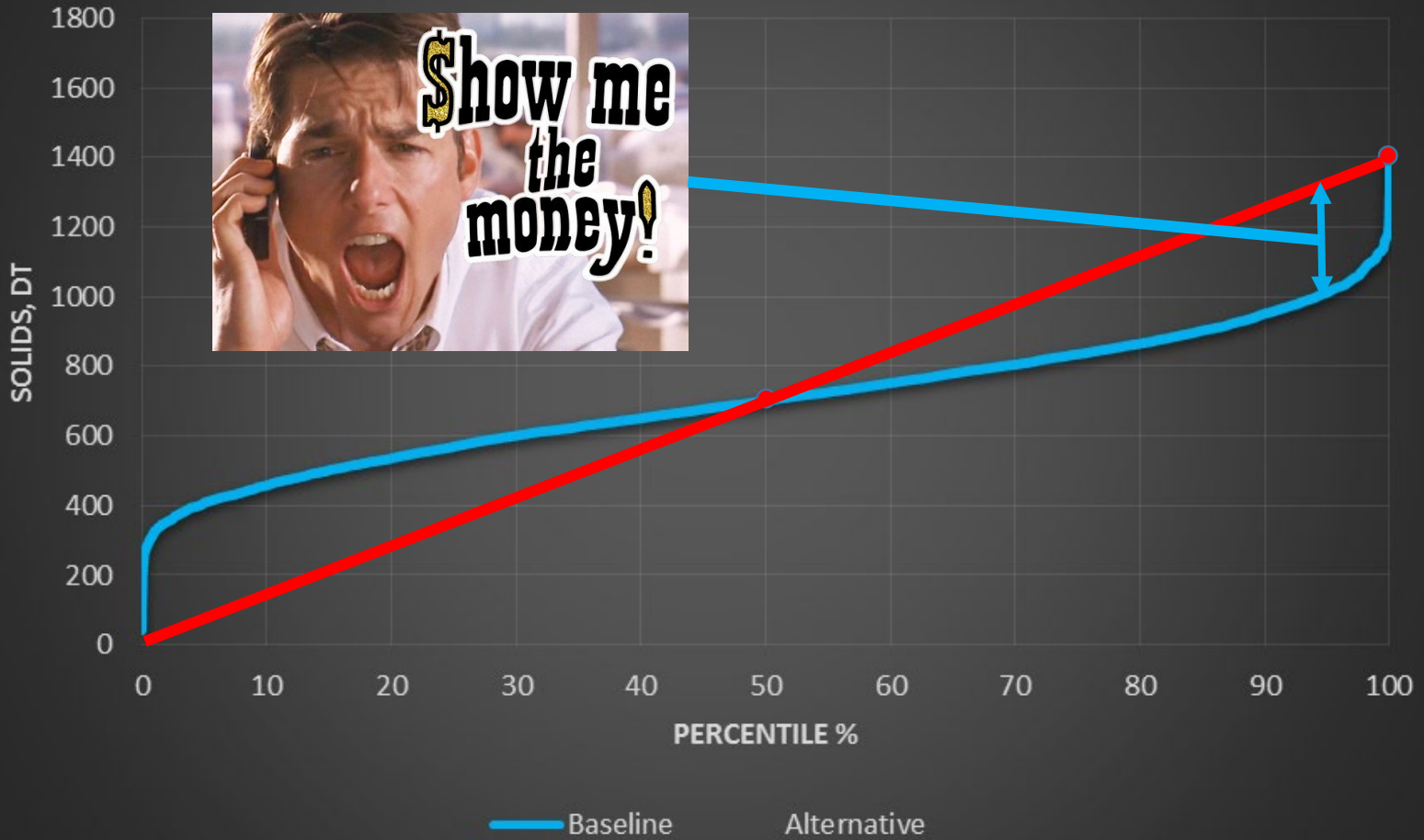


— Baseline Alternative

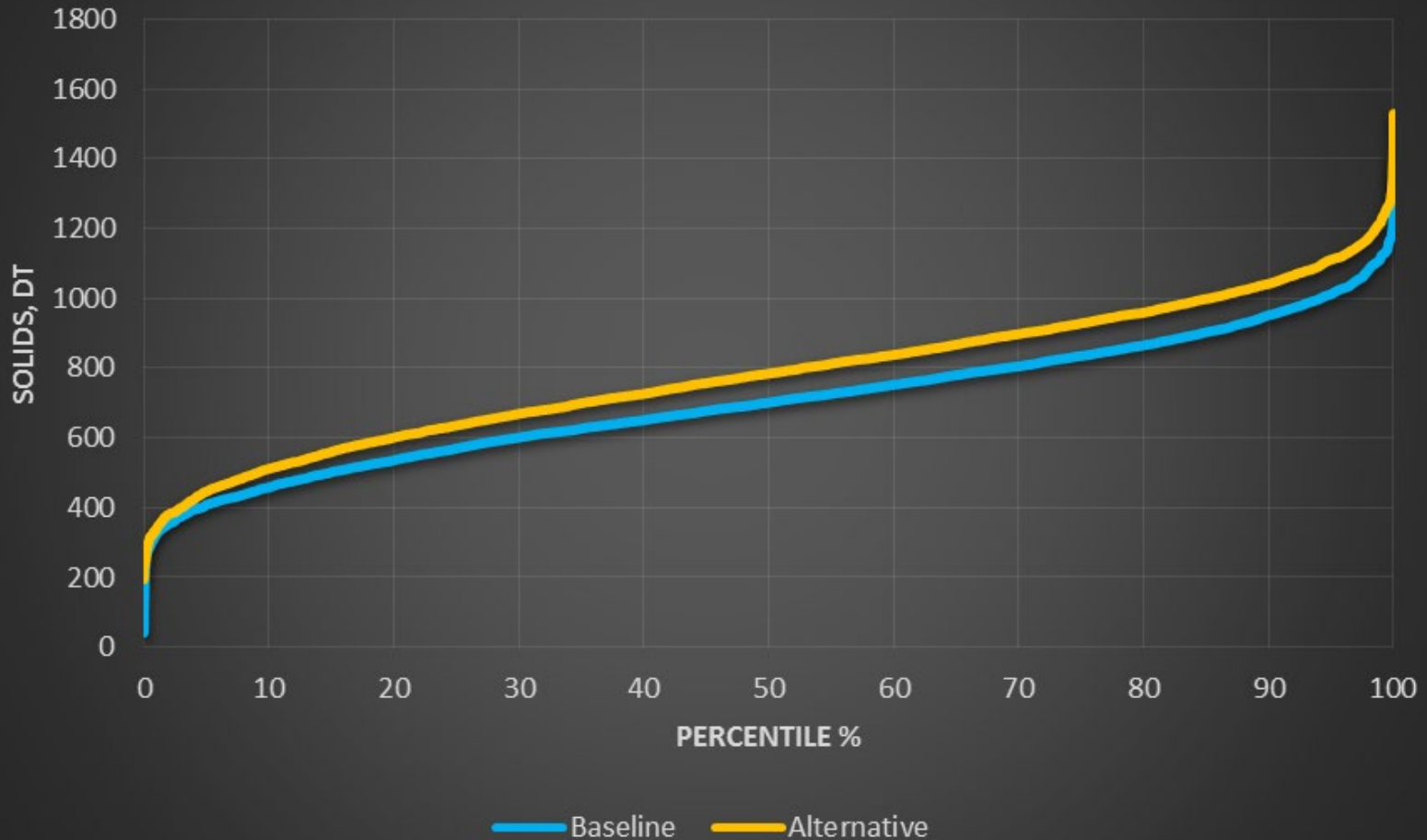
SWWTP Class B Solids to Compost



SWWTP Class B Solids to Compost



With MC: SWWTP Class B Solids to Compost



How Does Monte Carlo Work?



Rolling Two Dice

Known Dice Role Probabilities		
Dice Roll	# of Ways to Roll	Probability
2	1	0.028
3	2	0.056
4	3	0.083
5	4	0.111
6	5	0.139
7	6	0.167
8	5	0.139
9	4	0.111
10	3	0.083
11	2	0.056
12	1	0.028

What if you didn't know?

Die Roll	Probability
1	0.167
2	0.167
3	0.167
4	0.167
5	0.167
6	0.167

Iteration	Die 1	Die 2	Sum
1			0
2			0
3			0
4			0
5			0
6			0
7			0
8			0
9			0
10			0



Dice Roll	Frequency	Probability	Deviation from Known
2			
3			
4			
5			
6			
7			
8			
9			
10			
11			
12			

F	G	H	I	J	K	L	M	N	O	P	Q
Die Role Probabilities			Monte Carlo - 1000 Simulations					Monte Carlo Results			
Die Roll	Probability		Iteration	Die 1	Die 2	Sum		Dice Roll	Frequency	Probability	Deviation from Known
1	0.167		1	3	6	9		2	31	0.031	0.32%
2	0.167		2	4	2	6		3	58	0.058	0.24%
3	0.167		3	5	1	6		4	68	0.068	-1.53%
4	0.167		4	2	1	3		5	123	0.123	1.19%
5	0.167		5	6	1	7		6	149	0.149	1.01%
6	0.167		6	5	4	9		7	152	0.152	-1.47%
			7	2	5	7		8	139	0.139	0.01%
			8	1	3	4		9	112	0.112	0.09%
Probability Bins			9	6	5	11		10	66	0.066	-1.73%
0	1		10	2	2	4		11	66	0.066	1.04%
0.1666667	2		11	6	4	10		12	36	0.036	0.82%
0.3333333	3		12	4	2	6					
0.5	4		13	1	6	7					
0.6666667	5		14	6	3	9					
0.8333333	6		15	5	4	9					
1			16	6	1	7					
			17	4	2	6					
0.39027	random number1		18	1	6	7					
0.9489769	random number2		19	1	3	4					
			20	1	2	3					
			21	1	1	2					
			22	3	4	7					
			23	1	1	2					
			24	1	3	4					
			25	1	4	5					
			26	1	5	6					
			27	3	2	5					
			28	6	2	8					
			29	4	5	9					
			30	6	5	11					
			31	6	1	7					
			32	5	5	10					
			33	5	1	6					

Monte Carlo Analysis Elements and Structure

Models

Daily

- Production of solids (2 plants, 2 types), based on empirical data
- Solids outlet capacity based on empirical data
- Distribution of solids to six outlets
- BLAF storage utilization at both plants

Monthly

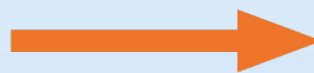
- Digester Reliability/ Catastrophic Failures

Annually

- Costs

Allows for **User-Defined** scenarios

Simulates a full calendar year



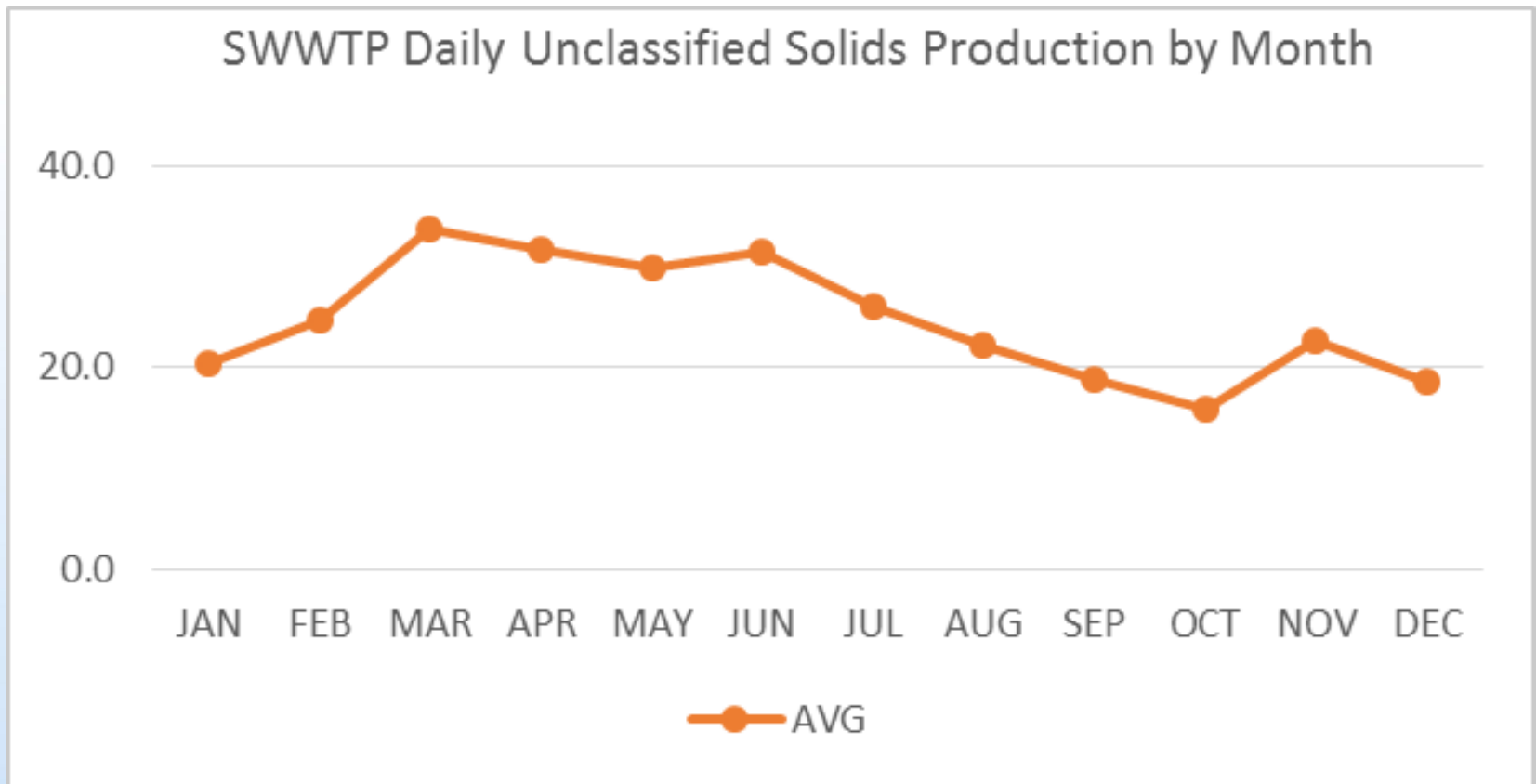
5,000
times

Monte Carlo Analysis for DOSD Solids Disposal

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Solids Production



Daily Solids Production Data 2013-2017

JACKSON PIKE Daily UNCLASSIFIED solids production - historical												JACKSON PIKE Daily CLASS B solids production - historical											
Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
1	2	3	4	5	6	7	8	9	10	11	12	1	2	3	4	5	6	7	8	9	10	11	12
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.57	20.64	14.14	17.22	28.28	28.70	27.68	20.89	17.82	24.53	21.54	9.63
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.54	24.98	15.55	19.33	29.31	27.08	25.76	20.56	19.28	26.72	21.47	9.63
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	13.55	28.45	13.43	19.59	22.70	20.65	17.76	21.04	19.39	30.32	22.37	13.06
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	17.65	25.96	15.40	22.57	17.88	18.98	22.28	20.29	23.87	26.73	21.22	16.63
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	21.67	16.75	6.01	22.58	16.93	23.23	18.30	20.13	27.28	42.76	21.41	15.23
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	19.83	21.60	22.19	22.12	18.32	19.63	19.02	19.85	21.35	34.39	16.22	10.76
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	26.43	35.23	15.39	18.06	18.30	18.40	2.81	14.42	27.42	31.76	14.15	12.31
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	25.44	30.61	11.17	16.79	18.31	23.39	20.02	19.25	27.55	33.06	14.43	10.50
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	25.05	28.15	13.93	16.14	16.17	20.01	34.38	27.71	27.77	31.63	17.44	17.72
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	19.93	28.15	11.12	22.68	18.34	15.90	35.44	29.15	27.61	21.57	19.83	18.42
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	12.82	26.05	12.84	19.58	18.89	16.92	33.94	29.57	6.14	19.92	19.95	20.43
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	6.55	26.07	26.52	11.07	18.34	17.73	28.10	31.50	11.34	14.12	19.13	17.84
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.21	25.02	26.36	13.21	18.85	15.83	23.36	30.46	20.73	2.46	9.24	18.04
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	17.24	25.01	23.28	17.44	18.31	16.00	23.19	32.62	0.00	12.02	0.00	18.04
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.19	24.60	22.26	16.16	21.76	14.84	22.13	17.44	0.00	13.52	0.00	18.89
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.94	26.04	23.26	22.47	28.18	15.53	20.39	13.86	0.00	15.20	0.00	18.03
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.68	26.04	22.25	28.05	26.06	14.97	12.41	12.35	0.00	18.61	4.65	21.96
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.12	24.99	12.32	22.39	26.19	19.47	0.00	0.00	0.00	17.70	16.21	22.27
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.20	25.00	10.10	0.00	20.21	27.48	11.96	0.00	0.00	12.39	14.14	20.54
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	23.00	14.88	12.56	22.54	18.88	27.23	17.95	8.71	0.00	19.56	28.26	19.69
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	23.74	10.10	16.01	19.39	17.92	37.09	19.06	18.28	0.00	17.93	19.10	8.99
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.74	9.65	23.85	17.58	19.62	44.21	19.85	31.66	0.00	16.01	12.61	12.05
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.12	10.12	24.78	38.49	21.25	61.37	14.84	34.56	18.28	16.52	8.81	19.36
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	30.25	10.58	20.36	17.88	19.90	0.00	10.65	19.05	24.00	18.31	14.91	20.12
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	26.26	13.31	17.18	17.37	19.43	0.00	31.76	38.27	19.80	17.61	21.57	13.64
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	20.36	14.88	37.44	14.47	22.04	0.00	28.62	19.98	20.86	18.09	24.17	6.87

Modeling Daily Solids Production using MC

G4 : *fx* =HLOOKUP(\$C4,JPClassBProduction,RANDBETWEEN(2,187))

	A	B	C	D	E	F	G	H	I	J
1										
2						Baseline Solids Production Data (assuming Digesters operational)				
						Overall Solids Production (DT)	JP Class B Solids Production (DT)	SWWTP Class B Solids Production (DT)	JP Unclassified Solids Production (DT)	SWWTP Unclassified Solids Production (DT)
3	Day of the week	Count of Days	Month	Week	Date					
4	Sunday	1	1	1	1/1/2017	73.50	25.31	28.32	0	19.87
5	Monday	2	1	1	1/2/2017	65.13	21.67	30.03	0	13.43
6	Tuesday	3	1	1	1/3/2017	72.09	20.97	29.00	0	22.12
7	Wednesday	4	1	1	1/4/2017	53.59	15.95	24.21	0	13.43
8	Thursday	5	1	1	1/5/2017	70.65	28.90	22.68	0	19.06
9	Friday	6	1	1	1/6/2017	60.89	18.25	24.21	0	18.43
10	Saturday	7	1	1	1/7/2017	79.25	31.56	27.29	0	20.40

Daily Solids Production Lookup Tables

JACKSON PIKE Daily UNCLASSIFIED solids production - historical													JACKSON PIKE Daily CLASS B solids production - historical												
Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec		
1	2	3	4	5	6	7	8	9	10	11	12	1	2	3	4	5	6	7	8	9	10	11	12		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.57	20.64	14.14	17.22	28.28	28.70	27.68	20.89	17.82	24.53	21.54	9.63		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.54	24.98	15.55	19.33	29.31	27.08	25.76	20.56	19.28	26.72	21.47	9.63		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	13.55	28.45	13.43	19.59	22.70	20.65	17.76	21.04	19.39	30.32	22.37	13.06		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	17.65	25.96	15.40	22.57	17.88	18.98	22.28	20.29	23.87	26.73	21.22	16.63		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	21.67	16.75	6.01	22.58	16.93	23.23	18.30	20.13	27.28	42.76	21.41	15.23		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	19.83	21.60	22.19	22.12	18.32	19.63	19.02	19.85	21.35	34.39	16.22	10.76		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	26.43	35.23	15.39	18.06	18.30	18.40	2.81	14.42	27.42	31.76	14.15	12.31		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	25.44	30.61	11.17	16.79	18.31	23.39	20.02	19.25	27.55	33.06	14.43	10.50		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	25.05	28.15	13.93	16.14	16.17	20.01	34.38	27.71	27.77	31.63	17.44	17.72		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	19.93	28.15	11.12	22.68	18.34	15.90	35.44	29.15	27.61	21.57	19.83	18.42		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	12.82	26.05	12.84	19.58	18.89	16.92	33.94	29.57	6.14	19.92	19.95	20.43		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	6.55	26.07	26.52	11.07	18.34	17.73	28.10	31.50	11.34	14.12	19.13	17.84		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.21	25.02	26.36	13.21	18.85	15.83	23.36	30.46	20.73	2.46	9.24	18.04		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	17.24	25.01	23.28	17.44	18.31	16.00	23.19	32.62	0.00	12.02	0.00	18.04		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.19	24.60	22.26	16.16	21.76	14.84	22.13	17.44	0.00	13.52	0.00	18.89		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.94	26.04	23.26	22.47	28.18	15.53	20.39	13.86	0.00	15.20	0.00	18.03		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.68	26.04	22.25	28.05	26.06	14.97	12.41	12.35	0.00	18.61	4.65	21.96		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.12	24.99	12.32	22.39	26.19	19.47	0.00	0.00	0.00	17.70	16.21	22.27		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.20	25.00	10.10	0.00	20.21	27.48	11.96	0.00	0.00	12.39	14.14	20.54		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	23.00	14.88	12.56	22.54	18.88	27.23	17.95	8.71	0.00	19.56	28.26	19.69		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	23.74	10.10	16.01	19.39	17.92	37.09	19.06	18.28	0.00	17.93	19.10	8.99		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	14.74	9.65	23.85	17.58	19.62	44.21	19.85	31.66	0.00	16.01	12.61	12.05		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	18.12	10.12	24.78	38.49	21.25	61.37	14.84	34.56	18.28	16.52	8.81	19.36		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	30.25	10.58	20.36	17.88	19.90	0.00	10.65	19.05	24.00	18.31	14.91	20.12		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	26.26	13.31	17.18	17.37	19.43	0.00	31.76	38.27	19.80	17.61	21.57	13.64		
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	20.36	14.88	37.44	14.47	22.04	0.00	28.62	19.98	20.86	18.09	24.17	6.87		

Digestion

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Why Model Digestion?



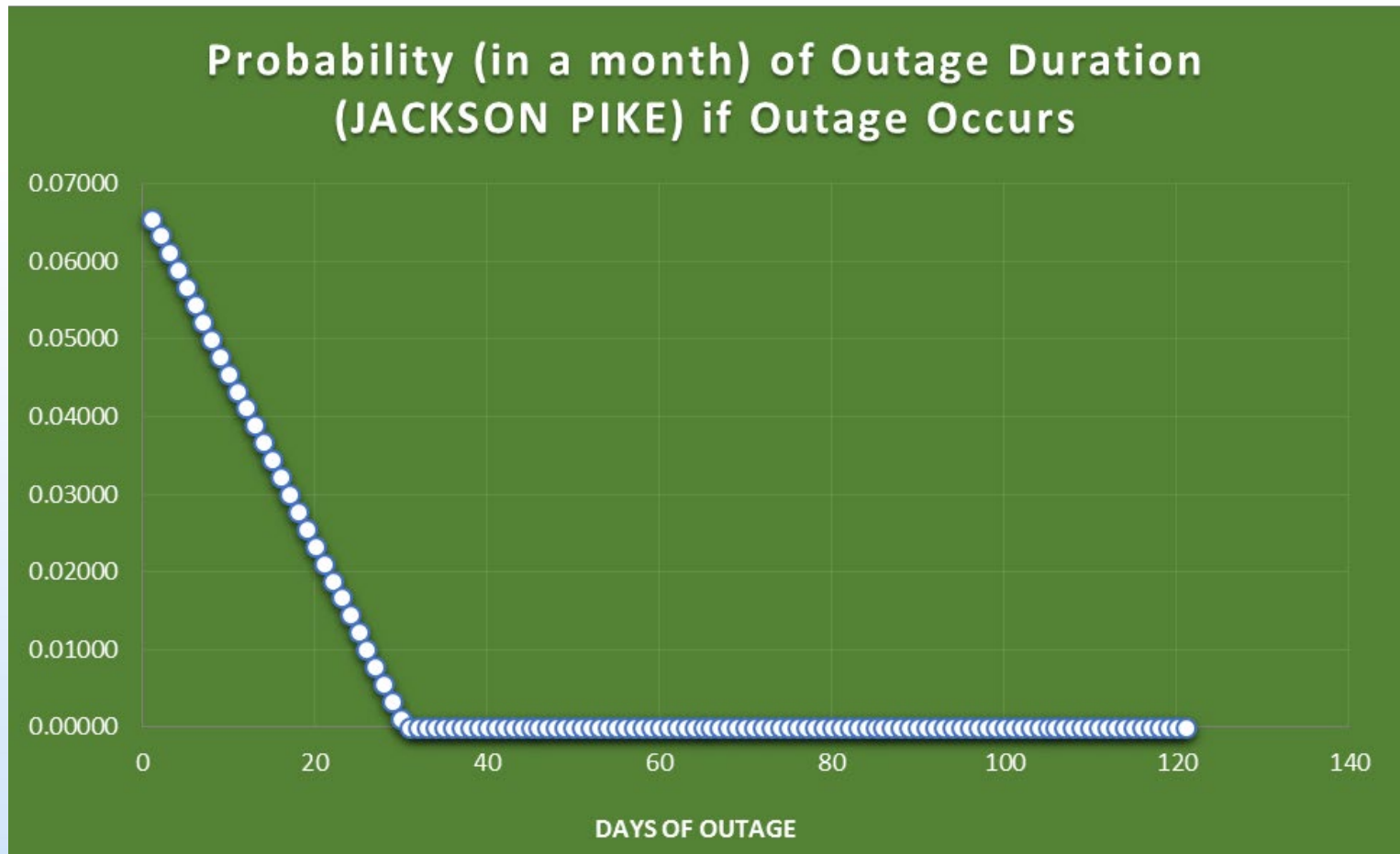
40%

10 DT Unclassified → Digestion → 6 DT Class B

User Inputs

Days to recover from digester failures (max)	30
Solids reduction	40%
Avg. Duration Between Failures	36

Assumptions for Digester Outage Probabilities



Probability Table for Digester Outages

Monthly Probability Bin	x = number of days of outage	y - likelihood of duration (given an outage)	Cumulative Probability of Duration (if outage)
0	0		0.000
0.97222	1	0.06556	0.066
0.97404	2	0.06333	0.129
0.97580	3	0.06111	0.190
0.97750	4	0.05889	0.249
0.97914	5	0.05667	0.306
0.98071	6	0.05444	0.360
0.98222	7	0.05222	0.412
0.98367	8	0.05000	0.462
0.98506	9	0.04778	0.510
0.98639	10	0.04556	0.556
0.98765	11	0.04333	0.599
0.98886	12	0.04111	0.640
0.99000	13	0.03889	0.679

Sample Results for Digester Outages

Failures (days/ month)			
	Month	Jackson Pike	Southerly
January	1	0	0
February	2	0	0
March	3	0	0
April	4	0	0
May	5	0	0
June	6	0	0
July	7	0	0
August	8	0	5
September	9	0	0
October	10	0	0
November	11	0	0
December	12	0	0

Land Application Weather Forecasting

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Forecasting Land App Spreadable Days

RELIABILITY OF LAND APPLICATION MODELED BELOW

Likelihoods of Spreadable Days (Wet year to Dry Year)						
	Probability Bin	Spring Days	Fall Days	Annual Disposal Capacity (DT)	Incremental Probability	
VERY WET	0	12	20	5990	5%	
WET-WET	1	0.05000	16	25	7675	10%
DRY-WET	2	0.15000	24	25	9173	20%
AVG-AVG	3	0.35000	20	30	9360	30%
WET-DRY	4	0.65000	16	35	9547	20%
DRY-DRY	5	0.85000	24	35	11045	10%
VERY DRY	6	0.95000	30	45	14040	5%
	7	1.00000				

ADAPTED from DOSD Evaluation

0.75746556 random number between 0 and 1 (15 decimals)

Spring Days	Fall Days
16	35

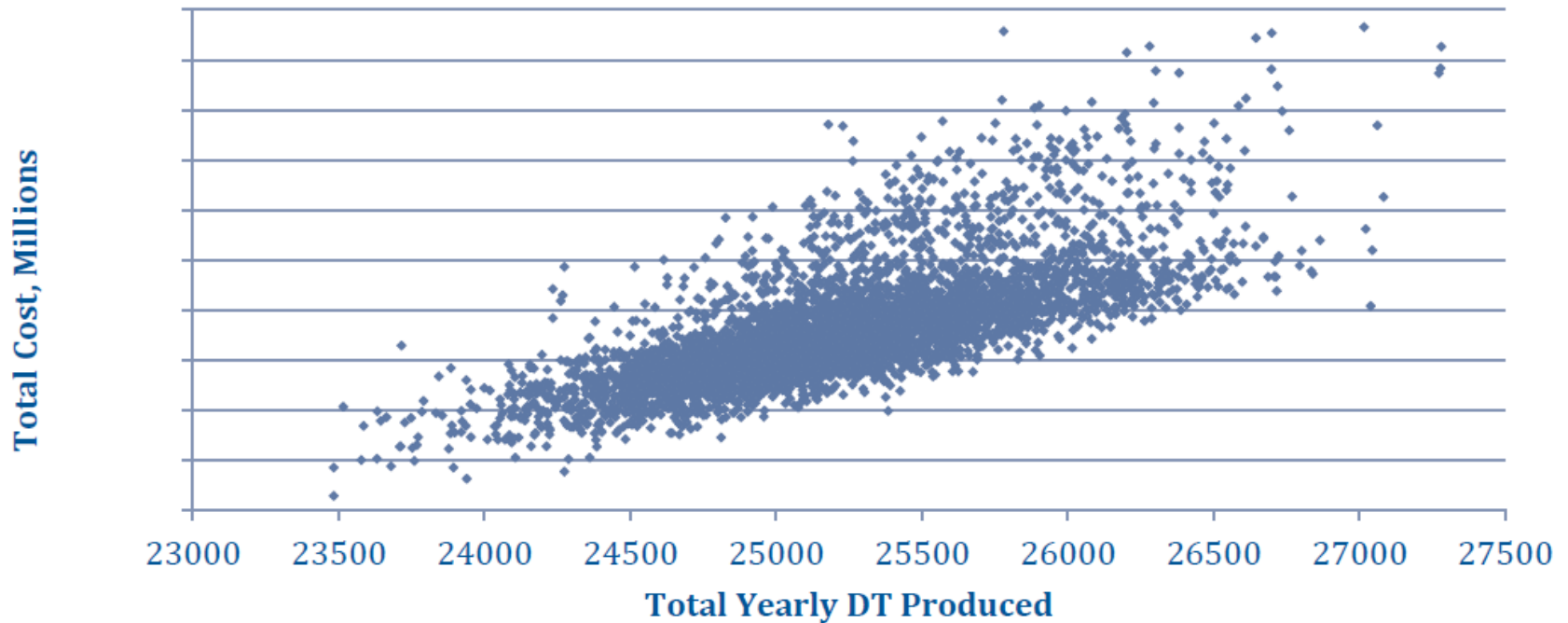
Modeling Costs

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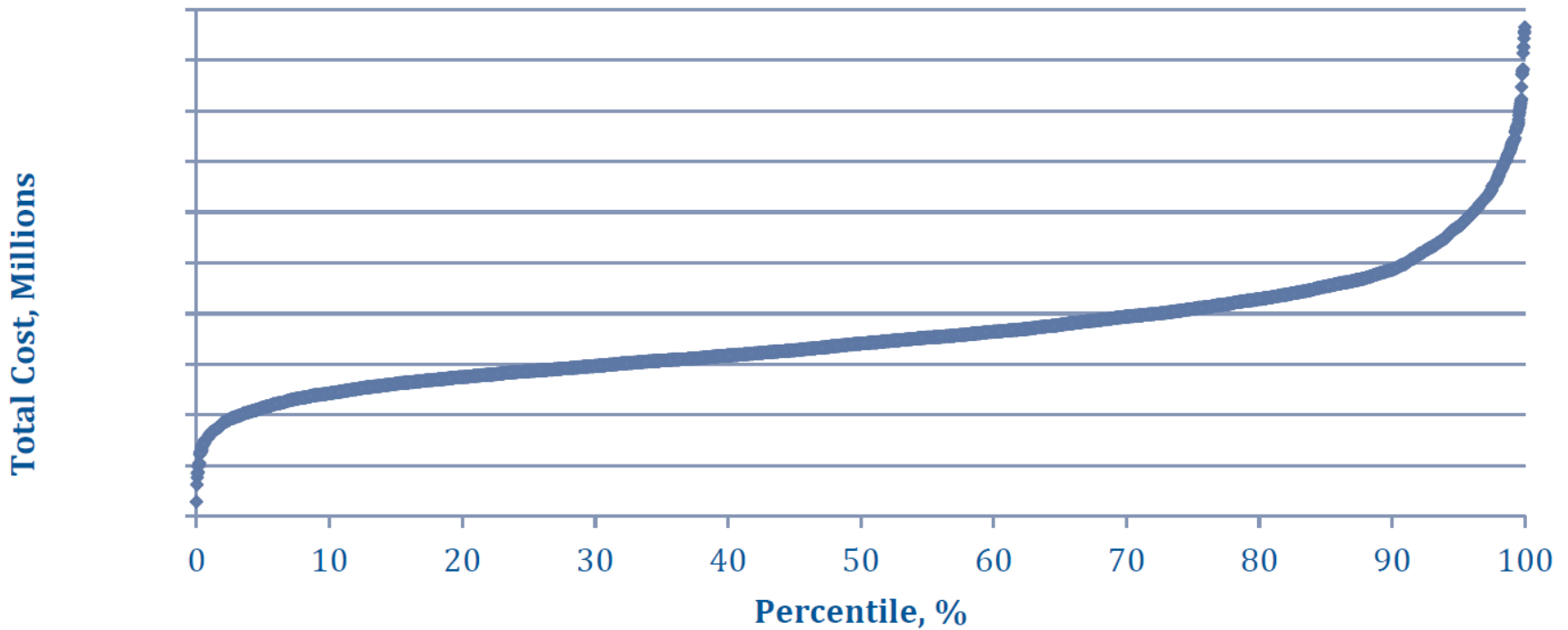
Cost per DT Solids Produced

Total Program Cost to Total Yearly DT Production, Annual



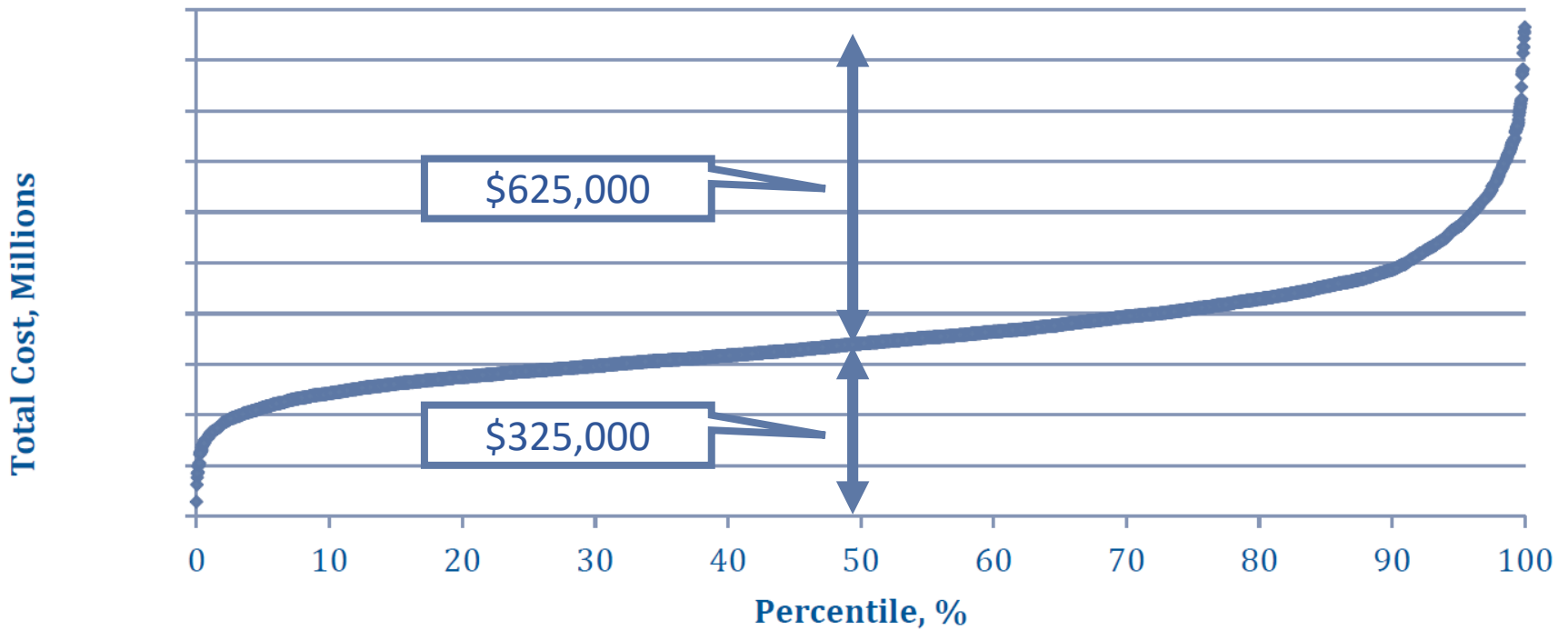
Cost Probability

Total Program Cost, Annual



Cost Probability

Total Program Cost, Annual



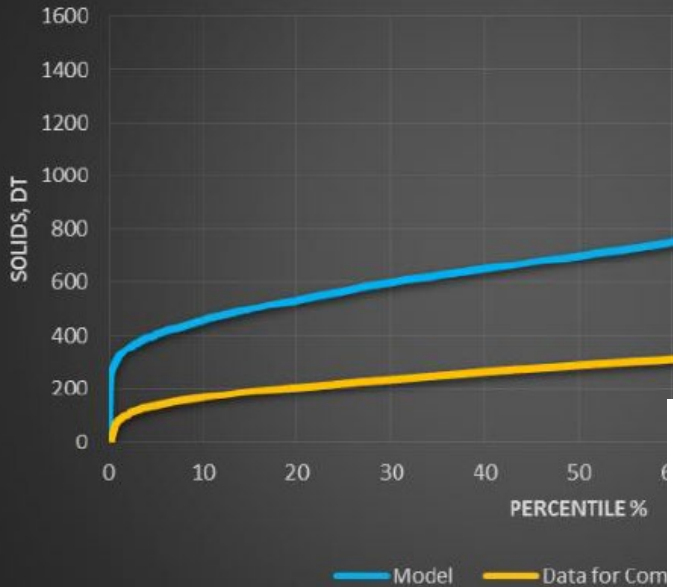
Other Benefits of Monte Carlo

What if.....? Monte Carlo Knows...

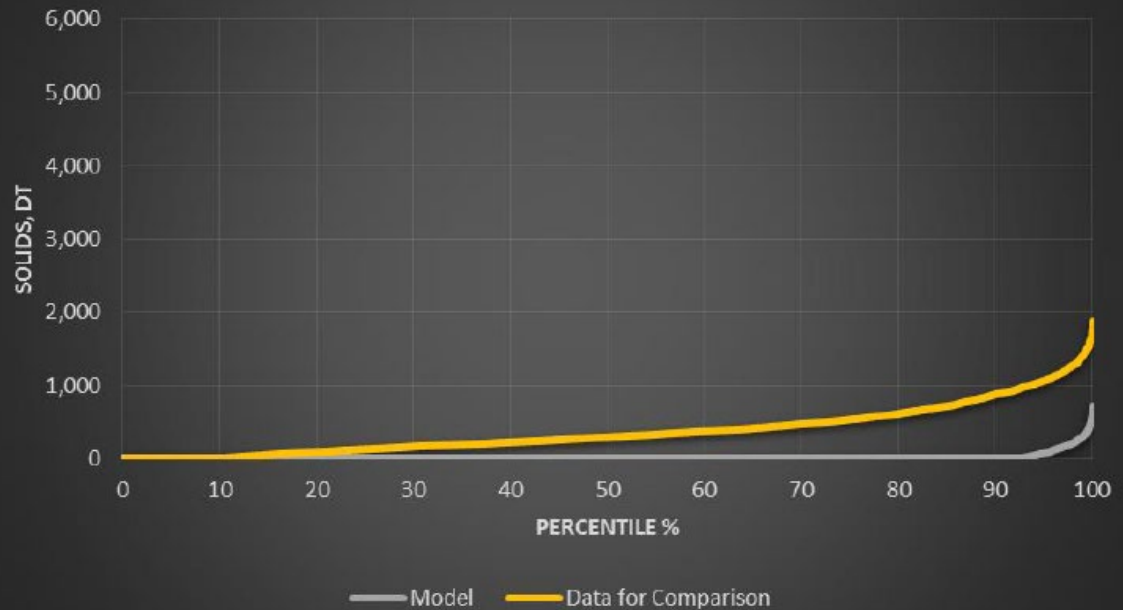
- What if I lose an outlet?
- Should I expand my portfolio to mitigate that risk?
- Should I increase staff at Compost in the summer?
- What if an outlet's capacity increases by 15%?
- What if digester reliability improves by 50%?
- What are my long-term costs if I design storage for the 90th percentile solids production?...80th percentile?

What if DRHP Disappeared?

SWWTP Class B Solids to Compost

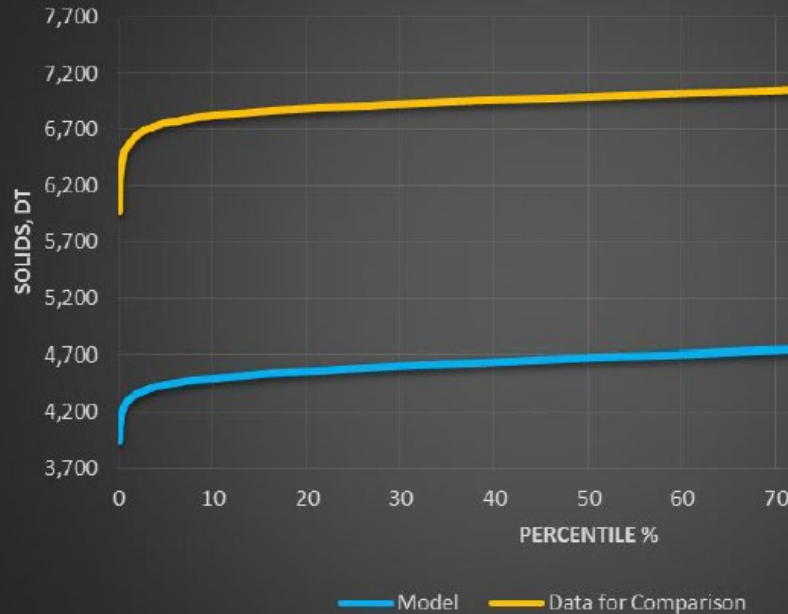


JP Class B Solids to Landfill

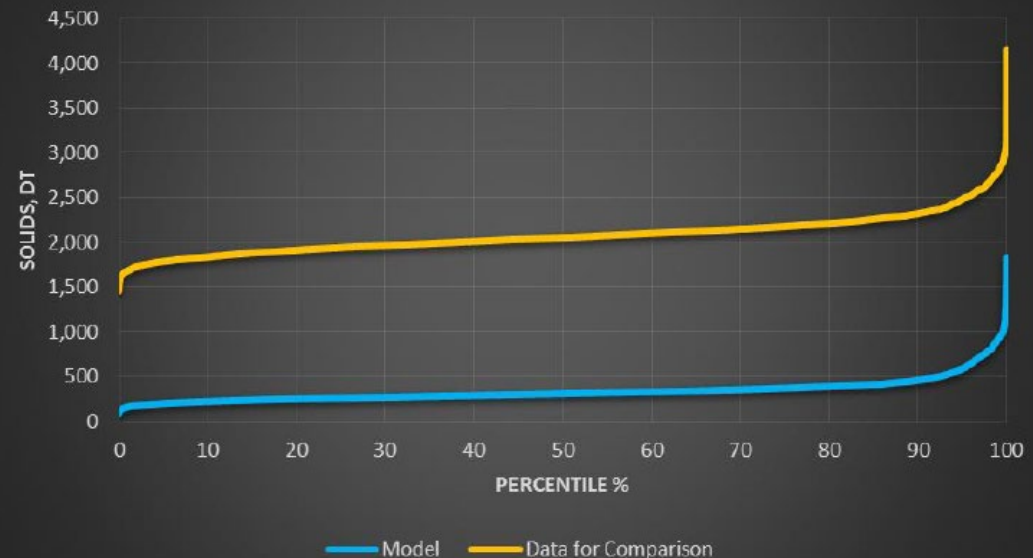


What if Anaerobic Digestion contract is not renewed?

SWWTP Unclassified Solids to Compost

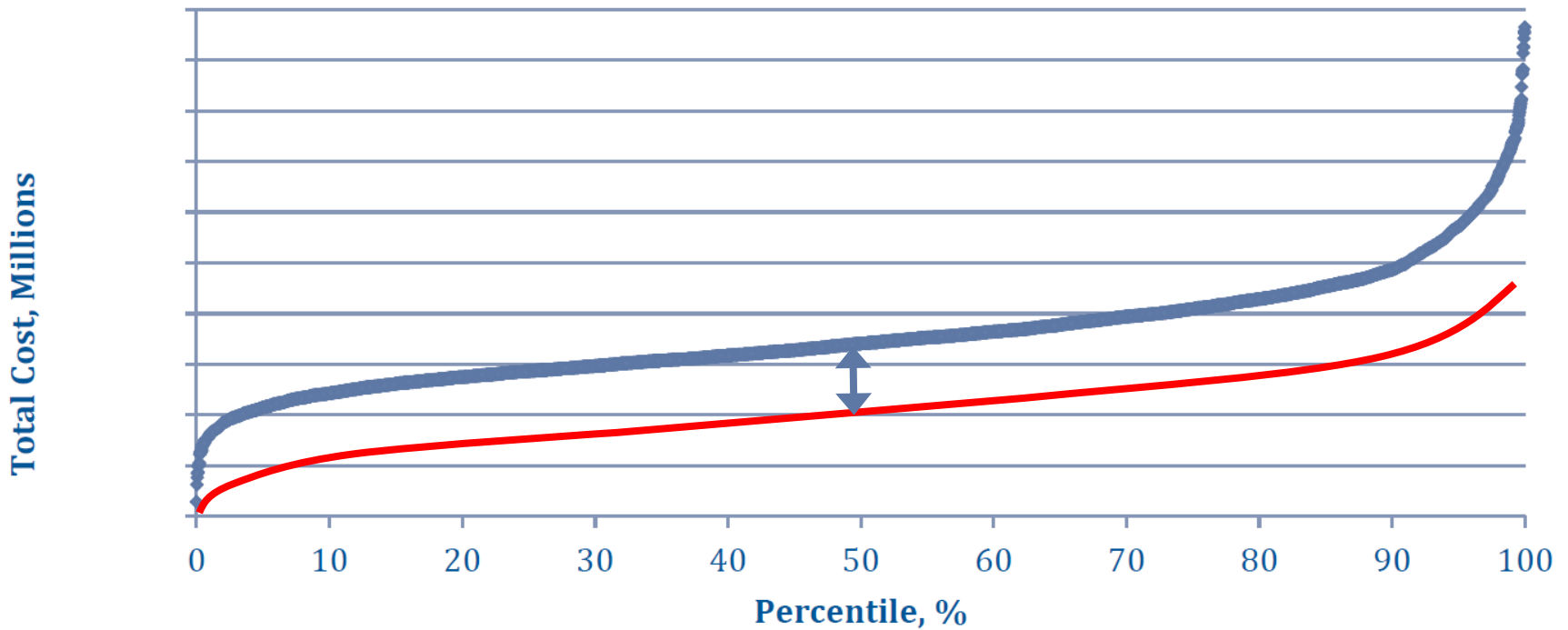


SWWTP Unclassified Solids to Landfill



Cost Probability

Total Program Cost, Annual



Conclusion



- Monte Carlo provide benefits with
 - Model diversity of scenarios, not discrete conditions
 - Understand issues involving many complex variables
 - Answer a wide range of “what if” scenarios quickly
 - Manage risks and optimize costs
- Other applications – optimizing investments
 - Sizing pipelines, treatment systems, storage facilities

Questions?

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