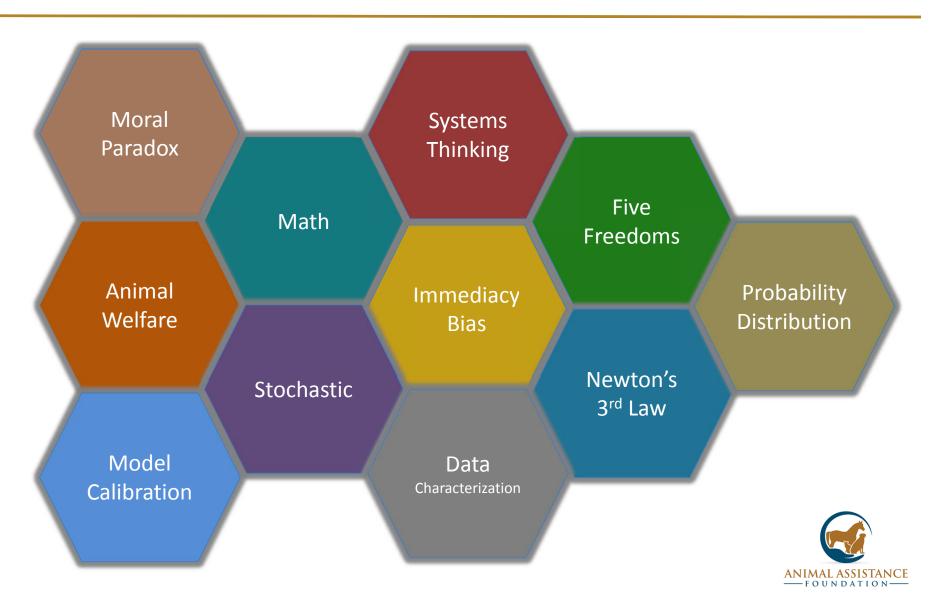
BEYOND LABELS: UNDERSTANDING THE TRUE IMPACTS OF LIVE RELEASE RATE AND INTAKE POLICIES

Roger Haston, Ph.D.

Animal Assistance Foundation



Connecting Concepts



What is Animal Welfare?

The Five Freedoms framework make up a concise and robust definition for animal welfare.

The Five Freedoms:

- Freedom from hunger or thirst
- Freedom from discomfort
- Freedom from pain, injury or disease
- Freedom to express (most) normal behavior
- Freedom from fear and distress

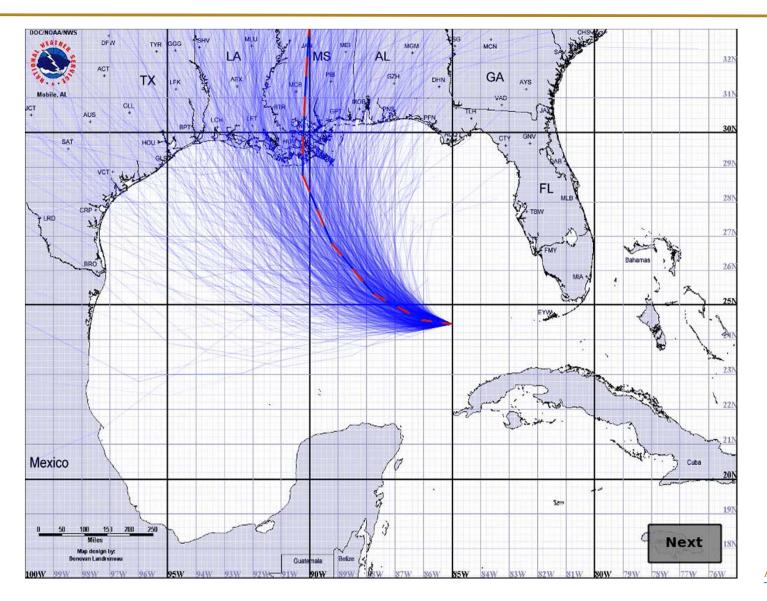


The Power of

STOCHASTIC MODELING & SIMULATION



Stochastic Modeling for Katrina

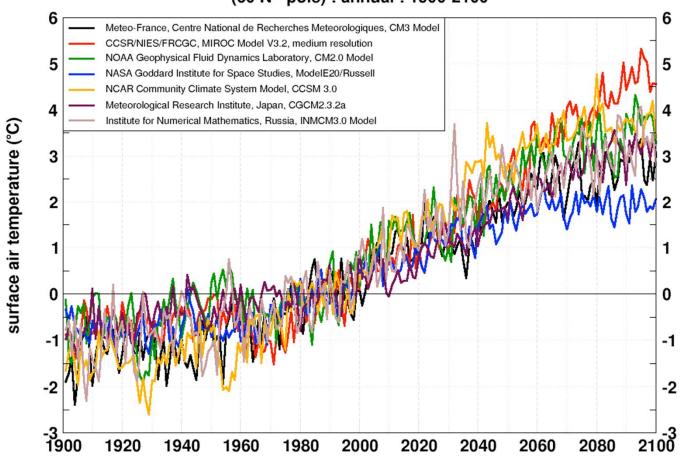




Global Warming

IPCC Projected Arctic Surface Air Temperature

(60°N - pole): annual: 1900-2100





Basics of Stochastic Modeling

- A technique for analyzing data and/or predicting outcomes for a set of variables and conditions that contain a degree of randomness or unpredictability (number of intakes on a given day, etc.)
- The model allows us to assess the interactions between all of the variables.
- In order to capture the unpredictable nature of shelter operations, a stochastic model was developed that incorporates the variability of many of the factors within a shelter (e.g. type of animal, daily intake, length of stay, cost).
- Each variable in the model were calibrated to data from actual shelter operations and/or validated with shelter operations personnel.

Problem Solving Theory

Step 1: Problem Identification

What is the discrepancy between what is expected and what is occurring

Step 5: Plan Evaluation

Is the plan effective?

Step 2: Problem Analysis

Why is the problem occurring?

Step 4: Plan Implementation

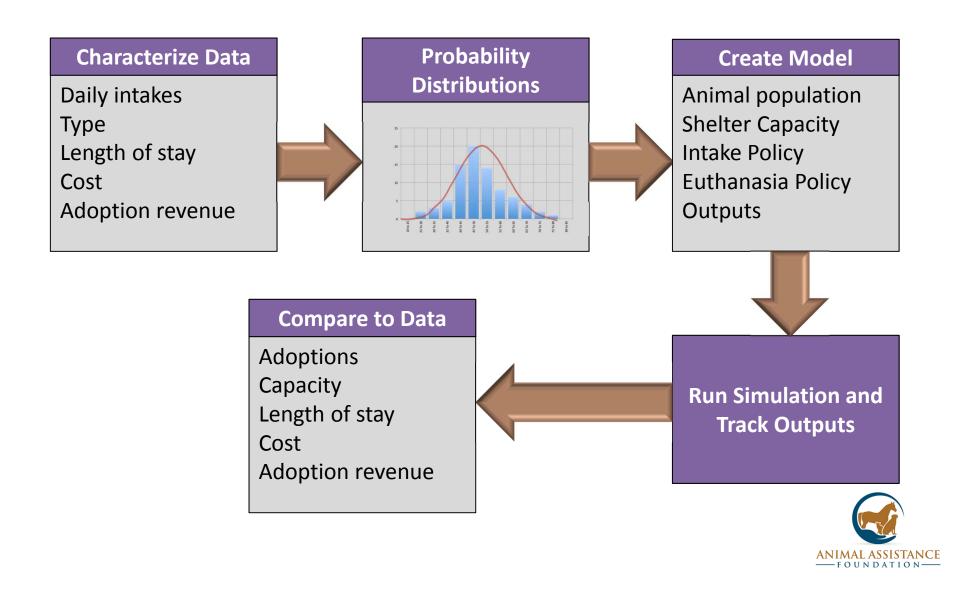
How will implementation integrity be ensured

Step 3: Plan Development

What is the goal? How will progress be measured



How the Process Works

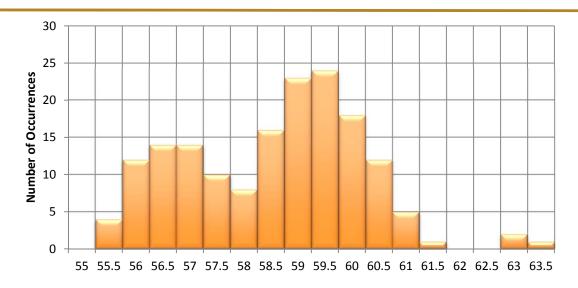


Seeing Through the Noise

DATA CHARACTERIZATION



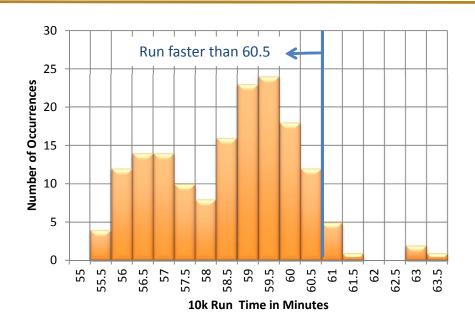
The Problem with Outcome Data



- Outcome data is the result of a system. For example adoptions.
 - Adoptions are the result of intakes, kennel space, advertising, etc.
- Understanding causality is difficult because outcome data are often the cumulative effect of multiple variables
- Variance in the outcome data is hard to interpret, however asking the right questions as to why the data has variance is critical



A Simple Example: Roger's Run Time



I am about to head out on my morning run, I bet you \$100 that I can run the 10k in 60.5 minutes or less

Should you take the bet?

Hint: 94% of the time in the past I have run faster than 60.5 minutes

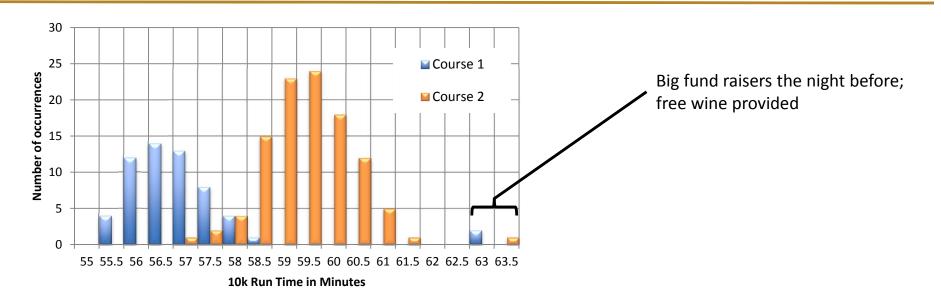
- We see that the data is spread over a large set of times with a few outliers.
- Key question: Why is there so much variability in the data?
- We need to understand what causes variability in the data

You find out that:

- I run on two different courses
- My performance is negatively effected by alcohol



Properly Characterizing Data



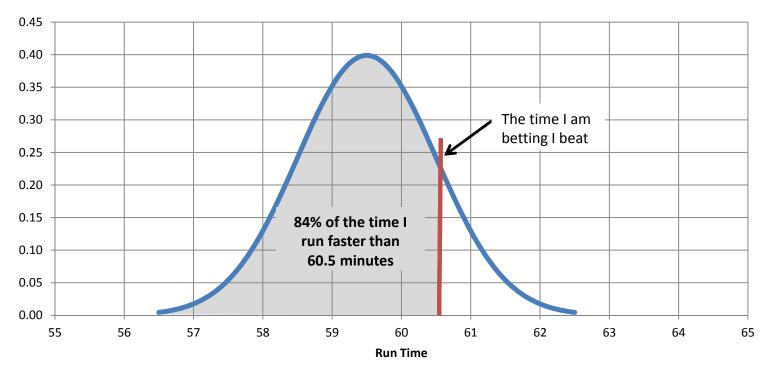
By splitting the data into a more relevant grouping, we can now see that you can more likely predict my run time if you know what course I am on.

- If I'm running course 1, I'm likely to be around 56 minutes
- if I am running course 2, then I'm likely to be around 59.5 minutes
- If I went to a big fundraiser I'll be very slow

You find out that I plan to run on Course 2.



Probability Distributions from Data



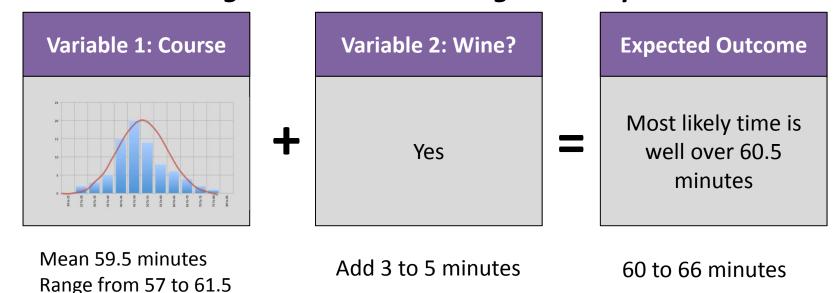
By looking at the data, we see it approximates a normal distribution. Some basic statistics allows us to build a curve that shows the probability of Roger making a particular time on Course 2. There is an 84% chance that my time will be faster than 60.5 minutes.

Should you take the bet?



Building the Full Model

- You find out I am going to a big fund raiser the night before that has Dave Mathews playing. From the data, you see this tends to add 3 to 5 minutes to my time
- The Model: Using all the variables together to predict



Should you take the bet?

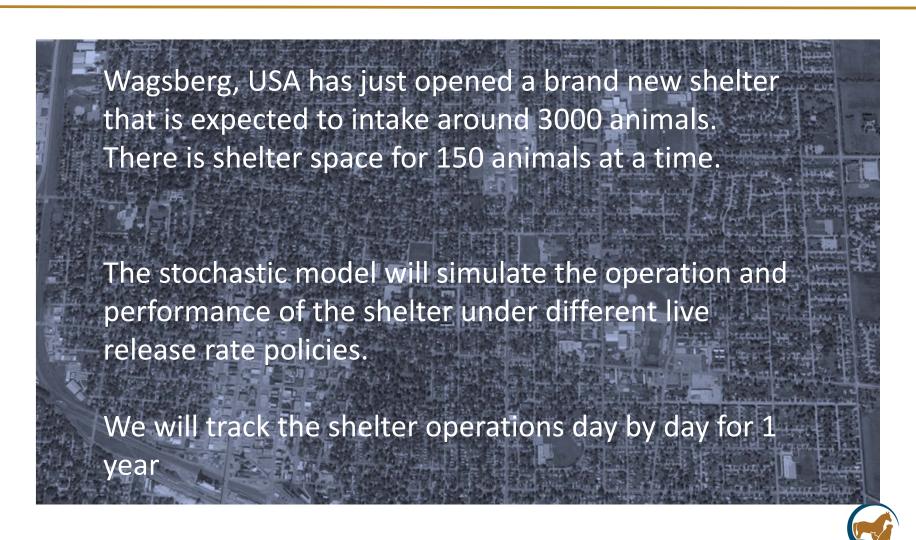


Animal Shelter Simulation

WELCOME TO WAGSBERG, USA



Wagsberg, USA



Definitions and Assumptions

- Live Release Rate:
 - number of adoption divided by the number of intakes
- Outcomes
 - There are only two possible outcomes
 - Adoption
 - Euthanasia
- All animals coming to the shelter are ownersurrender
- Unserviced Animals
 - Animals that were brought to the shelter but could not get in due to space



Key Input Parameters for Simulation

Anin	nal Population			
	Type (percent of population)	Calibrated to shelter data		
	Expected length of stay	Calibrated to shelter data		
	Expected cost for additional medical and behavior costs	Estimated		
	Expected Adoption Revenue	Calibrated to shelter data		
Shelter				
	Intake policy	Based on desired live release rate		
	Kennel capacity	No doubling up		
	Base cost per day	Calibrated to shelter information		
	Euthanasia policy	Based on policy setting		



Animal Population Characteristics

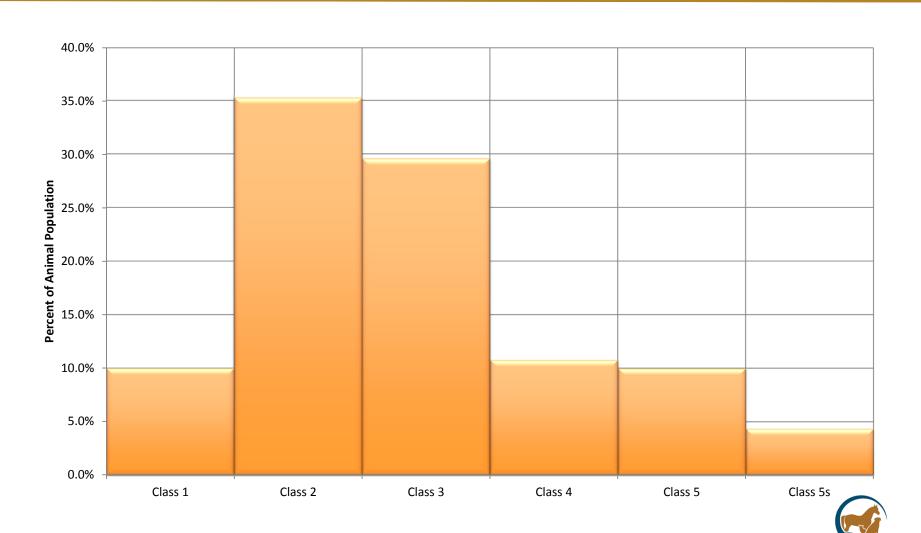
Animal Type	Description	Asilomar Class	% of Pop.	Expected Sta	Length of Y*	Base Cost/Day	Behavior & N Costs per		Adoption Revenue
				Mean	Std		Scale	Degrees Freedo m	
Class 1	Puppies	НН	10%	4	0.4	\$13.00	No Additiona	al Costs	\$350
Class 2	Highly adoptable	НН	35%	7	0.7	\$15.00	No Additiona	al Costs	\$250
Class 3	Minor issues	TR	30%	17	4.25	\$15.00	\$5.00/ day	3	\$100
Class 4	Some behavior and health issues; less desirable breeds	TM	10%	60	30	\$15.00	\$4.00/day	3	\$50
Class 5	Significant issues; adoptable under special circumstances	UU	10%	180	36	\$15.00	\$3.00/day	3	\$0
Class 5s	Unadoptable	UU	5%	Lifetime	n/a	\$15.00	\$3.00/day	3	\$0

^{*}Normal distribution



[^] Chi Square distribution

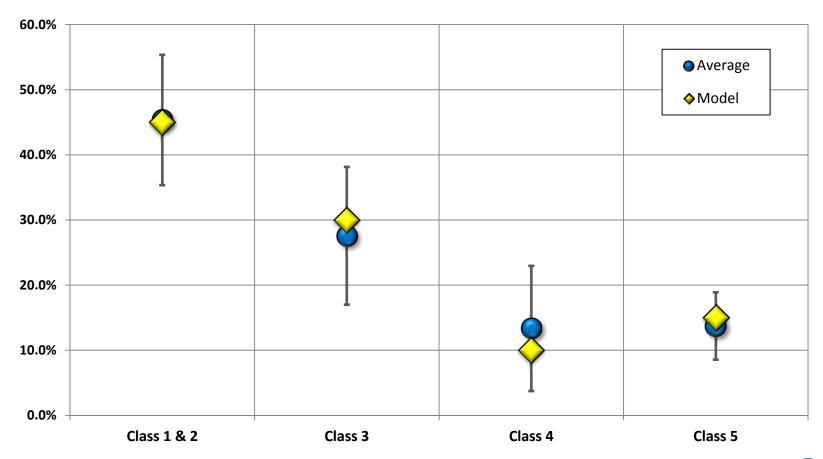
Population of Animals by Class



ANIMAL ASSISTANCE — FOUNDATION—

Population by Animal Class

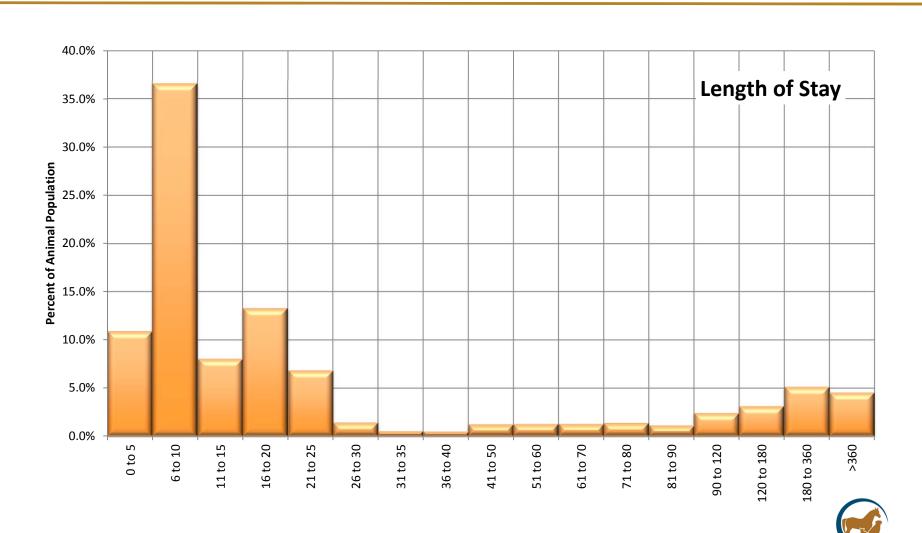
(Calibrated to Actual Shelter Data)



The average is reported data from 10 different shelters from around the country. The gray lines are the standard deviations (e.g. variability)

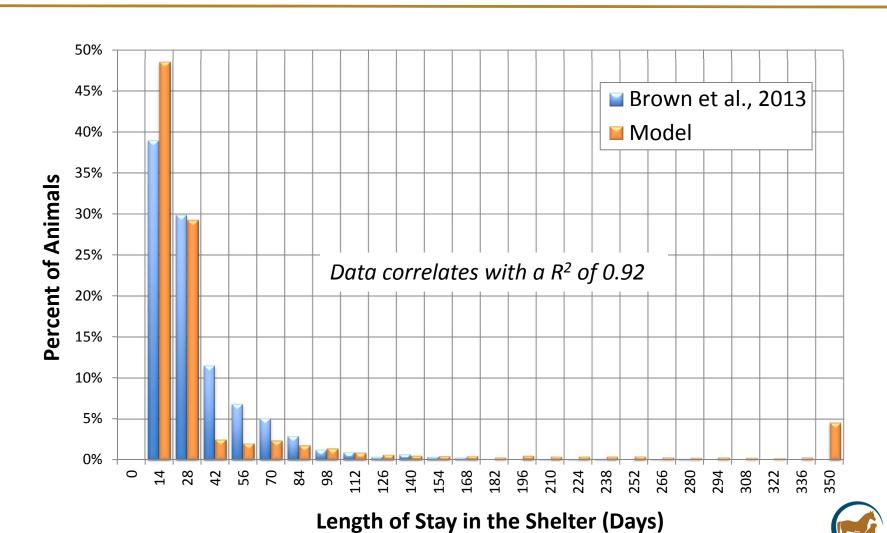


LOS for Entire Population



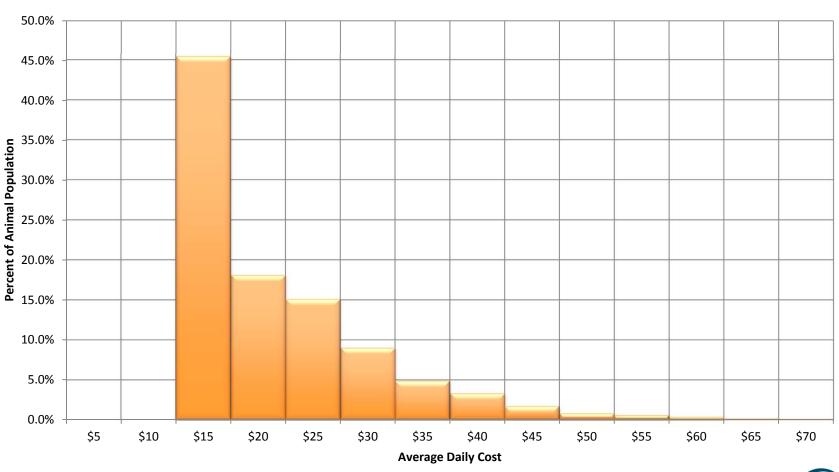
ANIMAL ASSISTANCE

Length of Stay Calibration



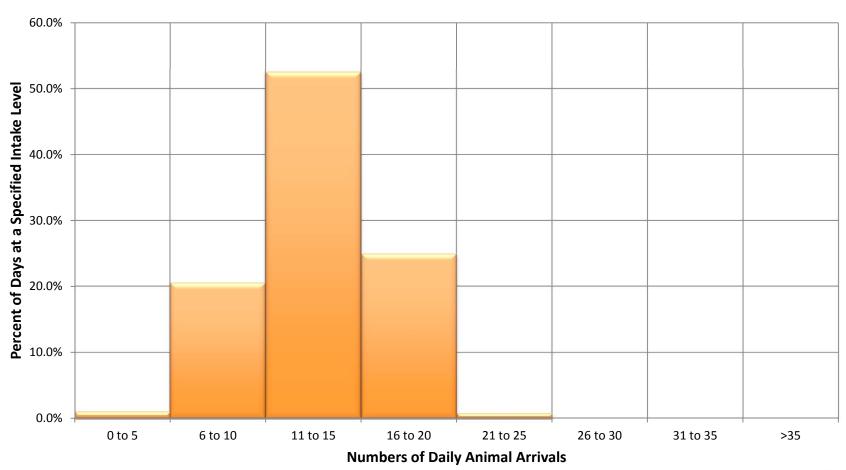
ANIMAL ASSISTANCE

Expected daily cost distribution





Expected Arrivals per Day





Shelter Operational Characteristics

	Kennels	Target LRR	Intake Policy	Euthanasia Policy	Expected Arrivals
Policy 1	150	85%	Accept if Room	Only Class 5	3000
Policy 2	150	90%	Accept if Room	Only Class 5; Class 5s first	3000
Policy 3	150	95%	Accept if Room	Only Class 5s	3000

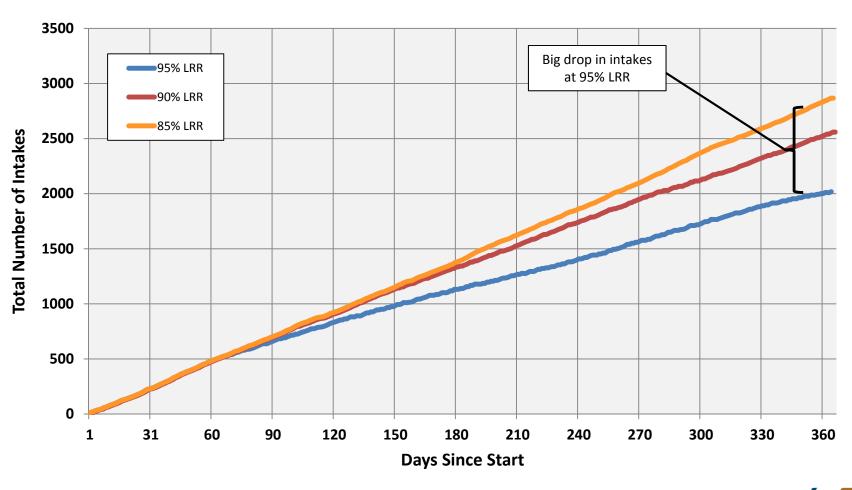


Results

SINGLE SHELTER LIVE RELEASE RATE IMPACT

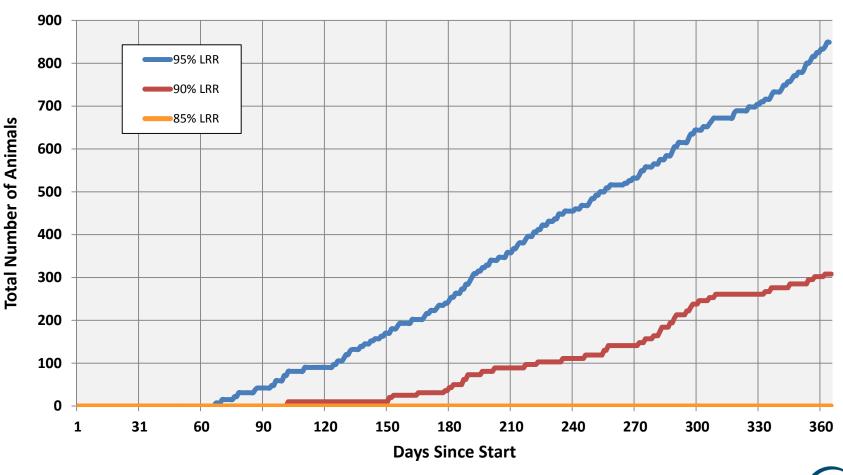


Cumulative Intakes



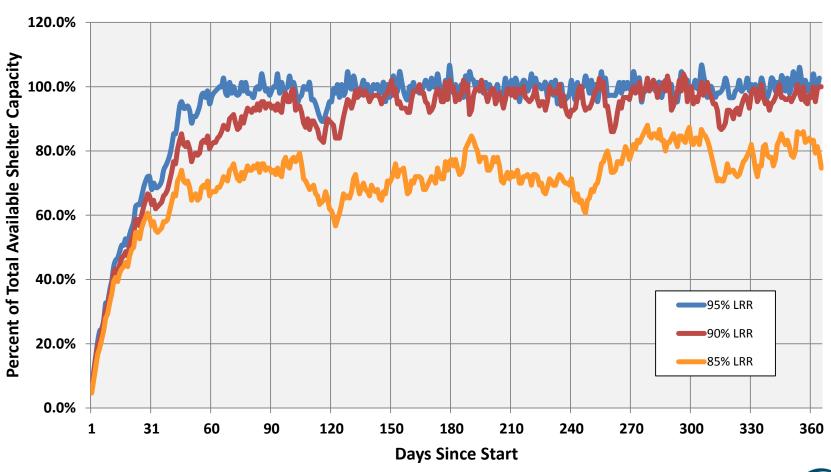


Animals Turned Away



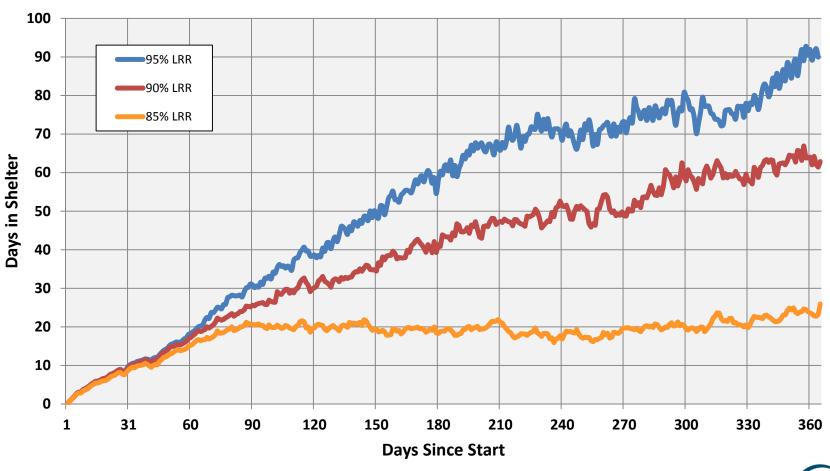


Shelter Capacity



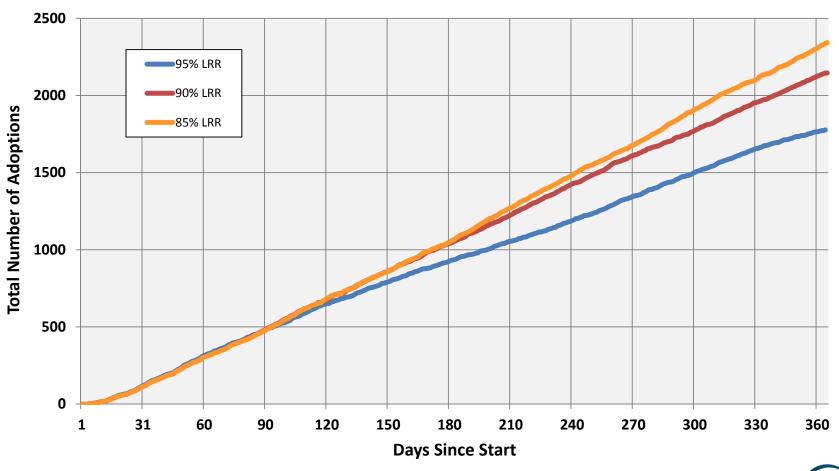


Average Length of Stay



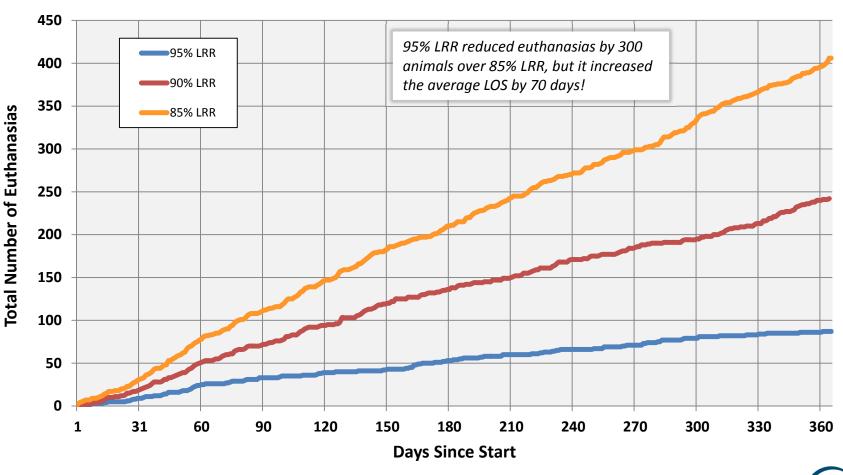


Adoptions



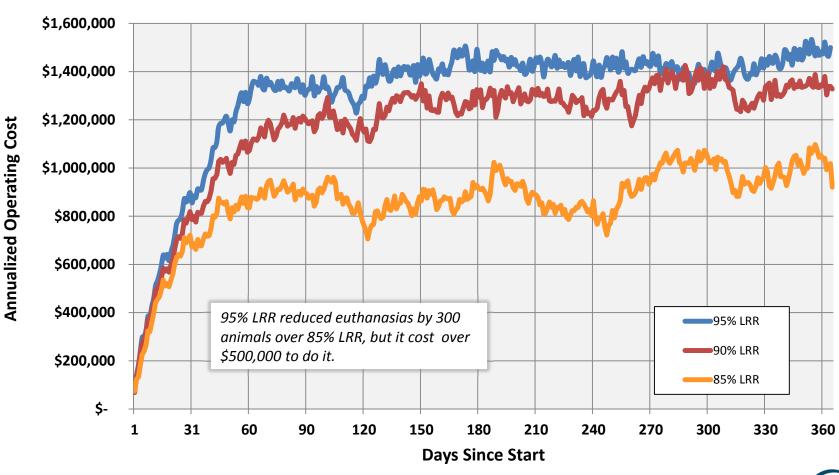


Euthanasias



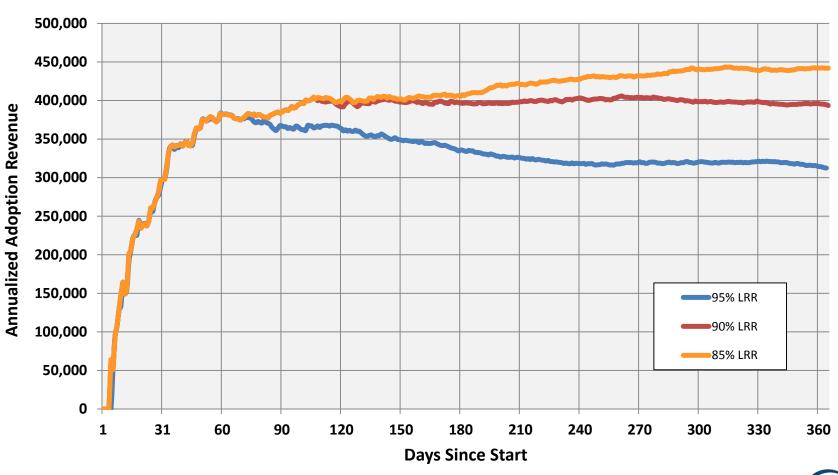


Operating Cost



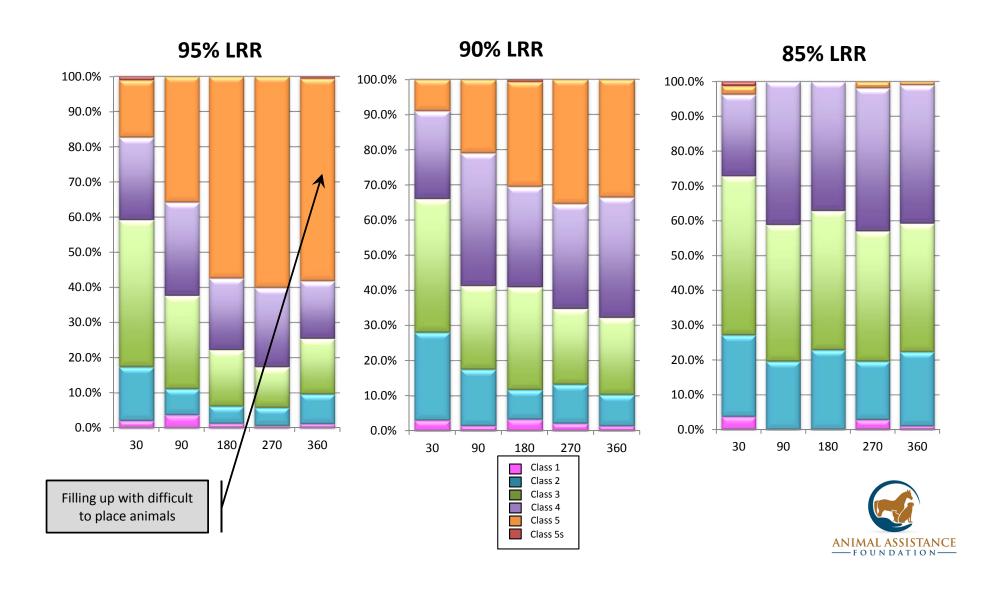


Annualized Adoption Revenue





Animal Inventory over Time



Complex systems require more than one metric to measure success

THE PROBLEM WITH LIVE RELEASE RATE AS A METRIC



Problems with Live Release Rate

- Only looks at one aspect of animal welfare
- A metric that is the outcome of division (one number divided by another) loses all information about scale
 - 9 animals adopted from 10 intakes is 90% LRR
 - 90,000 animals adopted from 110,000 intakes is 81% LRR
- Achieving a high live release rate is not necessarily a measure of a successful operation
 - Extremely limited admission (all puppies)
 - No available space for arrivals
 - Warehousing
- It is highly dependent on the input population

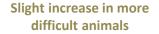


Impact of Input Population on LRR

Two exact shelters operating at a 90% LRR but with two different input populations. One population has more Class 4 and 5 and animals than the other. All outcome metrics are worse except euthanasias; this is only down because many animals are rejected at intake

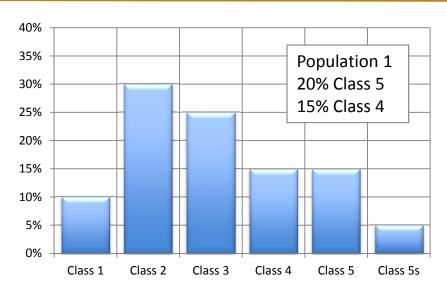
Animal Type	Shelter 1 Population	Shelter 2 Population
Class 1	10%	10%
Class 2	35%	30%
Class 3	30%	25%
Class 4	10%	15%
Class 5	10%	15%
Class 5s	5%	5%

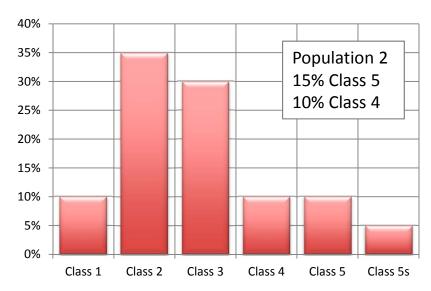
Outcome Metrics	Shelter 1	Shelter 2	Change	Metric Direction
Intakes	2776	2170	22% Down	Worse
Unserviced	30	517	1600% Up	Worse
Adoptions	2346	1769	25% Down	Worse
Average LOS	66	88	34% Up	Worse
Euthanasias	270	198	17.3% Down	Better
% Long Term	41%	58.3%	42% Up	Worse
Capacity	94.3%	98.6%	5% Up	Worse
Operating cost	\$1,240,032	\$1,645,671	33% Up	Worse
Cost per Adoption	\$529	\$930	76% Up	Worse
Adoption Revenue	\$438,752	\$309,548	29% Down	Worse

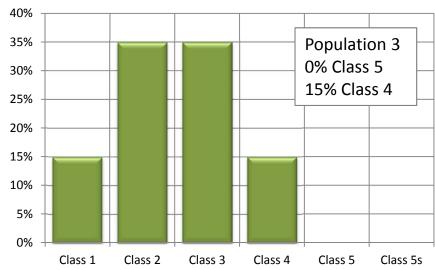




Sensitivity Analysis to Input Population

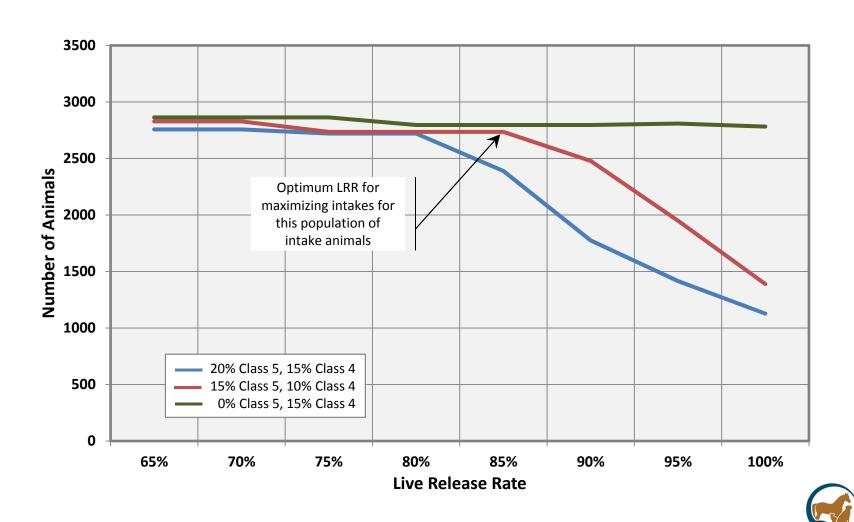






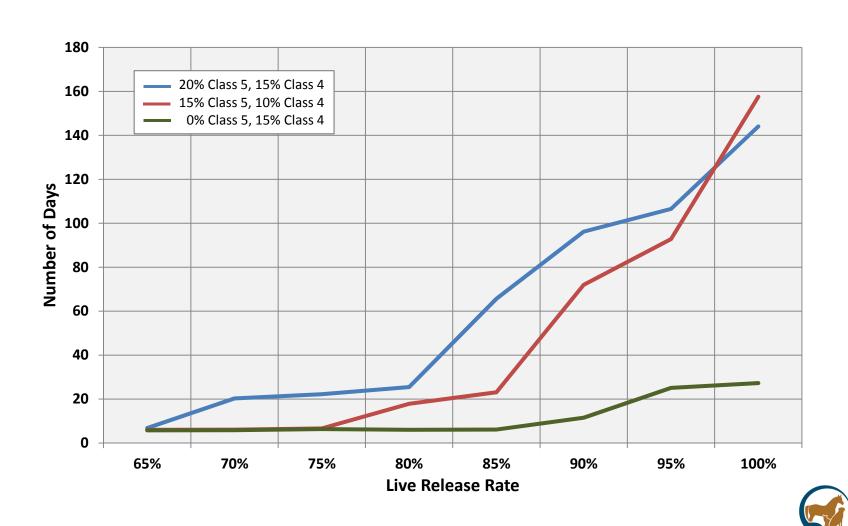


Total Intake at Different LRR



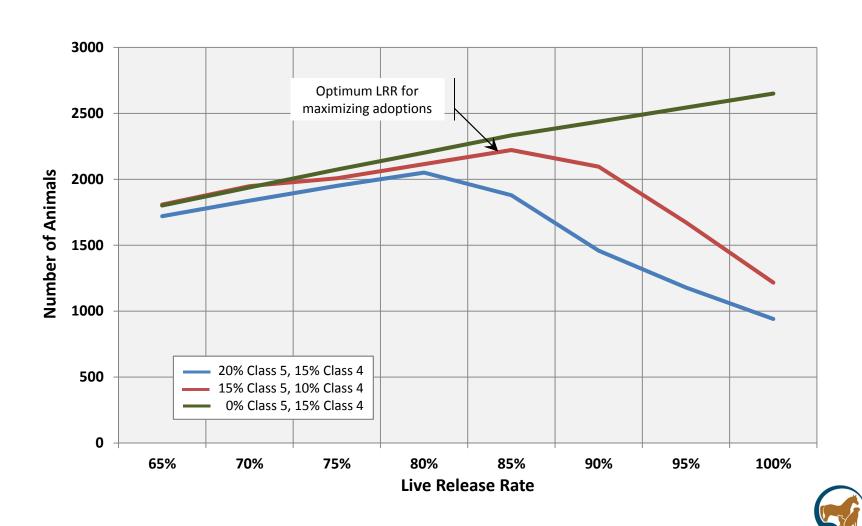
ANIMAL ASSISTANCE

Average LOS at Different LRR



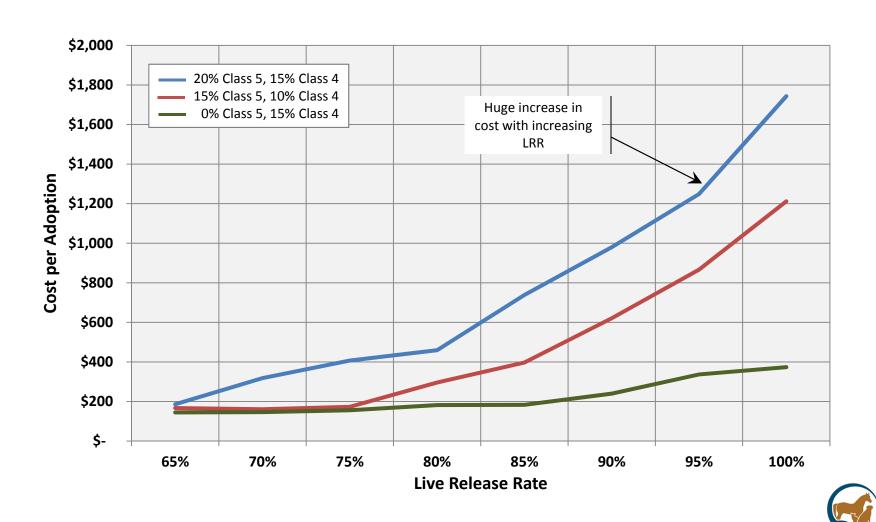
ANIMAL ASSISTANCE

Total Adoptions at Different LRR



ANIMAL ASSISTANCE

Cost/Adoption at Different LRR



ANIMAL ASSISTANCE — FOUNDATION—

Key Points on Intake Population

- Shelter performance is highly dependent on intake population
 - Managed intake facilities can achieve much higher LRR than open admission facilities
- The cost to increase LRR is nearly exponential
 - Both monetarily and in overall animal welfare
- Each population of animals has an optimum LRR that will maximize the number of adoptions and intakes

Where Does this Leave Us?

- Live Release Rate is a critical metric, but we must understand its weaknesses, biases and limitation
 - Dependent on intake population
 - Loses sight of scale
 - Can work against animal welfare as defined by the Five Freedoms if used in isolation



It's a Community Problem

MULTI-SHELTER IMPACT ANALYSIS

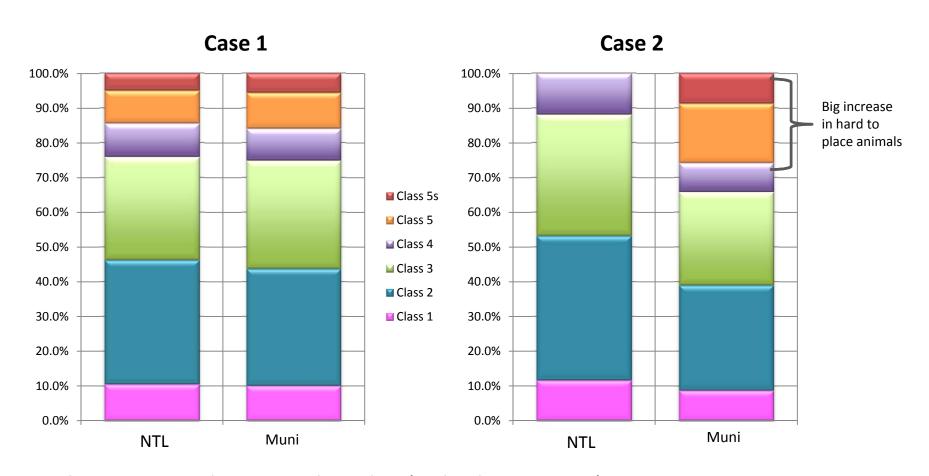


The Tale of Two Shelters

- Purrville has just opened up two identical shelters; they are exactly the same as the one opened in Wagsberg. The animal population has the same statistical parameters as well.
- Both shelters expect about 3000 animals
- This simulation will analyze the impact of intake and euthanasia policies on each of the shelters and the community as a whole. Two different cases highlighted below were analyzed

	Case 1		Case 2		
	No Time Limit	Municipal	No Time Limit	Municipal	
Intake	Any if room	Any if room	Only Class 1 to 4, if room	Any if room	
Euthanasia	None	Class 5 & 5s	None	Class 5s only	

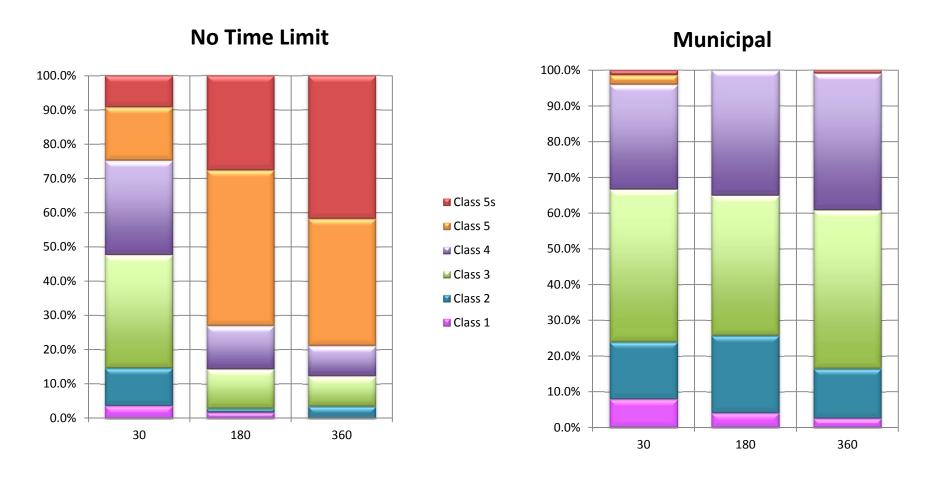
Arrival Population Demographics



Implementing a selective intake policy (Only Classes 1 to 4) at No-Time-Limit Shelter pushes a higher percentage of hard to place (Class 5 & 5s) animals to Municipal Shelter that has an open admission policy

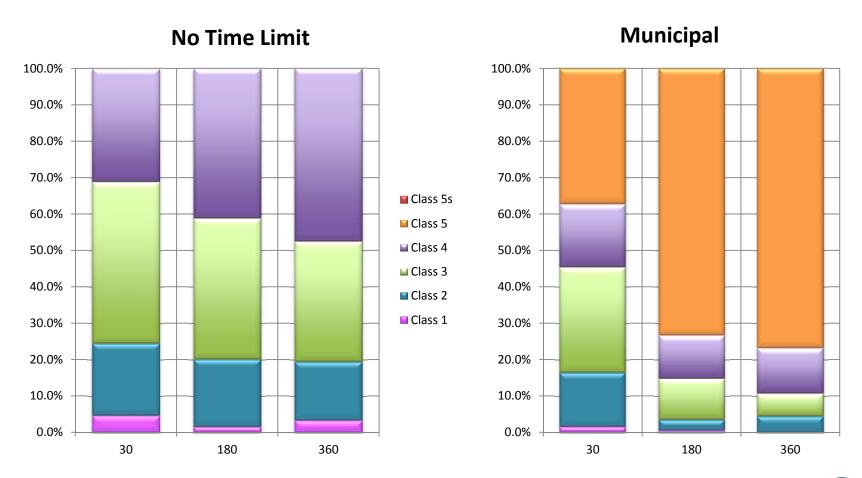


Case 1: Shelter Inventory over Time



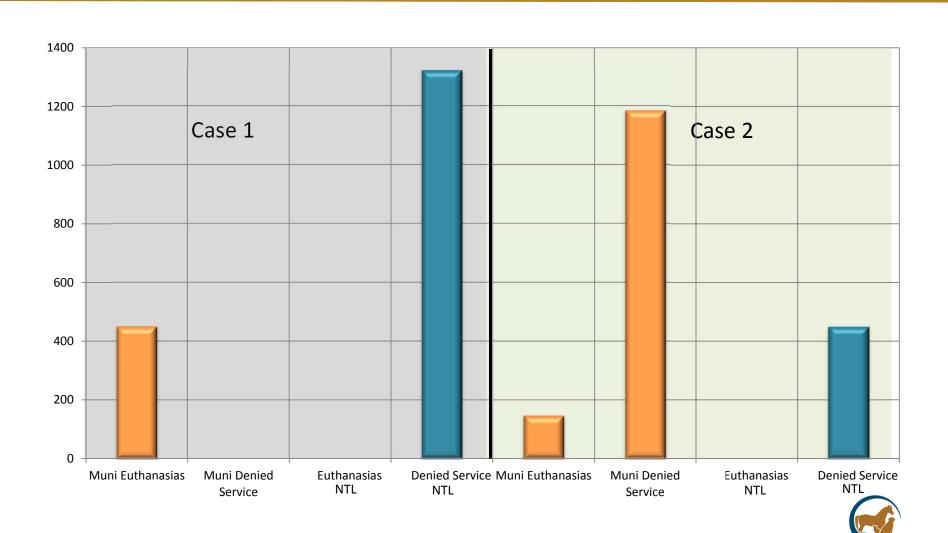


Case 2: Shelter Inventory over Time





Euthanasias & Denial of Service

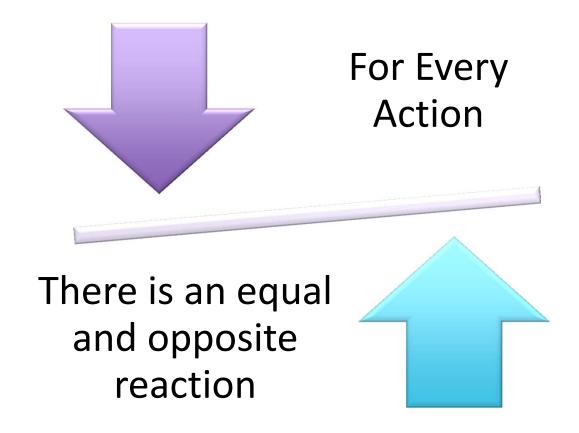


ANIMAL ASSISTANCE — FOUNDATION—

Comparison of Results

Metric	Municipal Shelter			No Time Limit Shelter					
	Case 1	Case 2	Diff			Case 1	Case 2	Diff	
Intakes	2866	1673	(1193)			1544	2858	1314	1
Rejected	0	1185	1185	1		1322	450	(872)	1
Adoptions	2301	1307	(994)	1		1377	2858	1481	1
Euthanasias	450	147	(303)	1		0	0	0	
Ave. Capacity	63%	101%	38%	1		101%	72%	(29%)	1
Ave. LOS	20.4	122	102	1		171	23	(149)	1
% Long-term	0%	73%	73%	1		81%	0%	(1)	1
Annual Cost	832,169	1,756,994	924,825	1		1,723,025	1,010,007	(713,018)	1
Adoption Rev.	441,560	221,457	(220,103)	1		247,026	515,512	268,486	1
\$/ Adoption	362	1,344	983	1		1,251	353	(898)	1
\$/Animal Served	290	1,050	760	1		1,116	353	(763)	↑

Newton's 3rd Law





Comparing expected outcome to real data

WHAT DOES THE MODEL PREDICT



Model Predication: Shelters are Full

Scenario to Evaluate Capacity in No Kill Shelters

Contacted 100 No-Kill shelters around the country

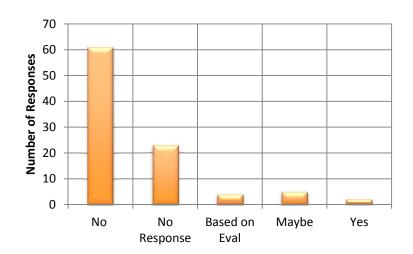


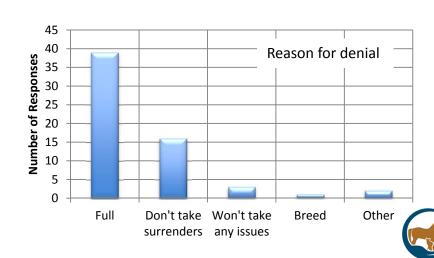
Mr. Squish

ANIMAL ASSISTANCE

50 with Boxer Mix; 50 with Lab Mix

"Hi, this is my dog, Mr. Squish. He is a nice dog, 3 years old, current on shots/heartworm/neutered. He is fighting with my other dog. It started 5-6 months ago and has gotten worse. He likes other people and hasn't fought with any other dogs, but isn't around other dogs much either. I really need a place for him to go so that my other dog can be happy again, he's 9, and I can't rehome him. Please let me know if you can take him into your shelter."





Model Prediction: Long LOS, High Costs

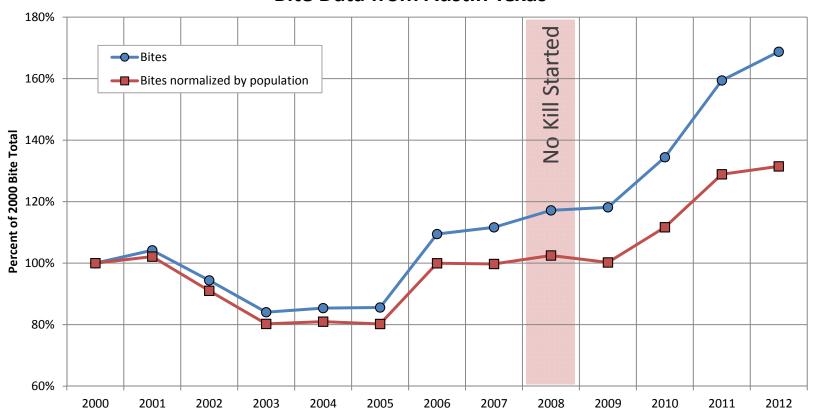
Data from a Major Sanctuary*		
	Model 90% LRR	Sanctuary
Average Length of Stay	4 months	22 months
Cost per Animal (Annual)	\$17,500	\$19,376
Inventory of Difficult Animals	42%	31%
Cost per Adoption	\$1,750	\$9,754
Euthanasia Rate	10%	13%



^{*} Reported to be 90% successful adoptions

Model Prediction: Increase in Unserved Animals

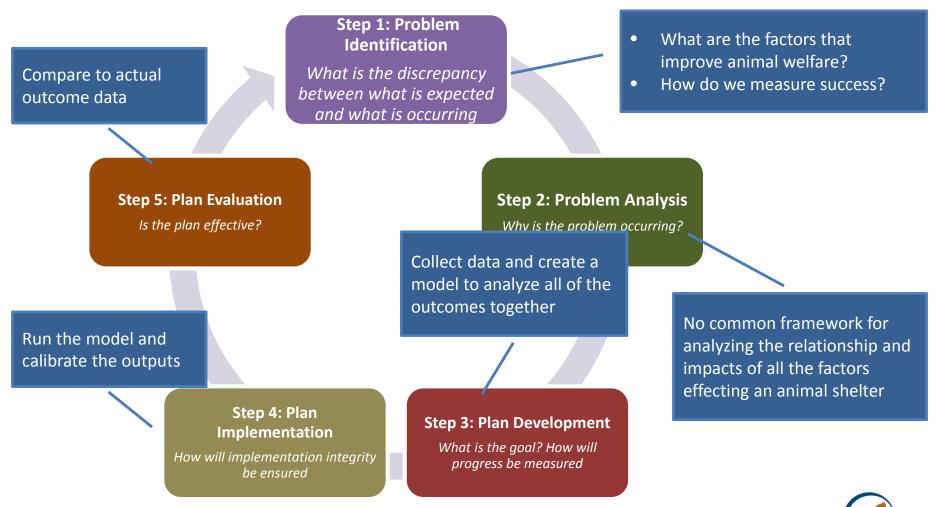
Bite Data from Austin Texas



Inference: More unserved animals means more problematic animals in the community resulting in an increase in dog bites



Stochastic Modeling as Problem Solving





Summary

- LRR is a critical metric, but is not a sufficient measure of success
 - You need counter balancing metrics like Length of Stay
- Increasing LRR from 85% to 95% causes a significant increase in LOS, cost, and denial of service.
- The optimal LRR for a shelter is highly dependent on the intake population and is unique for each community and situation
- Your decisions greatly effect those around you
 - Changing your intake policy by limiting types of animals can have a very negative effect on other shelters in your community
- Animal welfare is much more than LRR; we must measure our success on the sum total of all the factors that contribute to an animal's welfare

Definition of Paradox

A paradox is a statement or concept that contains conflicting ideas. In logic, a paradox is a statement that contradicts itself;

for example, the statement "I never tell the truth" is a paradox because if the statement is true (T), it must be false (F) and if it is false (F), it must be true (T).

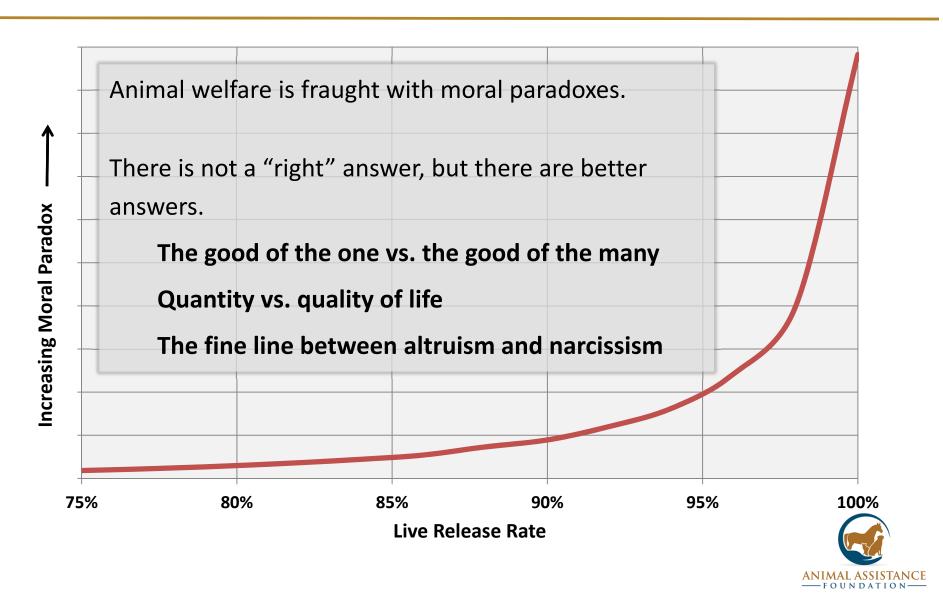


Mission: Improving Animal Welfare

- We have to recognize the we live in a system/community
- What we do effects those around us
- An organization's mission, vision and values must be in the context of the community
- As we get better and better at what we do, we are faced with decisions and consequences that cannot be taken in isolation
- Simplifying the world to a single attribute will ultimately drive organizations away from their intended mission



The Live Release Rate Paradox



Questions?

I am a Paradox. No you're not! Exactly. Ohh...

