### Predicting the effectiveness of Chronic Wasting Disease

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#### Introduction

Chronic Wasting Disease abbreviated as CWD is a transmissible neurological disease that affects cervids such as deer, elk, and moose causing severe impacts on these wildlife populations. The disease, part of a family of diseases known as transmissible spongiform encephalopathies (TSEs), is caused by misfolded proteins known as prions, leading to brain damage and ultimately death in infected animals. It is a fatal disease with no known cure, vaccine, nor environmental decontamination methods that affects the brain and lymphatic system causing neurodegeneration and eventually death. In the USA, it was first discovered in the 1960s, within a government research facility in Colorado, in a captive mule deer. Today CWD has been found in 31 states, 4 providences of Canada, Norway, South Korea, Finland, and Sweden [1]. CWD is the only TSE found in free ranging animals and has been in the spotlight for the last 50 years due to shown lasting negative effects on populations that were thought not possible [3]. Symptoms include loss of coordination, lethargy, drooling, gradual weight loss despite eating, loss of fear of humans, increased thirst and urination, and lowered head with blank stare. The disease can take 1.5 to 3 years to incubate but only 1 year to progress. The disease is originally known to be transmittable by direct contact between the cervids, however, there is an indirect transmission aspect through the depositing of prions in the natural environment through bodily fluids and waste that is part of the disease progression through the deer population [1]. Management culling efforts originally done as generalized mass reduction of the deer population (generalized culling) and recently more direct localized reduction (localized culling) because how generalized has not been proven to work the best due to an increase in prevalence in spite of generalized culling efforts, are the tools that organizations like the Minnesota Department of Natural Resources (MN DNR) have available for management of the disease [2]. The spread of CWD presents a significant challenge for wildlife management and conservation efforts, particularly in regions where hunting and wildlife tourism are critical components of the local economy. Effective management strategies are essential to control the spread of CWD and to maintain healthy wildlife populations.

**Problem Statement** 

Management efforts to control the growth and spread of CWD in deer have been hampered by disagreements with access to land and the efficacy of different types of culling. Our project's goal is to use a spatially explicit model individualized based model (IBM) that simulates deer movements, deer features, the spread of CWD, and hunting and culling, to match output data to as close to reality as possible. This model, being an IBM and spatially explicit, has the benefit that ecologists getting to link individual behavior to population dynamics and the predictions that come with certain changes with individuals with the bonus that spatially explicit models are good for simulating disease transmissions [3]. The results from the simulation allow us to determine whether there is a significant relationship between prevalence of CWD and land access for localized culling. Additionally, through the simulation we can avoid costly real-life testing and other hazards in the field.

#### Research goals

For this project, the main goals that were set up and accomplished were gaining an in depth understanding of Chronic Wasting Disease, learning R in order to understand the R script and the Monte Carlo simulation that was used, figuring out how parameters influence the four pieces of output from the model, and matching those four pieces with the field data the MN DNR supplied by having the closest parameter picks. After these initial goals, our subsequent goals were to use the baseline parameter choices to understand and predict the output for different land access management scenarios and gather our findings for a final presentation and report.

## Methodology

Our first approach to finding a good baseline involved changing the resolution of the simulation. We wanted to simulate a  $28 \times 28$ -mile area; however, a complete simulation of that resolution takes 2-3 days to complete (a complete simulation is one that was run 100 times). Keeping all other variables the same and just decreasing the resolution we realized drastically decreased the simulation time; however, it provided inaccurate results due to dear density and other variables not scaling down properly well as time constraints on figuring out how to properly scale the other variables down to correspond to the resolution. There were 3 variables that were completely adjustable within this simulation: Eta as Whether transmission is density (eta = 0) or frequency (eta = 1) dependent, B.d as the Direct transmission coefficient, and B.i as the Indirect transmission coefficient. See appendix 6 for how it translated to the model code.

Once we figured out that it was necessary to simulate in the 28 x 28 space instead of trying random values and letting the computer run for 3 days, we conducted a Design of Experiment (DOE). Within this DOE 72 tests for plausible variable combinations were come up with and ran a single time. Based on the results, we filtered out combinations that produced poor results and focused on the tests that yielded results close to the field data given by the MN DNR (figure 1). Figure 1 shows the 4 main output our model also gives as number of deer harvested

and tested positive, and number of culled and from the culled positive. From the DOE we found 15 potential combinations of the 3 variables, and from there were further able to get the number down to 7 using different statistical measures such as RMSE, MAE, MAPE, and R – squared. With the 7 combinations with the best results, we ran these simulations multiple times and took the average of the best one to form our baseline. Once we decided on the baseline we decided to run 4 scenarios of 0% management area, 10% area, 50% area, and 100% area and given the results decide if there was an impact in controlling the disease based on land access.

Year	Number of deer harvested	Number of Hunter harvested deer tested	Number of Hunter Harvested deer positive	Shooting Permits	Shooting Permits Positive	Number of deer Culled and Sampled	Number of culled deer positive	Percent of acres accessable for culling		
2016	3578	1239	6	214	2	238	2	UNK		
2017	3167	1665	6	19	0	0	0	NA		
2018	2269	2018	14	328	0	455	12	18.30%		
2019	3096	2991	16	9	0	214	4	9.80%		
2020	3206	1105	9	18	0	282	3	6.70%		
2021	2814	1586	16	21	0	189	8	9.10%		
2022	2647	908	9	0	0	402	16	7.70%		

Figure 1: MN DNR Field Data

#### **Results**

We agreed on a baseline set of parameters that we would not change at all but instead change the amount of land access the spatially explicit model has available to cull deer. Our main 3 parameters eta, B.d, B.i we chose as the numbers 0.0004, 0.002, and 0.0002 respectively. See appendix 6 and 7 for the expanded code version of the full list of parameters such as mortality rates and number of sick deer introduced that still was relevant to our 4-piece output goal. B.d and B.i were important due to being direct and indirect transmission components about how they are semi based on literature as direct transmission has a 100 times greater chance to spread a disease rather than indirect means like in this case prion deposits in carcasses and in the environment. However, these numbers are still unknown on the correct parameter choices, so our choices did not reflect 100 times increase but ten times increase. From our baseline we got a rough estimate of the graph in appendix 2 that corresponded to positive deer prevalence as one of the main outputs but was not the very best we could find if we had more time. From figure 3 we can also see that the number of deer hunted is close on average to figure1 and the number culled is close. However, it is hard to tell if the proportion of deer positive is close to field data, but this is where our checking prevalence graphs (appendix 1) tells us if we are on track with field data.

For the first management scenario as the worst-case scenario for MN DNR to have 0% access of the land to form a localized culling solution in appendix 3 we saw a sharp increase in the average prevalence of the disease over the 26 simulations ran with a 95% confidence interval over the recorded years. The second management scenario is if the MN DNR had slightly more than the usual 8.5% pre given by our model for an increase to 10% land access. If we consult appendix 4, we can see the prevalence is being controlled slightly better than the original 8.5% baseline in appendix 2 with a 95% confidence interval shaded in red for the multiple simulations. The third scenario is if the MN DNR had half the land in the study zone (figure 2) as appendix 4

showed slight peaks in prevalence within the confidence interval range but overall was lower than field data. Finally, the management scenario with 100% access as appendix 5 shows a similar trend that appendix 4 has but there are no peaks of increased prevalence. These results tell the MN DNR that the baseline and model are useful in matching reality (figure 1) and can be used to help convince landowners that more land access can help drop the increase of CWD prevalence.

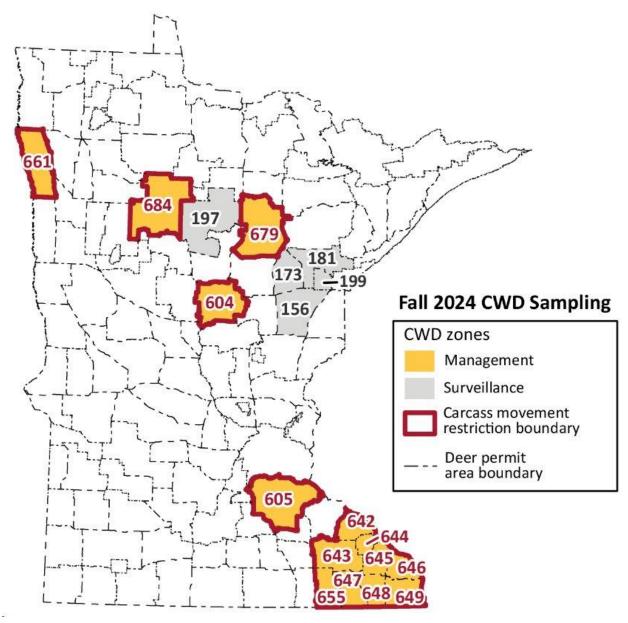


Figure 2: study zone of MN DNR field work

&_Munt_I	N_Hunt_2	N_Hunt_3	N_Hun	r_4 N_Hur	nt_5 N	Munt_6	N_Hunt_	P_Hunt,	1 P_Hunt_2	P_Hunt_3	P_Hunt_4	P_Hunt_5	P_Hunt_6	P_Hunt_7	N_Cutt_1	N_Cull_	2 N_Cul	L3 N_C	Cull_4	N_Cutt_5	N_Cull_6	N_Cull_7	P_Cull_1	P_Cull_2	P_Cull_3	P_Cull_4	P_Cull_5	P_Cull_6	P_Cuil_7
NA	NA	NA	NA	NA	N.	IA.	NA .	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA	NA.	NA	NA	NA
3286	3528	3725			1588	5023	571	7 0.0020	2 0.00292	4 0.001823	0.002457	0.001541	0.003546	0	17	4 3	90	190	131	149	336		0.005747	0.071053	0.1	0.21374	0.147651	0.08631	. NA
2965	3109	3491			985	4351	489	0.0011	0.0010	0.00198	0.008219	0.003549	0.006279	0.002185	15	8 5	97	286	638	465	389	21	0.025316	0.020619	0.122378	0.109718	0.148387	0.143959	0.080952
3022	3013	3367	34	670 4	1041	4520	483	2 0.0045	15	0.00207	0	0.003478	0.007886	0.003644	43	5	0	296	0	239	626	40	3 0.013793	NA.	0.060811	NA	0.079498	0.095847	0.141439
3004	3152	3388			1044	4495			88 0.00208						27		32	241	97	289	754	70						0.098143	
3086	3238	3548			1136	4542			88 0.00100					0.002007	22		81	332	254	363	0		0.026667						0.111538
3017	3139	3504			1043	4537	483	0.0021	16	0.004032	0.002705	0.001779	0.003165	0.004342	23	5	0	332	339	143	329	39	4 0.029787	NA	0.114458	0.097345	0.153846	0.06079	0.045685
3003	3028	3419			1028	4236			0.00106						14			238	537	792	407		3 0.02027						
3047	3081	3426			1027	4570			0.00213						17	3 2		122	165	311	423		4 0.028902						
3175	3509	3663			1472	4957			71 0.00389						37	2 3	07	512	248	57	301		1 0.034946						
3040	3199	3408			1131	4347		0.002		0.002039				0.002939	27	7	0	126	560	1096	421		4 0.025271					0.104513	
3044	3214	3457			1157	4569			0.00211						34		24	192	382	192	68		0.02346						
2941	3136	3484			1075	4436		6 0.0025						0.005185	18	5	0	0	319	770	799		3 0.032432		NA:			0.051314	
3040	3099	3403			1122	4405		0.0011		0.006054					19	В	0	593	199	430	731		8 0.025253					0.082079	
3088	3282	3488			1011	4550		0.0011		0.001951				0.003336	18	2	0	275	153	0	353		4 0.005495			0.202614			0.074924
3005	3196	3377			1010	4501			17 0.00408						32	9 30		204	71	172			6. 0.006079						
3068	3134	3366			1092	4621			6 0.00304						10		21	126	359	489	363		5 0.009259						
3184	3127	3466			1046	4479		9 0.0032		0.005128					43		0	425	540	73	340		5 0.013921					0.061765	
2988	3170	3485			1192	4483			15 0,00324						9		14	185	444	269	81		6 0.022222					0.148148	
3053	3160	3367			1213	4623			0.00108					0.002761	43		14	0	252	87	469		3 0.016018					0.044776	
3017	3220	3501			1024	4531			14 0.00099						17		30	129	872	371	194		1 0.023121					0.268041	
2989	3171	3528			1132	4546			44 0.00210						18	-	74	134	393	329	702		0.022099						
3043	3035	3396			3987	4596			19 0.00109						12		56	116	153	180	468		3 0.015748						
3545	3780	4059			1837	5374			76 0.00623						59			490	185	217	444		6 0.08769						
2987	3180	3452		665 3	3922	4256			0.00097						22	2 16	52	349	564	255	201		7 0.009009						
3080	3063	3473	3	866 4	1167	4413	482	7 0.0011	14	0.003883	0.011111	0.004219	0.00464	0.002235	12	9	0	562	747	307	354	23	8 0.015504	NA.	0.072954	0.123159	0.068404	0.19209	0.151261

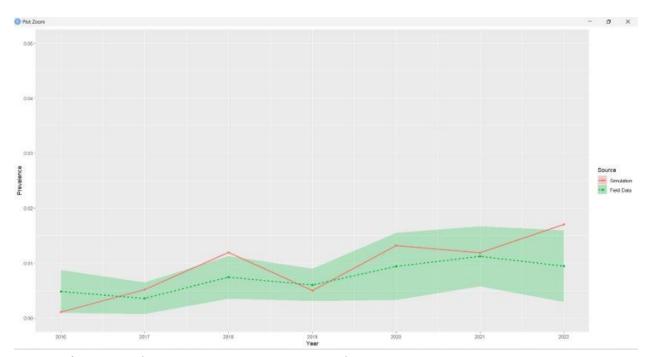
Figure 3: baseline output

#### Conclusion

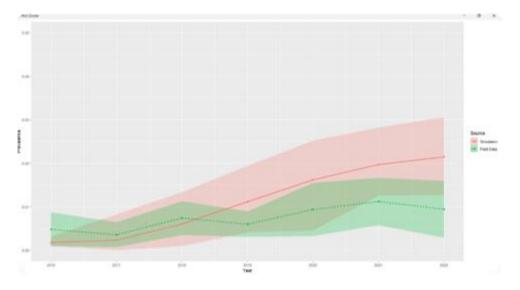
In conclusion, we achieved our main goals of better understanding the parameters that make the model, finding a baseline to match the field data as close to reality and getting results for management scenarios. However, there were some limitations in the process that hampered the work regarding time constraints because finding parameters for the baseline takes days, thus it isn't perfect. However we saw a significant decrease in prevalence of the management graphs and can apply this work to MN DNR work by achieving three things: helping the organization save time and money due to the simulation being useful, give findings to convince land owners to increase land access, and finally helping the MN DNR understand where to allocate more resources such as testing, culling, or projects that help communicate with landowners about the disease management. For future goals for further work on the project we could use the expanded time to better optimize the code for faster run times, test out better baseline parameters with help from statistical measures and explore more specific management scenarios.

# Appendix

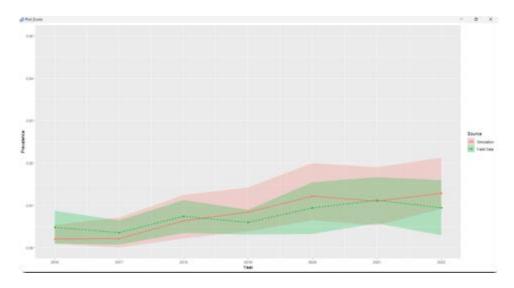
**Appendix 1:** Baseline parameter output graph



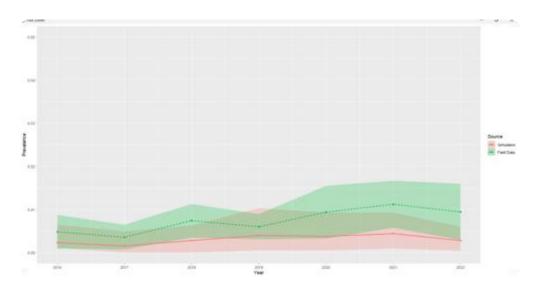
**Appendix 2:** Baseline parameter output graph with 0% land access



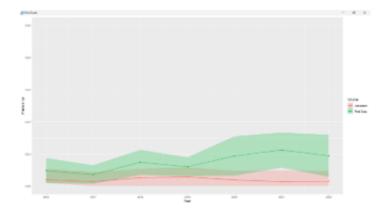
Appendix 3: Baseline parameter output graph with 10% land access



**Appendix** 4: Baseline parameter output graph with 50% land access



**Appendix 5**: Baseline parameter output graph with 100% land access



**Appendix 6:** Code snippets of main general changeable parameters

```
26 - {
27
      deer.dens <- 25 # Deer per square mile
                                                              # MINORLY ADJUSTABLE
28
29
      simulationAreaDims <- 1760 * 28.0 # In denominations of 1 mile = 1760 yards
30
      managementCellSize <- 1760 * 1.00 # In denominations of 1 mile = 1760 yards
      cellularResolution <- 1760 * 0.25 # In denominations of 1 mile = 1760 yards
31
      originatedAreaDims <- 1760 * 5.00 # In denominations of 1 mile = 1760 yards
32
33 - }
34
35 # Define parameters for disease transmission (ALL FULLY ADJUSTABLE)
36 - {
37
      eta <- 0.0004 ## Whether transmission is density (eta = 0) or frequency (eta = 1) dependent
      B.d <- 0.002 ## Direct transmission coefficient (100x larger than indirect)
38
      B.i <- 0.0002 ## Indirect transmission coefficient
```

#### **Appendix 7:** Code snippets of general changeable parameters continued

```
# FULLY ADJUSTABLE
45
      numberOfIntroduced <- 25
46
      yearsOfManagement <- 7
47
48
      managementThreshold <-c(0.001, 0.008) # Desiring a mean of 0.0048
49
50
      probabilityOfTesting <- 0.400 # Excludes fawns because they were not tested</pre>
51
      probabilityOfCulling <- 0.850
52
      probabilityOfAccess <- 0.1 #0.085
53
54
       ## Establish priority and marginal cull rates across eight-week periods
55 -
         priorityCullRate <- 1 - (1 - probabilityOfCulling) \land (1 / 8) marginalCullRate <- 1 - (1 - probabilityOfCulling) \land (1 / 8)
56
57
58 -
59 - }
60
61 # Define parameters for mortality (ALL MINORLY ADJUSTABLE)
62 - {
       naturalMortalityRate <- c(0.55, 0.10, 0.05) # Non-harvest mortality for juveniles, yearlings, adults
63
64
      hunt.antl <- 0.15 # Proportion of bucks that get harvested #0.25 og
      hunt.less <- 0.10 # Proportion of does/fawns that get harvestetd #0.20 og
```

#### References

[1] Thompson, N. E. (2023). *Management of free-ranging white-tailed deer with chronic wasting disease in Michigan* (dissertation).

- [2] Almberg, Emily S., Cross, Paul C., Johnson, Christopher J., Heisey, Dennis M., & Richards, Bryan J. (2011). *Modeling routes of Chronic Wasting Disease Transmission:*Environmental prion persistence promotes deer population decline and extinction (dissertation).
- [3] Kjaer, L. J. (2010). *Individual-based modeling of white-tailed deer (Odocoileus virginianus) movements and epizootiology* (dissertation).