

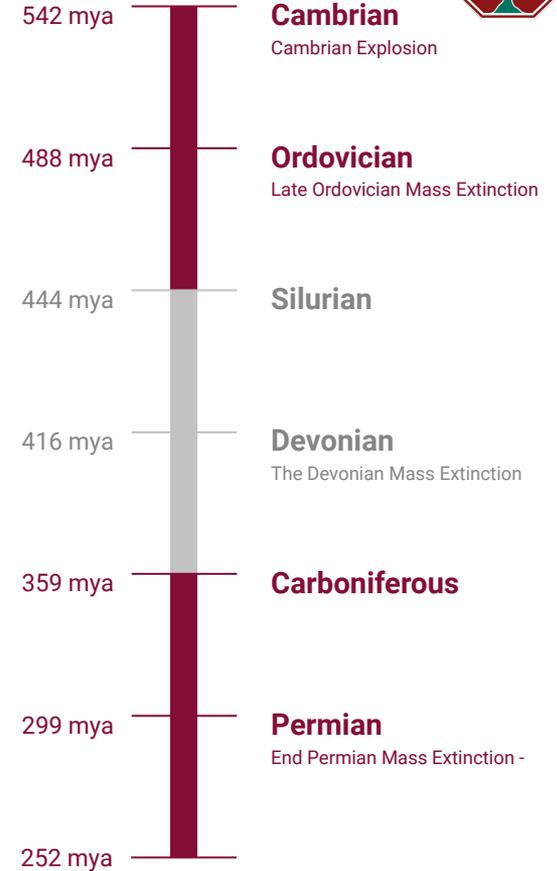
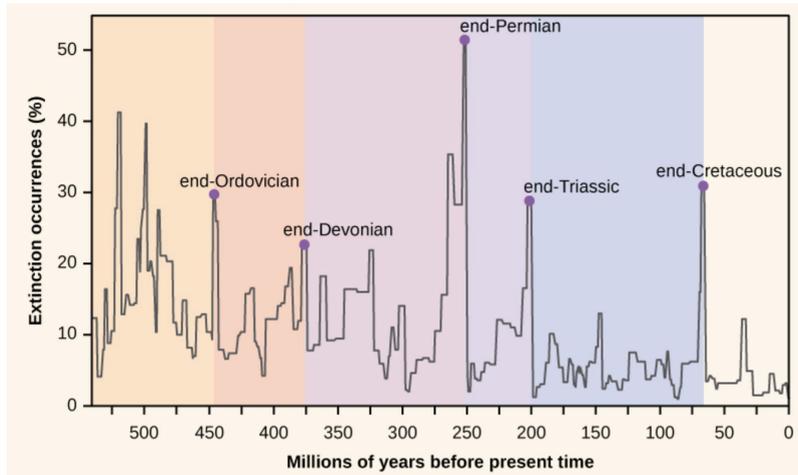
Using Machine Learning Models and Logistic Regression Analyses to Develop a Comprehensive Understanding of Extinction Risk For Marine Animal Phyla Across the Paleozoic

Adarsh S. Ambati, Anya Sengupta, Theo Chiang, Dr. Pedro Monarrez, Michael Pimentel-Galvan, Dr. Noel Heim, Dr. Jonathan Payne



Basic Info/Background

- Extinction is volatile and relies on many ecological and biological variables
- Nature of extinction risk varies across geologic time; we focus on the Paleozoic Era





Questions

1. Were certain taxonomic groups of organisms preferentially selected for extinction during the Paleozoic era?
2. Were certain types of organisms preferentially selected for extinction during the Paleozoic era?
3. Is it possible to predict when a particular genus goes extinct during the Paleozoic using characteristics/descriptors?



Methodology

- Phyla - Echinodermata, Mollusca, Chordata, Arthropoda, Brachiopoda
- Descriptors - buffering, feeding patterns, motility, oceanic tiering, respiratory organ type, circulatory system type, length, surface area, volume
- Regularized binomial regression models(Step 3) + logistic binomial regression(Step 1/2)
 - Built Using R
 - Uses Stanford Earth Body Size Dataset (n=8816)

```
Environment History Connections Tutorial
R • Global Environment
Data
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$ tiering      : int 3 3 3 3 3 3 3 3 3 ...
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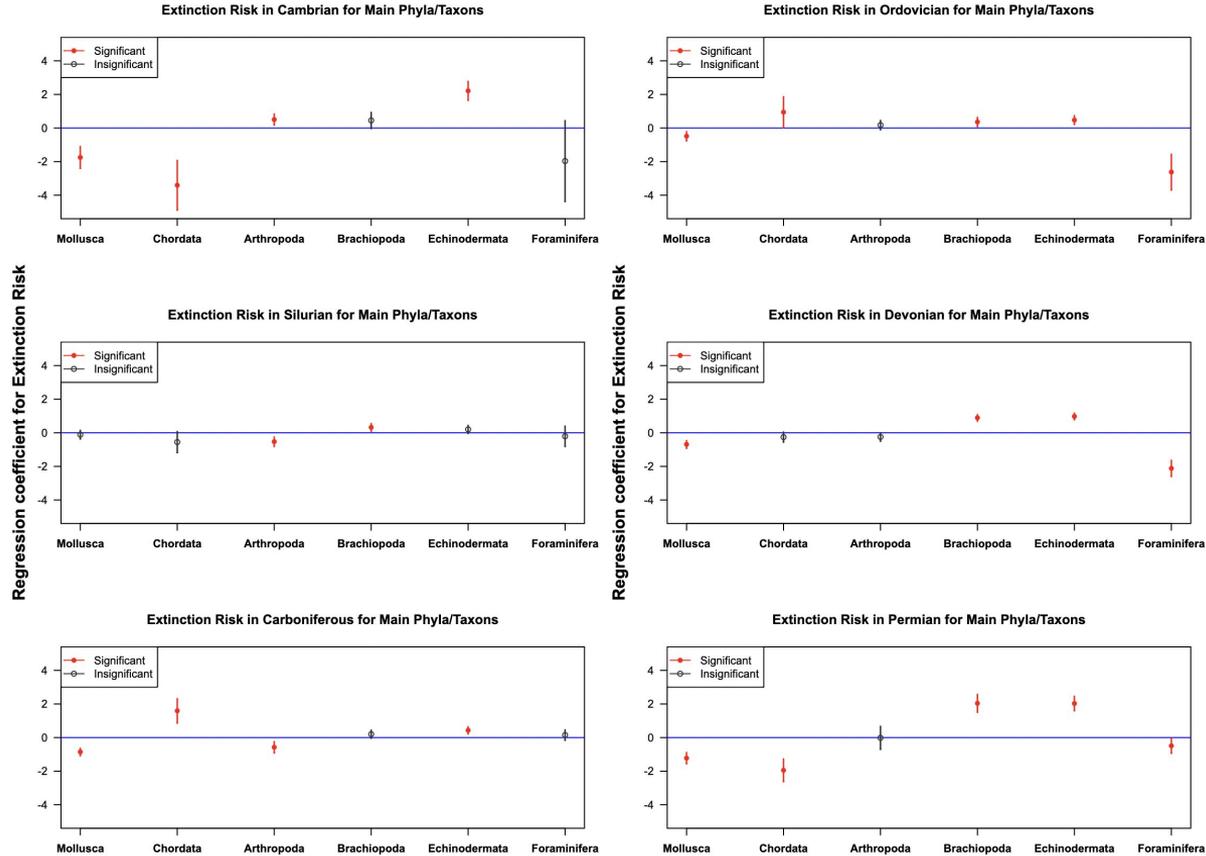


Methodology Cont.

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\$ brachiopoda	: num	0	0	0	0	0	0	0	0	0	0	0	0	...
\$ echinodermata	: num	1	1	1	1	1	1	1	1	1	1	1	1	...
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\$ excambrian	: num	0	0	0	0	0	0	0	0	0	0	0	0	...

- Step 1: Binomial Logistic Regression Analysis on Phyla/Class/Order during each stage of Paleozoic identifying likeliness of extinction on each class
- Step 2: Binomial Regression Analysis on Phyla/Class/Order during each stage of Paleozoic identifying likeliness of extinction on each genus descriptor (ie predatory feeding, facultative motility, benthic tiering)
- Step 3: Developing simple machine learning model (using regularization) to predict whether a taxonomic group/specimen goes extinct in a specific period

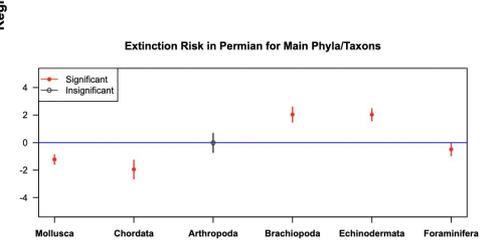
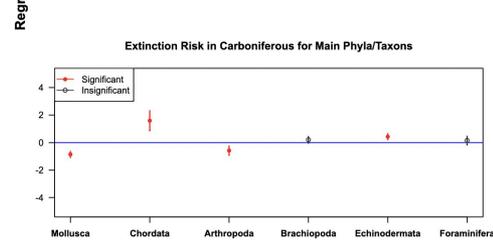
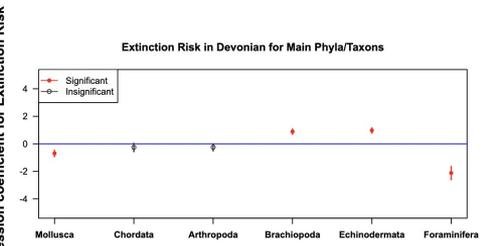
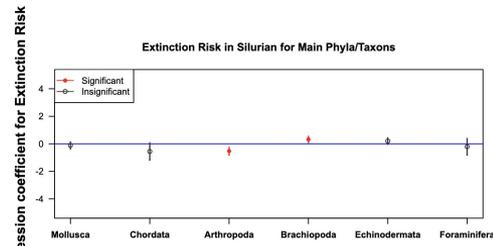
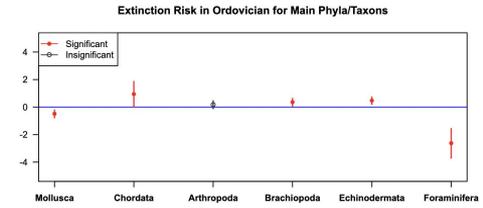
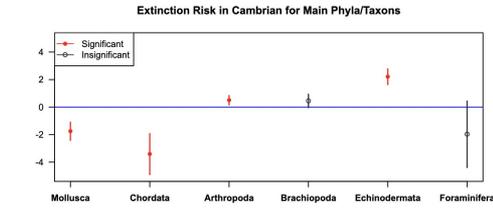
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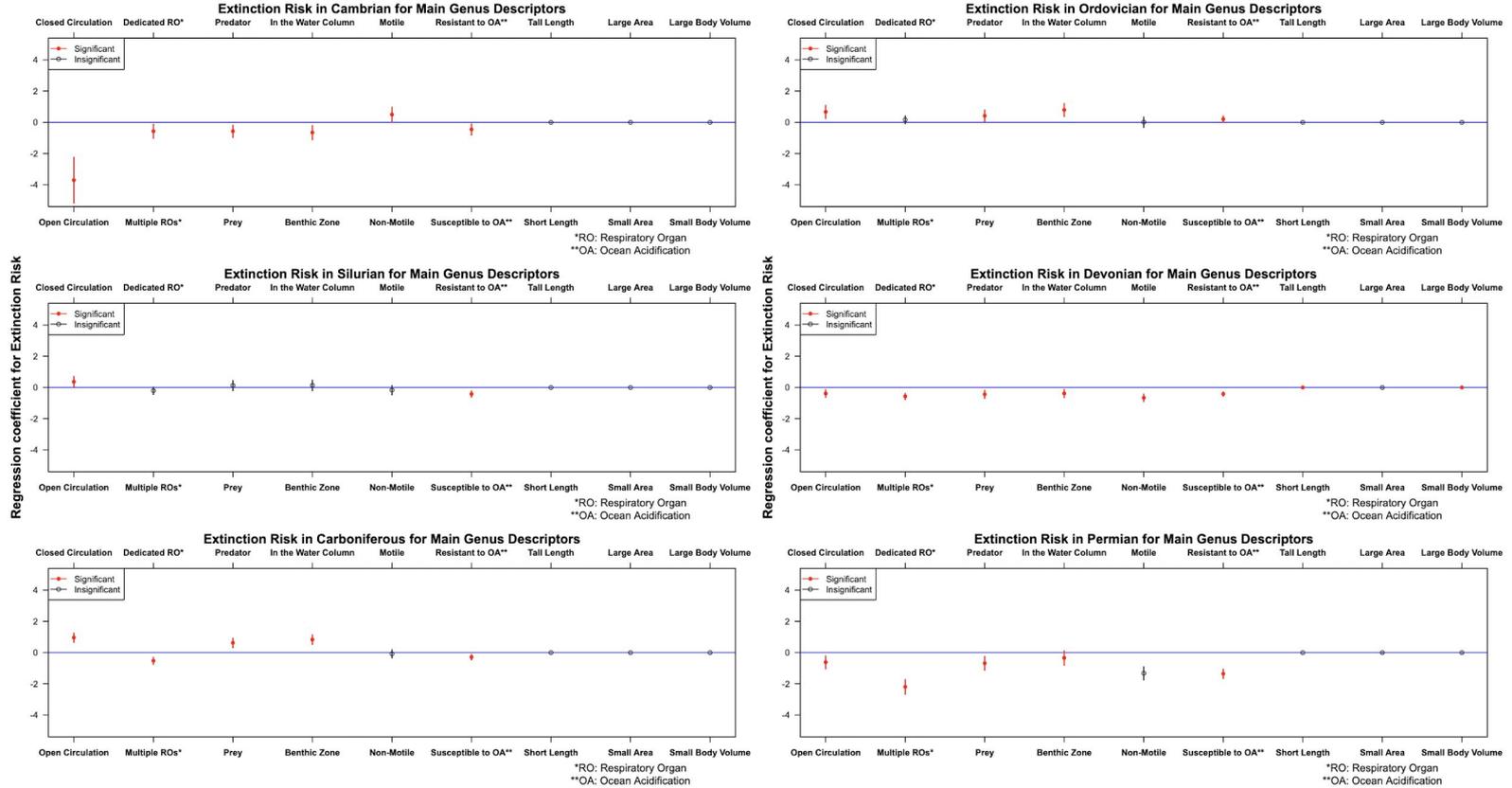


Step 1: Key Findings

- **23 of 36** data points were significant
 - 12 data points had a significant greater extinction risk
 - 11 data points were significantly selected for survival.
- **Patterns**
 - Mollusca — consistently selected for survival
 - Brachiopoda— (major extinction events in Devonian and Permian, reflected in graphs)
 - Echinodermata — consistently selected for extinction
- **Extinction Risk was not uniformly felt across all major phyla in the Paleozoic**



Step 2:

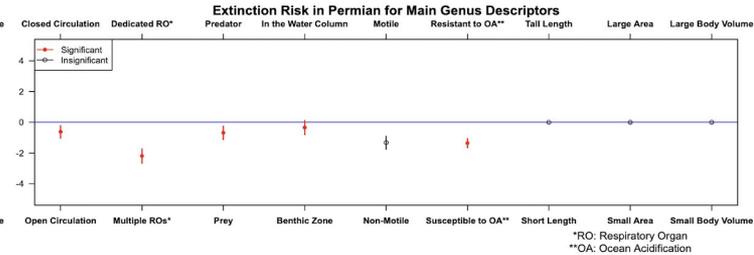
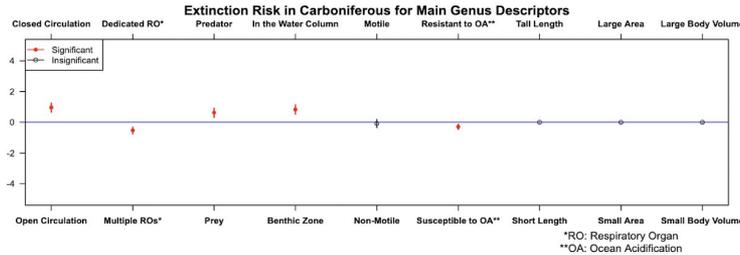
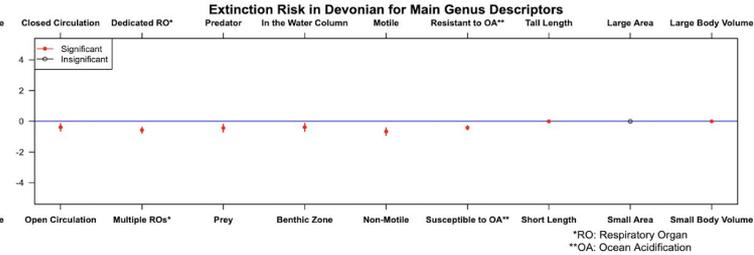
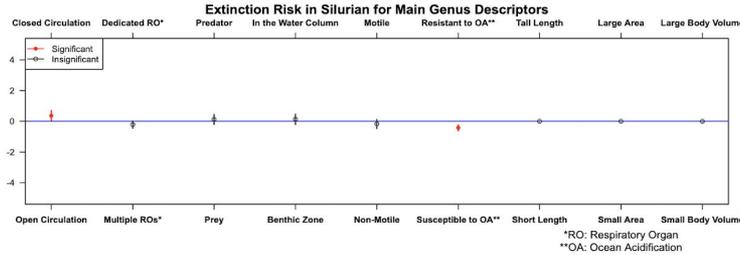
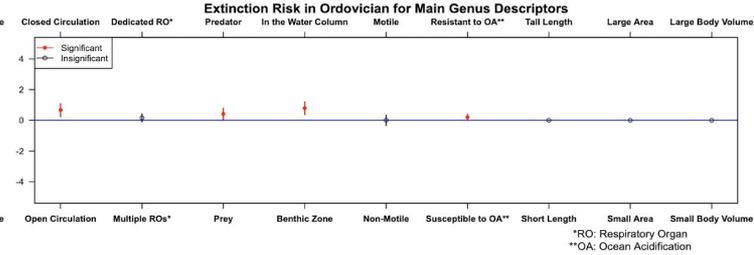
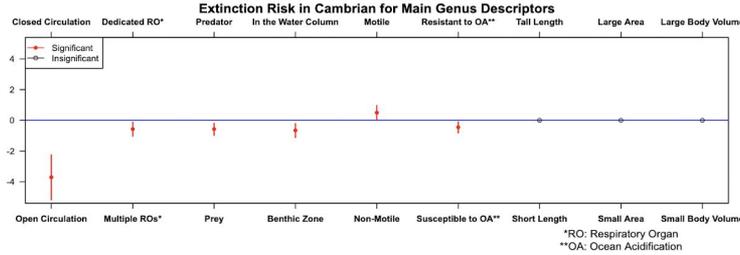




Step 2: Key Findings

➤ 30 of 54 data points were significant

- Patterns
- Body Size
 - Circulatory Systems (very volatile)
 - Respiratory Organ Type





Step 3: Machine Learning Model

- 6 binomial regression models with regularization for each period
 - Regularization helps in two major ways:
 - Reducing the variance of the model so as to not cause it to become overfit to the training data
 - Helps determine the features that cause the model to increase in variance and removes or shrinks their contribution to the model. (Essentially, features that do not predict have their coefficients reduced)
- Features:
 - Circulatory systems
 - Type of Respiratory Organ
 - Feeding (Predator or Prey)
 - Tiering (Water Column or Benthic)
 - Motility (Freely Moving or Non-motile)
 - Ability to withstand Ocean Acidification
 - Maximum length
 - Maximum Area
 - Maximum Volume
- **Predicting Outcome:** whether or not a genus went extinct in a particular period



Step 3: Machine Learning Model

Three types of regression models (lasso, elastic net, and ridge regression) were tested:

- ❖ **Lasso:** Focused on Feature Elimination (penalizes by removing)
- ❖ **Ridge Regression:** Focused on Feature Coefficient Reduction (penalizes by shrinking)
- ❖ **Elastic Net:** middle of both
- ❖ Nomenclature: Lasso : $\alpha = 1$; Ridge : $\alpha = 0$; Elastic net: $0 < \alpha < 1$
- ❖ The best model identified by running the program for α values between 0-1



Step 3: Machine Learning Model

circ	respOrgan	feeding	tiering	motility	oceanacidificationwithstand	max_length	max_area	calc_max_vol	excambrian
0	0	1	0	1	0	34.700000	3782.7603	1.499532e+04	0
0	0	1	0	1	0	181.200000	103149.2939	1.332594e+06	0
0	0	1	0	1	0	18.232000	1044.2837	2.613302e+03	0
0	0	1	0	1	0	55.450000	9659.4628	5.353129e+04	0
0	0	1	0	1	0	64.057376	12891.0460	7.920197e+04	0
0	0	1	0	1	0	37.300395	4370.9591	1.824551e+04	0
0	0	0	0	1	0	37.269746	4363.7789	1.820484e+04	0
0	0	0	0	1	0	40.702486	5204.6525	2.312418e+04	0
0	0	0	0	1	0	91.940000	26555.7700	2.112418e+05	0
0	0	0	0	1	0	41.897815	5514.8362	2.501452e+04	0
0	0	0	0	1	0	43.493046	5942.7783	2.768527e+04	0
0	0	0	0	1	0	10.141566	323.1171	5.316888e+02	0
0	0	0	0	1	0	45.810729	6593.0184	3.187570e+04	0
0	0	1	0	1	0	46.668914	6842.3497	3.352294e+04	0
0	0	0	0	0	0	27.050000	2298.7112	3.482660e+04	0

Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Cambrian	0	0.01	0.92	0.93	0.8
Cambrian	0.1	0.01	0.92	0.93	0.8
Cambrian	0.2	0.01	0.92	0.93	0.8
Cambrian	0.3	0.01	0.92	0.93	0.8
Cambrian	0.4	0.01	0.92	0.93	0.8
Cambrian	0.5	0.01	0.92	0.93	0.8
Cambrian	0.6	0.01	0.92	0.93	0.8
Cambrian	0.7	0.01	0.92	0.93	0.8
Cambrian	0.8	0.01	0.92	0.93	0.8
Cambrian	0.9	0.01	0.08	0.5	0.79
Cambrian	1	0.01	0.91	0.5	0

Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Ordovician	0	0.012589254	0.83	0.67	0.73
Ordovician	0.1	0.01	0.83	0.68	0.73
Ordovician	0.2	0.012589254	0.83	0.67	0.73
Ordovician	0.3	0.01	0.83	0.67	0.73
Ordovician	0.4	0.01	0.83	0.67	0.73
Ordovician	0.5	0.01	0.83	0.66	0.73
Ordovician	0.6	0.01	0.82	0.66	0.73
Ordovician	0.7	0.01	0.82	0.66	0.73
Ordovician	0.8	0.01	0.82	0.66	0.74
Ordovician	0.9	0.01	0.82	0.61	0.74
Ordovician	1	0.01	0.82	0.61	0.75

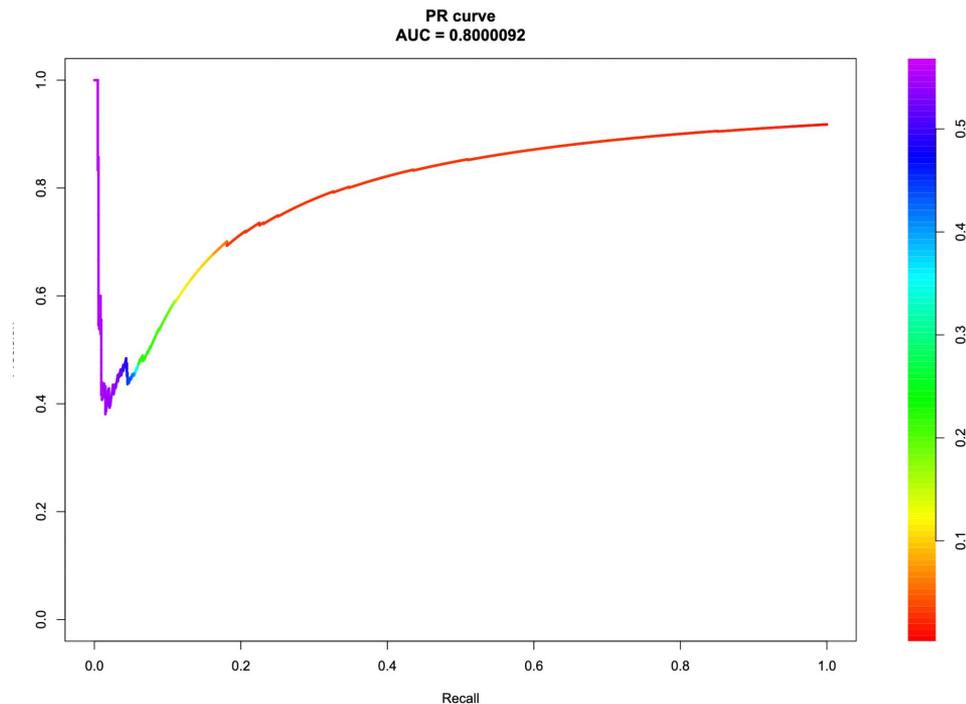
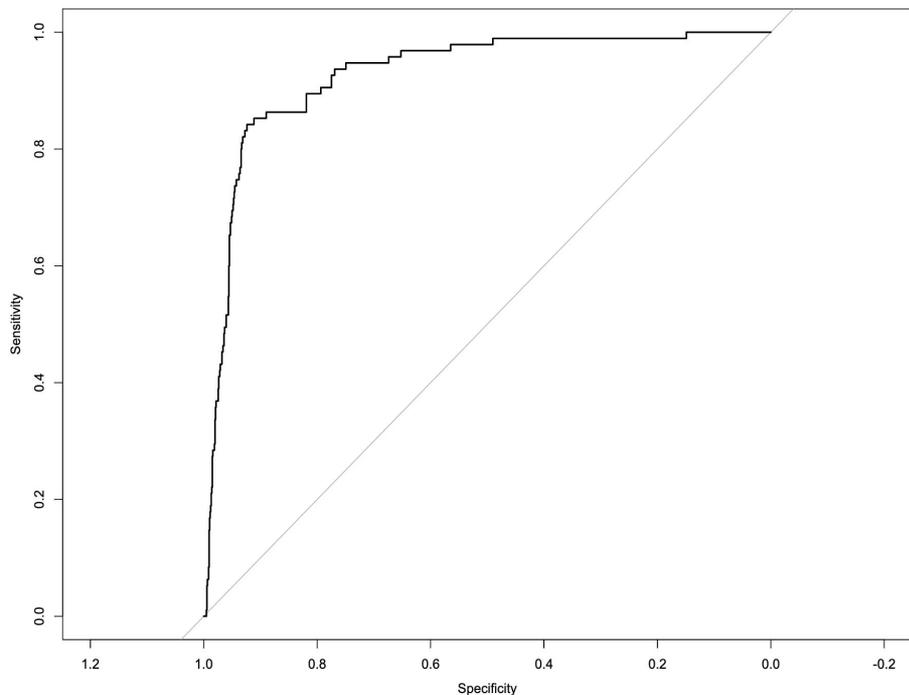
Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Silurian	0	0.005416508	0.91	0.6	0.89
Silurian	0.1	0.001219581	0.91	0.6	0.89
Silurian	0.2	0.000578704	0.91	0.6	0.89
Silurian	0.3	0.000578704	0.91	0.6	0.89
Silurian	0.4	0.000578704	0.91	0.6	0.89
Silurian	0.5	0.000578704	0.91	0.6	0.89
Silurian	0.6	0.000578704	0.91	0.6	0.89
Silurian	0.7	0.000578704	0.91	0.6	0.89
Silurian	0.8	0.000578704	0.91	0.6	0.89
Silurian	0.9	0.000578704	0.91	0.6	0.89
Silurian	1	0.000578704	0.91	0.6	0.89

Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Devonian	0	0.025118864	0.79	0.64	0.75
Devonian	0.1	0.015848932	0.79	0.64	0.75
Devonian	0.2	0.019952623	0.79	0.64	0.75
Devonian	0.3	0.006309573	0.79	0.64	0.75
Devonian	0.4	0.003981072	0.79	0.64	0.75
Devonian	0.5	0.003981072	0.79	0.64	0.75
Devonian	0.6	0.003162278	0.79	0.64	0.75
Devonian	0.7	0.001584893	0.79	0.64	0.75
Devonian	0.8	0.002511886	0.79	0.64	0.75
Devonian	0.9	0.001995262	0.79	0.64	0.75
Devonian	1	0.003162278	0.79	0.64	0.75

Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Carboniferous	0	0.003541732	0.83	0.6	0.77
Carboniferous	0.1	0.002915452	0.83	0.6	0.77
Carboniferous	0.2	0.011383466	0.83	0.6	0.77
Carboniferous	0.3	0.002915452	0.83	0.6	0.77
Carboniferous	0.4	0.002915452	0.83	0.6	0.77
Carboniferous	0.5	0.002915452	0.83	0.6	0.77
Carboniferous	0.6	0.003541732	0.83	0.6	0.77
Carboniferous	0.7	0.002915452	0.83	0.6	0.77
Carboniferous	0.8	0.002915452	0.83	0.6	0.77
Carboniferous	0.9	0.002915452	0.83	0.6	0.77
Carboniferous	1	0.002915452	0.83	0.6	0.77

Period	Alpha Value	Lambda Value	Accuracy	AUROC	AUPR
Permian	0	0.002915452	0.84	0.62	0.79
Permian	0.1	0.002915452	0.84	0.62	0.79
Permian	0.2	0.002915452	0.84	0.63	0.78
Permian	0.3	0.002915452	0.84	0.62	0.79
Permian	0.4	0.002915452	0.84	0.56	0.82
Permian	0.5	0.002915452	0.84	0.56	0.82
Permian	0.6	0.002915452	0.84	0.56	0.82
Permian	0.7	0.002915452	0.84	0.56	0.82
Permian	0.8	0.002915452	0.84	0.45	0.83
Permian	0.9	0.002915452	0.84	0.46	0.83
Permian	1	0.002915452	0.84	0.46	0.83

Step 3: Machine Learning Model Results



Area under ROC Curve of **0.93** (ability to distinguish extinction and survival) and Area under PR Curve of **0.80** (fewer prediction errors)



Step 3: Machine Learning Model Results

	Cambrian Model Features	Score
1	Circulation System	3.3422079
2	Motility	2.27727063
3	Ocean Acidification Resistance	1.12284708
4	Feeding Patterns	0.42942449
5	Respiratory Organ System	0.37019192
6	Tiering	0.02890546
7	Maximum Length	0.00479994
8	Maximum Area	4.23E-07
9	Calculated Maximum Volume	2.39E-09

	Ordovician Model Features	Score
1	Motility	1.06918333
2	Ocean Acidification Resistance	0.85323525
3	Tiering	0.33049509
4	Feeding Patterns	0.32654485
5	Respiratory Organ System	0.15505047
6	Circulation System	0.02294367
7	Maximum Length	0.00124485
8	Maximum Area	4.10E-06
9	Calculated Maximum Volume	1.08E-08

	Silurian Model Features	Score
1	Motility	1.32779174
2	Circulation System	1.06206068
3	Respiratory Organ System	0.70851463
4	Feeding Patterns	0.6192809
5	Ocean Acidification Resistance	0.59558913
6	Tiering	0.59509338
7	Maximum Length	0.00258455
8	Maximum Area	3.22E-06
9	Calculated Maximum Volume	3.32E-08

	Devonian Model Features	Score
1	Ocean Acidification Resistance	0.70835892
2	Respiratory Organ System	0.45876009
3	Motility	0.22309167
4	Circulation System	0.06226351
5	Maximum Length	0.00555858
6	Maximum Area	2.00E-05
7	Calculated Maximum Volume	4.32E-08
8	Feeding Patterns	0
9	Tiering	0

	Carboniferous Model Features	Score
1	Circulation System	2.32544688
2	Ocean Acidification Resistance	0.73146419
3	Respiratory Organ System	0.65025381
4	Maximum Length	0.00189793
5	Feeding Patterns	0
6	Tiering	0
7	Motility	0
8	Maximum Area	0
9	Calculated Maximum Volume	0

	Permian Model Features	Score
1	Circulation System	2.18390017
2	Respiratory Organ System	1.04694225
3	Ocean Acidification Resistance	0.93551679
4	Feeding Patterns	0.86094481
5	Tiering	0.74849748
6	Motility	0.70993468
7	Maximum Length	0.000241
8	Calculated Maximum Volume	2.56E-08
9	Maximum Area	0



Major Takeaways

- Extinction Risk is not uniform across both geologic history or across taxonomic groups
- Certain traits can act as indicators for higher extinction risk; however, these too vary across geologic history
- These traits can even be used to create relatively accurate predictive models approximating what period a genus went extinct
- **One major goal of our project was to act as a comprehensive foundation for future research:** each one of the traits or phyla from stages one and two can be further studied.



Future Research

- Completing analysis across the rest of geologic history
 - Identifying patterns of how extinction risk changes for each phyla and each trait across every period in Earth's history
 - Ie. How anoxic conditions affect extinction risk for each phyla/trait
 - Ie. How mass extinctions affect extinction risk for each phyla/trait
- Testing out Decision Tree or Random Forest Regression Models to predict exact first and last appearance in geologic history for each genus
- Look into Building Neural Nets for this type of prediction

Acknowledgements



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Citations

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