

# SeeGull: Pacific Sea Gull Activity Classification and Path Prediction

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# **Abstract**

**Project goal:** Pacific sea gull behavior prediction, including:

- activity
   classification based
   on positional data
- path prediction
   based on past
   positions and
   weather

# **Activity Classification**

Achieved reasonable results with feature mapping and logistic regression with some dimensionality reduction.

# Path prediction

Further work is needed to resolve issues with the LSTM-based path prediction algorithm.



# **Activity Classification**

**Task:** Given 5 positions, determine class of gull movement.

Class 1: Minor to none

Class 1: Million to hone

Class 2: Minor, multi-directional

Class 3: Moderate, unidirectional Class 4: Significant, unidirectional

#### Data:

Labeled by a group member. Train data uses positions of 5 gulls. Test data uses positions of 2 gulls.

## **Feature maps:**

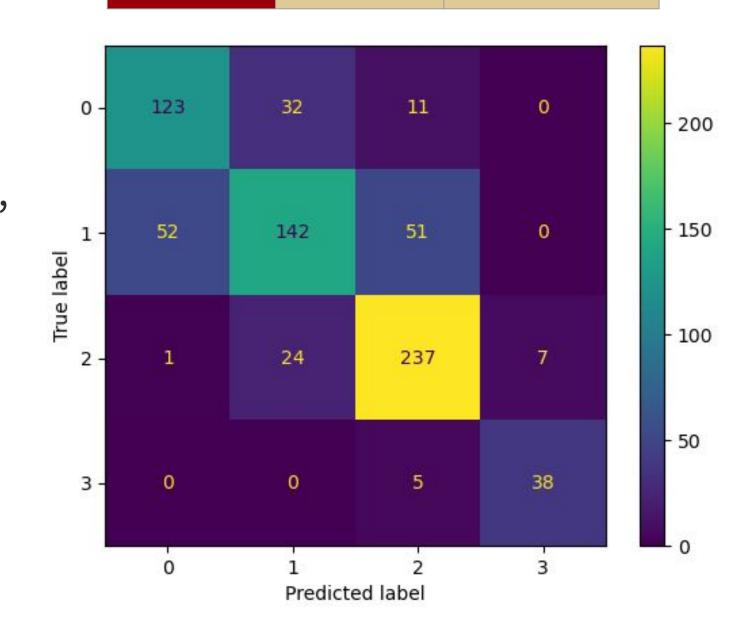
latitude, directions longitude standard deviations

Models Considered: Support Vector Machine (SVM), Naive Bayes, Logistic Regression

Dimensionality Reduction: After choosing best model, reduce feature dimensionality with principal component analysis (PCA).

Results: Logistic regression shows the best performance. Logistic regression maintains reasonable performance with number of dimensions reduced to 10 by PCA.

Algorithm	Average F1	Weighted Average F1
SVM, RBF	51%	46%
SVM, poly	59%	55%
SVM, sigmoid	36%	32%
Naive Bayes	67%	62%
Logistic (PCA=19)	77%	75%
Logistic (PCA=9)	66%	66%
Logistic (PCA=10)	76%	74%



Confusion matrix for logistic regression, PCA-reduced to 10 dimensions

# **Path Prediction**

**Task:** Given 5 prior locations of the gull as well as weather data across the geographical area at the same respective times, predict the next position of the gull.

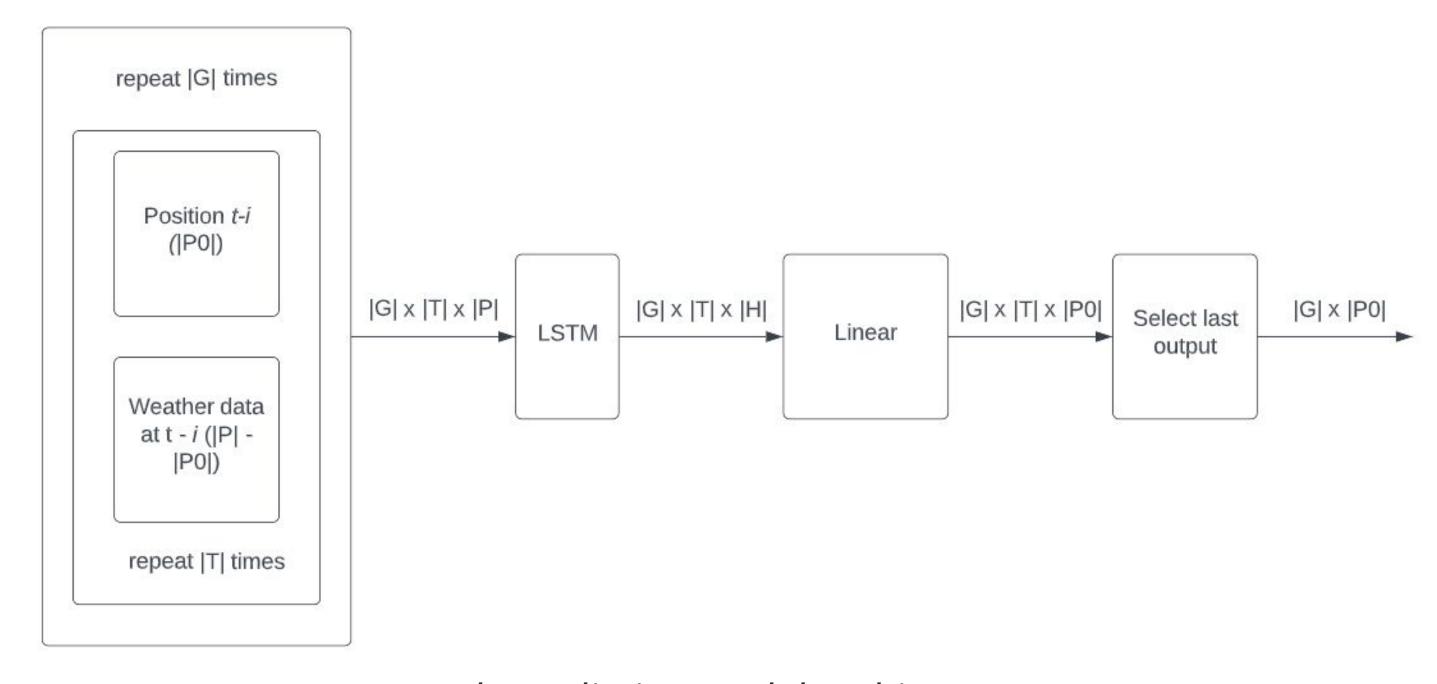
#### Data:

*Train data*: first 70% of times.

Test data: Last 25% of

times.

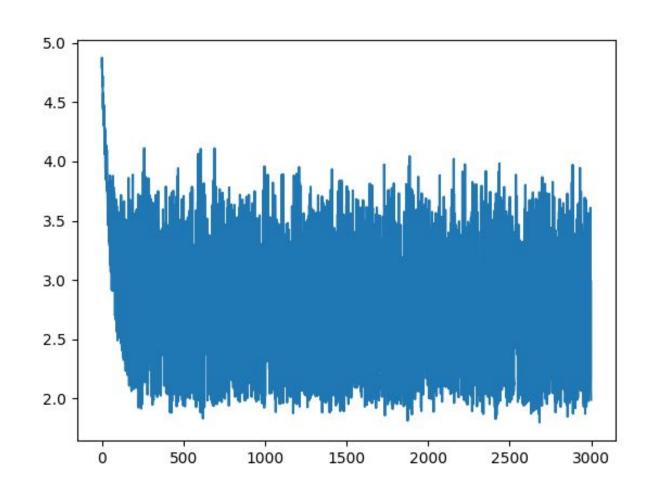
Model: Neural network with long short-term memory and linear layers.



Path Prediction Model Architecture

## **Performance**

Model failed to converge on train data and yielded repeated cross-entropy values in the range of 3.3-3.5 on test data.



Training loss as a function of epoch