What's the Catch? Recreational Fishing Trends in North Carolina (1990-2019)

 $https://github.com/ardathdixon/Data_FinalProject$

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1 Rationale and Research Questions

As public awareness of increasing strains on ocean resources and organisms grows (e.g. see the reach of films such as Seaspiracy and Sonic Sea), more attention is being placed on understanding fishing patterns and the impacts they have on the oceans. Most of this attention is placed on commercial and industrial fishing operations, which are studied and managed by the federal agency NOAA, the National Oceanic and Atmospheric Administration. However, there are fewer studies on recreational fishing, for which NOAA also aids in overseeing and collects data.

For this study, we chose to investigate recreational fishing trends in North Carolina over a thirty-year period. The data, whose origins are discussed more below, initially included 27 variables, detailing information such as mode of fishing and wave (2-month time period) the data was caught in. For this analysis, we wanted to look specifically at total catch during each wave, as we were running time series analyses during the course of the project. Therefore, we focused on only one catch estimation variable: total catch. We chose North Carolina due to our familiarity with species here, and we chose two popular recreational fishing species to investigate alongside all species combined. Trends in recreational fishing data can give us information about human behavior, species populations, and species movement patterns, which is why we found this topic interesting and wanted to investigate it further.

We chose the following questions to guide our work:

- 1. Are there trends in the amount of these fish caught over time? How do they compare?
- 2. What could these trends look like in the future?

2 Dataset Information

2.1 Data Retrieval:

For this analysis, we used data collected during Marine Recreational Information Program (MRIP) surveys conducted by NOAA (NOAA n.d.). NOAA works with state and local partners to collect information on the species and number of fish caught by fishers via in-person communication, telephone surveys, and mail-in surveys. We retrieved the data using the NOAA Recreational Fisheries Statistics Queries "download query" tool (found here). We created three separate queries to download data: one for all species, one for bluefish (*Pomatomus saltatrix*), and one for black sea bass (*Centropristis striata*). For each query, we used the date range 1990-2019 and requested catch estimate data by "wave", or two-month period, for all waves, fishing modes, and areas of fishing for the state of North Carolina (Table 1). We downloaded the CSVs as ZIP files and added them to our project repository. All data and code for this project can be retrieved from the **GitHub repository**.

Detail Description

Data Source NOAA MRIP

Retrieved from https://www.fisheries.noaa.gov/data-tools/recreational-fisheries-statistics-queries

Variables Used Year, Wave, Total Catch (Number of fish), Mode, Area

Date Range January 1990 - December 2019

Table 1: General Information About the Data Used

2.2 Data Wrangling:

We began our analysis by selecting the following columns from the raw data CSV files: Year, Wave, Mode, Area, and Total Catch. Next, we created a custom function to convert waves to months in order to process the six annual waves using time series analysis. For NOAA fishing records, wave 1 represents January and February, wave 2 represents March and April, and this continues through the year. Therefore, we assigned wave 1 catches to the date of January 1, wave 2 catches to March 1, and beyond.

After applying the custom wave-to-month function to the data, we created a date variable to capture year and month together (the day value for each of these dates was 01). Next, we used a split-apply-combine approach to find the sum of total catch for each wave. Each of these sums contains the catch from all unique combinations of fishing mode and area that were recorded during a given wave. Finally, to create the time series, we selected only the columns for date (DATE) and the sum of total catch (TOT_CAT_ALL). We computed summary statistics for these total catch values, as shown in Table 2. We then proceeded with exploratory analysis and interpolation of missing data points, which are described below.

Table 2: Total Catch Summaries (Number of Fish)

Summary Statistics	All Fish	Bluefish	Black Sea Bass
Minimum	11869	26	1168
Mean	12402954	1342064	411196
Median	11292146	1064369	313437
Maximum	34932698	5254124	1746847

3 Exploratory Analysis

Following initial wrangling, we checked the number of waves without catch records for each dataset by joining the existing data to a list of all possible waves between wave 1 of 1990 (represented by 1990-01-01) and wave 6 of 2019 (represented by 2019-11-01). The results of this exploration, which informed our approach for interpolation, can be found in Table 3.

Table 3: Number of missing values from NOAA MRIP data

Dataset	Number of missing values
All fish	11
Bluefish	17
Black sea bass	13

To fill the gaps with no data, we interpolated the likely values of missing time periods. This linear interpolation incorporated the catch numbers on either side chronologically. If wave 1 of 1990 or wave 6 of 2019 was missing, we did not interpolate the value, as there would not be a measurement available on both sides. We graphed the total catch trends over time (with the newly interpolated values for missing periods) as shown in Figure 1. With this visualization, we could compare the three categories' recreational fishing catch patterns: all fish, bluefish, and black sea bass.

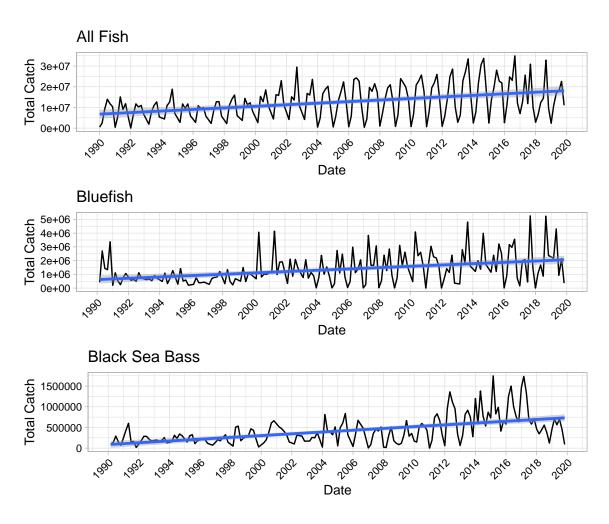


Figure 1: Total Catch Patterns over Time

4 Analysis

4.1 Question 1: Are there trends in the amount of these fish caught over time? How do they compare?

After our exploratory analysis and interpolation of the missing data points for each dataset, we created three time series for further analysis. We investigated the trends in total catch for all fish (Figure 2), bluefish (Figure 3), and black sea bass (Figure 4) by decomposing the time series into their seasonal, trend, and remainder components. For all three time series, we observed a strong seasonal trend with low catch totals in the winter and high catch totals in the summer. Furthermore, each trend component showed an apparent increase in total catch over time.

Though all three datasets show an increasing trend over time, the magnitude of this increase varies. The trend component of the time series has a range of 12,667,496 for all fish, 2,237,223.2 for bluefish, and 1,115,112.3 for black sea bass. The IQR for the trend component of the time series was 5,528,819 for all fish, 976,185 for bluefish, and 245,287 for black sea bass.

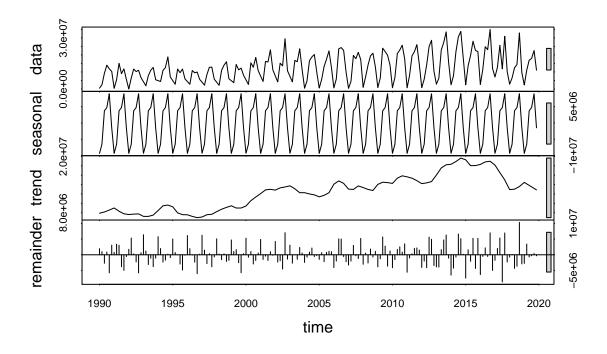


Figure 2: Seasonal and Trend Decomposition for All Fish Total Catch

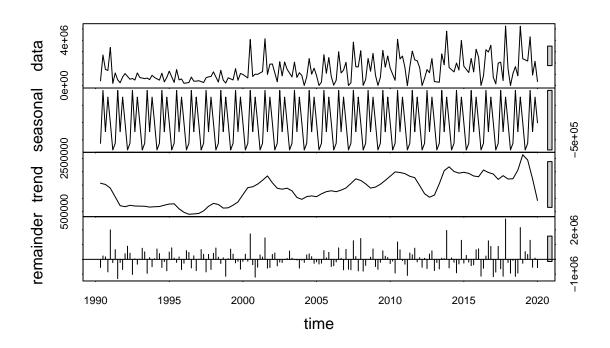


Figure 3: Seasonal and Trend Decomposition for Bluefish Total Catch

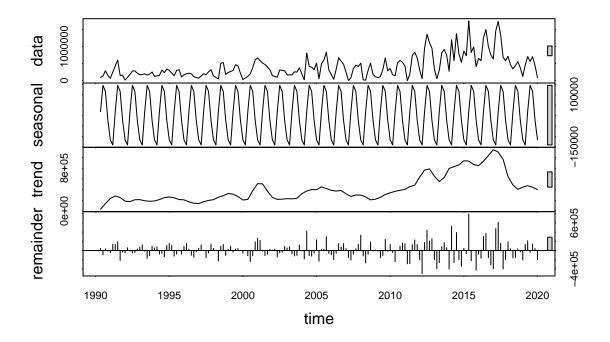


Figure 4: Seasonal and Trend Decomposition for Black Sea Bass Total Catch

We ran a Seasonal Mann-Kendall test on each time series to test whether there was a monotonic trend in the total number of fish caught over time (Table 4). All three tests returned a statistically significant result (all fish: tau = 0.49, $p < 2.22 \times 10^{-16}$; bluefish: tau = 0.32, $p = 8.75 \times 10^{-10}$; black sea bass: tau = 0.41, $p = 8.44 \times 10^{-15}$). Therefore, for all three time series, we reject the null hypothesis that there is no monotonic trend in favor of the alternative hypothesis that there is a trend in the data over time.

Table 4: Seasonal Mann Kendall Tests

Fish Category	tau	2-Sided P-value
All Fish	0.4896552	0.000000e+00
Bluefish	0.3235180	8.748902e-10
Black Sea Bass	0.4095312	8.437695e-15

Note on Table 4: p-values in this table were generated from wrangling the outputs from the seasonal Mann-Kendall tests. This wrangling rounded the p-value for all fish from < 2.22 x 10^-16 to 0.

The strengths of these trends vary; based on tau values, catch for all fish has the strongest overall trend while bluefish has the weakest overall trend. Nonetheless, total catch for both

individual species and all species combined increases between the beginning of 1990 and the end of 2019.

4.2 Question 2: What could these trends look like in the future?

To investigate future trends, we used forecasting in R to predict future data based on the existing past data we pulled from NOAA. There are several different methods through which to forecast, though only some of them account for seasonality, which was necessary here due to the seasonal components found in all three of our time series. We chose to use the Holt-Winters forecasting method for this data. Holt-Winters is more complex than simpler methods like naive forecasting, but it requires knowledge of fewer additional input variables that other models, like SARIMA, need in order to run. Holt-Winters uses smooth exponentiating and varying weights of past data - with more recent data weighed more - to predict future data. Here, we predicted five years of data (where h = number of periods, and each period = two months). In the resulting plots, the dark blue area represents the 80% confidence interval level for predicted data, while the light blue area represents the 95% confidence interval level.

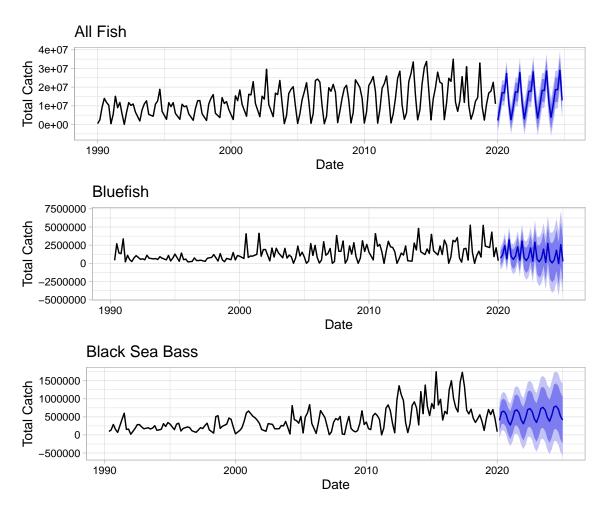


Figure 5: Holt-Winters Catch Forecasts

The Holt-Winters plots (Figure 5) all show clear continued seasonal patterns and trends. For black sea bass and all species combined, overall future trends are expected to be positive, like the past trends. For bluefish, the forecasted trend is slightly negative – this is likely because the most recent years of bluefish data do show a slight negative slope, even though the trend is positive for the full thirty years in the dataset (though as noted above, bluefish has the smallest tau, and therefore the weakest trend). These forecasting plots are useful visualizations but should not be considered fully accurate because of the inherent complexities in fishing catch data that models like Holt-Winters cannot account for.

5 Summary and Conclusions

5.1 Strong seasonal trends

NOAA marine recreational fishing catch totals for North Carolina show strong seasonal trends. Many more fish are caught in the summer, and much fewer fish are caught in the winter, as demonstrated above (Figure 1). This seasonality is likely influenced by recreational fishing patterns, where fishers are more likely to fish in warm summer weather than cool winter weather. Another potential cause for the seasonal trends is fish abundance and migration patterns, with higher populations of fish in North Carolina waters during the summer than during the winter. Total catch trends for all fish and black sea bass showed unimodal peaks and valleys overall, while bluefish showed bimodal trends (Figure 1). These bimodal bluefish peaks could be due to their seasonal migration patterns (ASMFC 2021). Another possible explanation for their bimodal trends is that bluefish in the northern Atlantic Ocean spawn twice annually, once in the spring and once in the summer (Arthur & Walford 1979).

5.2 Overall positive trend

There was an increase in total catch of bluefish, black sea bass, and all species combined over time. The driver of this increase in recreational fishery landings is unknown, but it could be attributed to increased fishing effort, with more fishers participating in recreational fishing or individual fishers catching more fish.

Although the trend was generally positive for all three time series, it was not uniform; in other words, the rate of change varied over time, and there were brief periods where catch plateaued or decreased. This variation could be caused by changes in recreational fishing regulations over time, such as the increase or decrease of catch limits and size limits, or temporary area closures. Furthermore, recreational fishing is subject to the myriad of factors that can influence human behavior, including but not limited to climate variation (Dundas & von Haefen 2020) and events with an environmental impact such as hurricanes (Smee et al. 2020) or the Deepwater Horizon Oil Spill (Alvarez et al. 2014). In a recent example of how other current events could impact recreational fishing that is beyond the scope of our analysis extending only to the end of 2019, the COVID-19 pandemic could have either increased recreational fishing in 2020, when people were spending more time outside, or decreased recreational fishing, as people were less able to travel to the coast.

5.3 Limitations

The MRIP system works very well for its intended purposes, but it is ultimately still a system based on estimation. Through MRIP, NOAA interviews only some fishers, then uses mathematical modeling to extrapolate on this collected survey data in order to create statewide (or area-wide, depending on the dataset) estimates. While MRIP is the best source of recreational fishing catch data, there is always room for some error in its estimations.

Within each of our datasets, there were some missing values. These values were not side by side or frequent, so we chose to linearly interpolate our data in order to fill them in. This

helped our figures and analyses (e.g. seasonal Mann-Kendall) to appear and run cleaner, but interpolating has some drawbacks. While interpolations seemed to follow the clear seasonal pattern in each dataset, most often the interpolated data was in wave periods that were minimums in other years, so our interpolated values may have been a bit higher than actual catch rates during those times. Interpolation is an estimator for missing data, and it is important to acknowledge that our interpolated data may not be representative of true values.

Finally, our forecasted data was much cleaner and less noisy than our existing data (see Figure 5). Noise in these datasets comes from external factors like changes in fish catch limits or weather patterns, which this forecasting method can neither take into account nor predict. Our forecasting outputs were therefore limited by the relatively simple methodology we chose to employ. Holt-Winters remains a popular forecasting method and is reasonable to use for our purposes here, but the inherent uncertainty in this predicted data is also important to acknowledge.

5.4 Future recommendations

Future research could extend our analysis to include a greater scope regarding specific species, areas, and timelines. NOAA provides catch data for an extensive number of fish species, and therefore time-series and forecasting analyses could also be applied to other species of interest. Likewise, the widespread area available for NOAA catch data means that future research could translate our analysis toward catch numbers from other states. Additionally, the available data goes beyond the timeline we analyzed, dating back to 1981. Future analysis could adjust variables of interest accordingly thanks to the vast NOAA data available for public access.

In addition to adjusting the sample for analysis, future research could compare changes of catch per unit of effort over time. Analyzing total catch does not incorporate the number of fishers or success rates. Incorporating the catch per unit of effort could reveal possible causes of the increasing catch trends over time (e.g. more fishers in total or more efficient fishers).

6 References

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