Ecological time series for the Celtic Sea

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Objectives:

- Collate existing data resources for the Celtic Sea;
- Tidy, filter and remove duplicates;
- Understand relationships between biological data and environment;
- Develop models for identifying synchronous changes across multiple species/taxa, detecting step changes and potential regime shifts

Questions:

- 1. Have there been gradual or sudden changes in abundance or life history traits (e.g. growth) of selected species?
- 2.Has the occurence of unusual events increased over time?
- 3.Are such trends related to changes in environment (e.g. wind, circulation patterns and temperature)?
- 4.Are there synchronous changes across multiple species?

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The eco audit

Are jellyfish going to take over the oceans?

Like a karmic device come to punish our planetary transgressions, jellyfish thrive on the environmental chaos humans create. Is the age of the jellyfish upon us?



Friday 21 August 2015 15.29 BST

email



Barrel Jellyfish at Kynance Cove in Cornwall, UK. Photograph: Andrew Pearson/Alamy

Most popular



North Korea claims successful nuclear test using hydrogen bomb -



North Korean nuclear test confirmed in major escalation by Kim Jong-

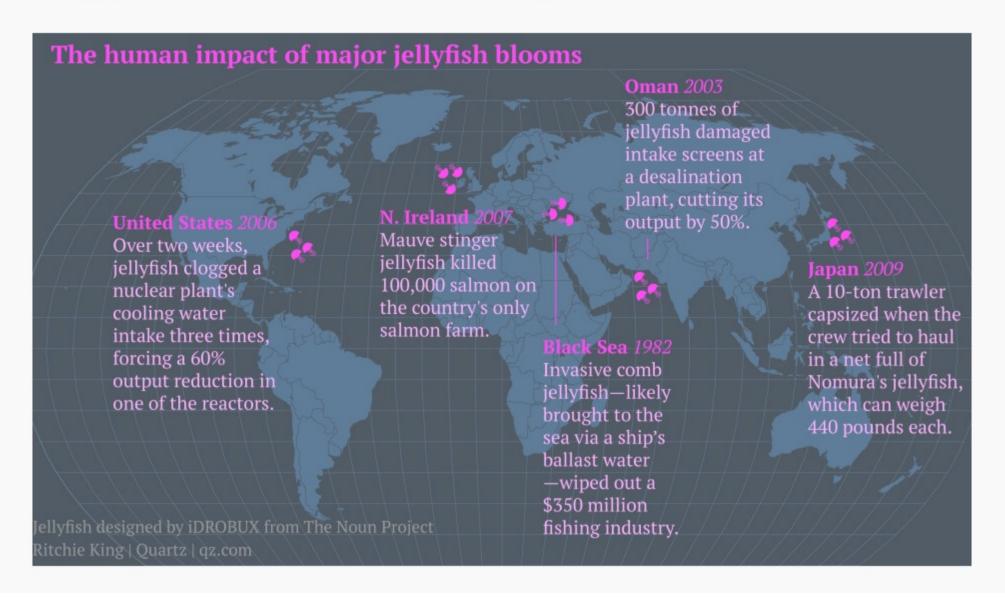


This North Korean nuclear test means Trump must now start talks



Fabio Fognini thrown out of HS Open over

Jellyfish are taking over the seas, and it might be too late to stop them



Why this is happening?

"They've got this unique life cycle where they can tolerate harsh conditions and then rapidly thrive when conditions are favourable. So when a stressor like climate change or overfishing opens up a niche for them they can really take advantage of that and rapidly proliferate," said Lucas Brotz, a researcher at the University of British Columbia. Not all species of jelly benefit, rather there tends to be a reduction in the diversity of species and vast, homogenous masses emerge.

"They can make millions and millions of copies of themselves and clone asexually. That's when you get these massive blooms. I think that's the secret to the success of jellyfish, the reason they've been around for hundreds of millions of years."

But whether there is <u>strong evidence of a global increase in jellyfish</u> populations is difficult to answer.

Some believe the current observed rise may represent a natural cycle.

Data description

Biological:

- Jellyfish blooms (occurrence, 1890 2016);
- Loggerhead turtle strandings (abundance, 1990 2015)
- Cape Clear Bird Observatory:
 - Basking sharks sightings (abundance, 1971 2008)
 - **Sunfish sightings** (abundance, 1971 2008)
 - Leatherback turtle sightings (abundance, 1971 2008)

Data provided by Thomas Doyle and Aidan Long from NUIG

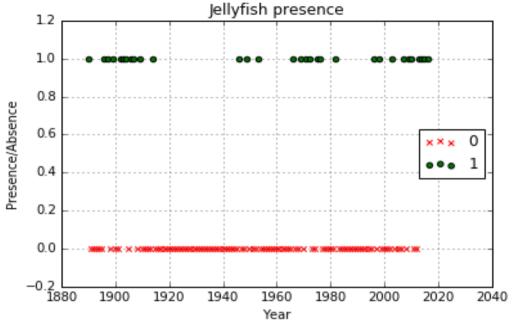
- Phytoplankton and jellyfish data (Continuous Plankton Recorder)
- Zooplankton (Station L4, Plymouth Marine Laboratory)

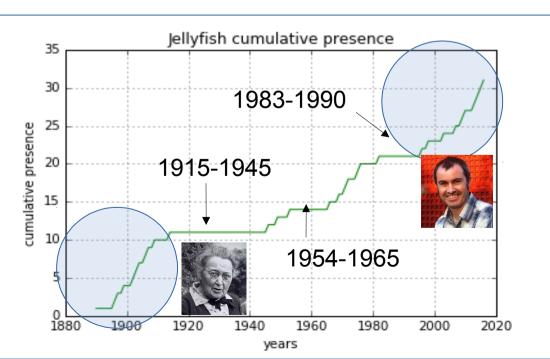
Environmental:

- Wind data (Cork Airport station, hourly 1962 2016)
- Sea Surface Temperature (monthly, 1854 2017)



Jellyfish blooms





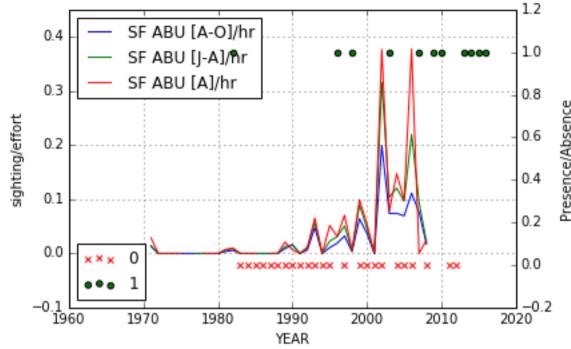


www.wikipedia.org

Data compiled from soft reports:
Irish Naturalist Journal, Irish Times,
Glaucus, Scottish Naturalist,
Marine Fauna
Effort not quantified and inconsistent
over time

Two comparable periods: indicates that current swarms similar to historical evidence

Sunfish sightings

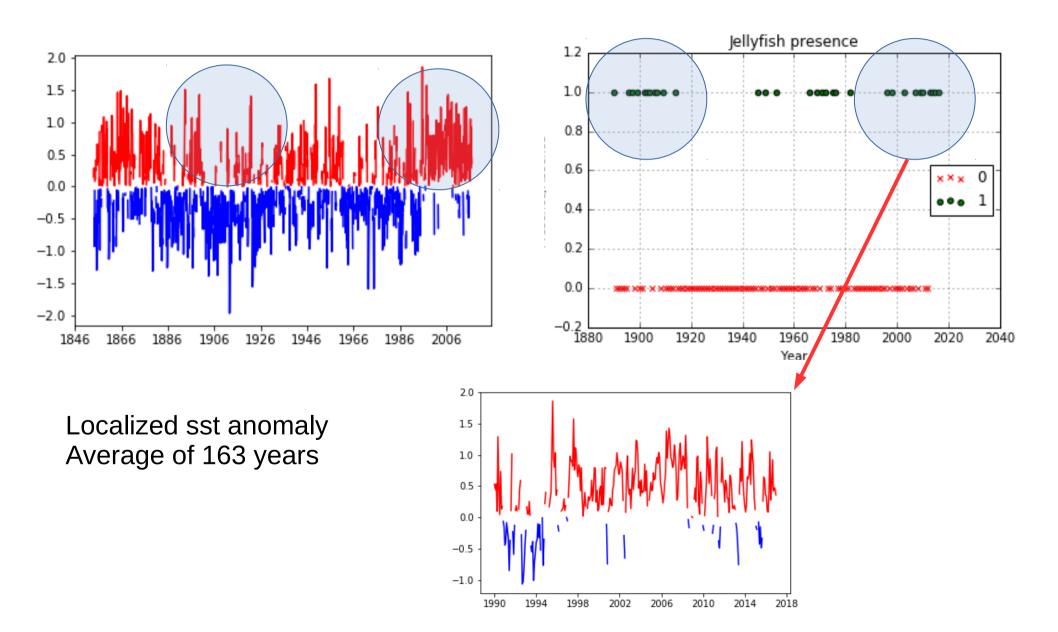


- Jellyfish predator abundance may correlate with jellyfish swarms
- Data shows a clear increase in sunfish sightings from 1999 onwards



Source: www.montereybayaquarium.org

Examining correlations with temperature

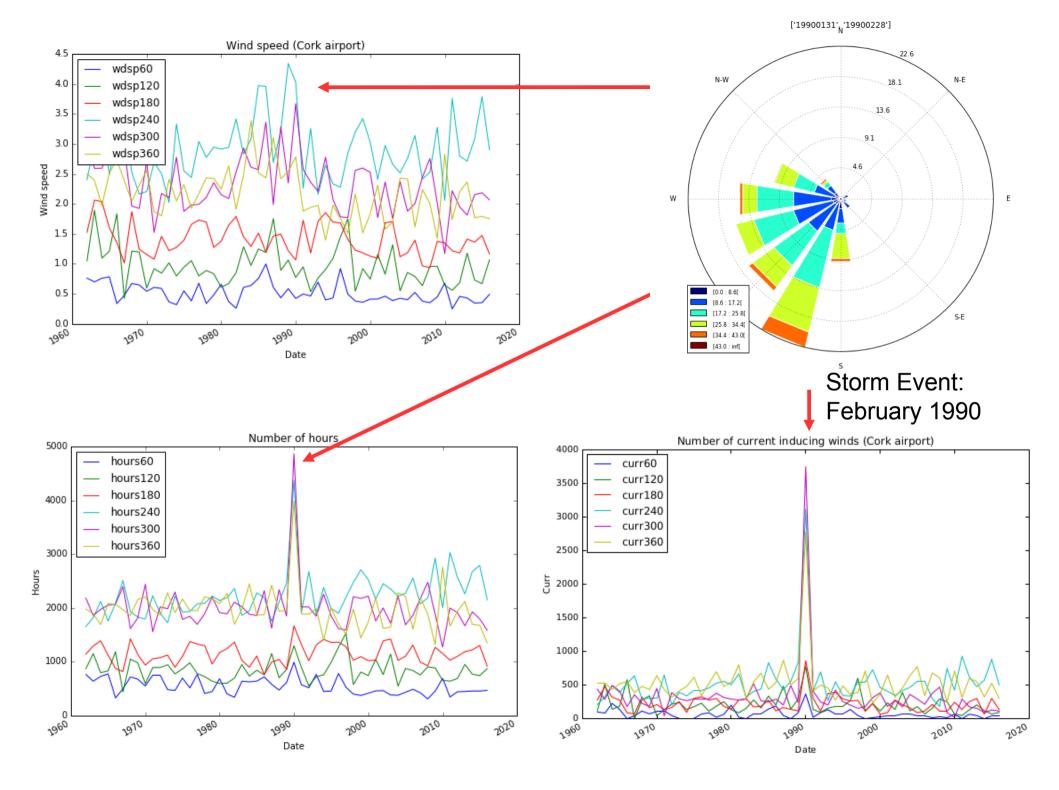


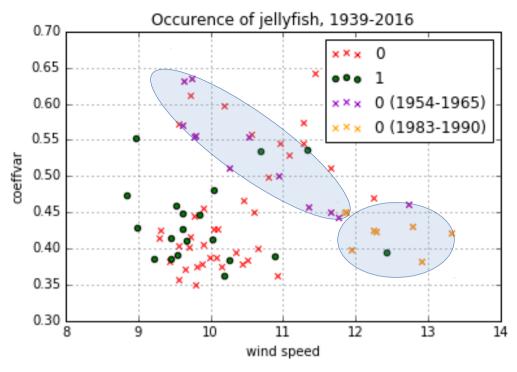
Wind reshape tool

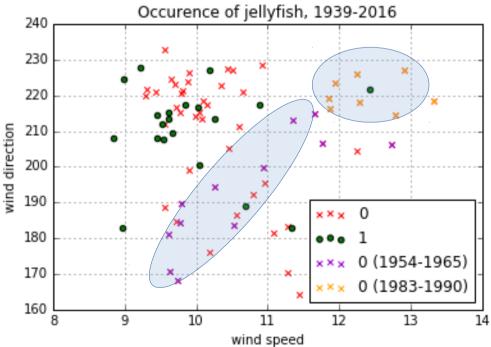
```
In [90]: df.head()
                                                                Out[90]:
def split(c):
                                                                       presence
    if c['wddir'] in range(0,60):
                                                                year
        val = 1
                                                                1982
                                                                              1 10.884018
                                                                                           217,248858
                                                                                                       0.259475
                                                                                                                 0.851256
                                                                                                                           1.789612
                                                                1983
                                                                                11.877283
                                                                                           216.304795
                                                                                                       0.605936
                                                                                                                 1.283790
                                                                                                                           1.448630
    else:
                                                                1984
                                                                               12,249431
                                                                                           225.826503
                                                                                                                 0.973588
                                                                                                       0.637637
                                                                                                                           1.289959
        val = 0
                                                                1985
                                                                                           214.503425
                                                                                                       0.754680
                                                                                                                 1.247831 1.510845
    return val
                                                                1986
                                                                                13.323858
                                                                                           218.509132 0.995776 1.188470 1.193607
d['wdsp60'] = d.apply(split, axis=1)*d.wdsp
                                                                                 wdsp360
                                                                                                              curr300 curr360 hours60 \
                                                                        wdsp300
                                                                                          curr60
                                                                year
def split(c):
                                                                      2.531279 1.908790
                                                                                                                293.0
                                                                                                                         297.0
                                                                                                                                  334.0
                                                                1982
                                                                                              0.0
    if c['wddir'] in range(60,120):
                                                                      2.930023
                                                                                2.488128
                                                                                             64.0
                                                                                                                520.0
                                                                                                                         545.0
                                                                                                                                  633.0
        val = 1
                                                                      2.611339 3.387978
                                                                                             68.0
                                                                                                                268.0
                                                                                                                         678.0
                                                                                                                                  622.0
    else:
                                                                      2.570776 2.517009
                                                                                                                         436.0
                                                                                                                                  635.0
                                                                                            144.0
                                                                                                                248.0
        val = 0
                                                                      3.362100
                                                                                            189.0
                                                                                                                         536.0
                                                                                                                                  714.0
                                                                                2.434247
                                                                                                                291.0
    return val
                                                                                                              hours360 coeffvar
                                                                       hours120 hours180
                                                                                          hours240
                                                                                                    hours300
                                                                                                                                         sst
d['wdsp120'] = d.apply(split, axis=1)*d.wdsp
                                                                year
                                                                 1982
                                                                         691.0
                                                                                  1362.0
                                                                                             2358.0
                                                                                                       2106.0
                                                                                                                 1695.0
                                                                                                                         0.389622 12.033112
                                                                1983
                                                                         945.0
                                                                                  1022.0
                                                                                             1859.0
                                                                                                       2018.0
                                                                                                                 1975.0
                                                                                                                        0.450120 12.176555
                                                                1984
                                                                         727.0
                                                                                   893.0
                                                                                             1945.0
                                                                                                       1879.0
                                                                                                                 2445.0 0.425079
                                                                                                                                 12.191222
                                                                1985
                                                                         838.0
                                                                                  1103.0
                                                                                             2277.0
                                                                                                       1856.0
                                                                                                                 1865.0 0.429986 12.067223
                                                                1986
                                                                         753.0
                                                                                   755.0
                                                                                             2189.0
                                                                                                       2318.0
                                                                                                                 1871.0 0.422112 11.561333
                                                                 [5 rows x 23 columns]
```

Divide wind data into 6 categories according to direction (60 degrees in each) Calculate:

- the average speed of the wind from that direction
- the number of hours that the wind blew from that direction
- the total number of current inducing winds from that direction (winds greater than 5 knots for 20 hours or more).
- coefficient of variation standard deviation of wind direction relative to the mean







Clusters in years:

1954-1965 1983-1990

False zeros?

False zeros

Non-detection does not imply that event did not take place unless the probability of detection is 1.

Presence-absence with true zeros:

Consider a subset of jellyfish data where effort was reasonably consistent (i.e. 1890-1915, 1990-2016);

Presence-absence with imperfect detectability:

 Consider all jellyfish data allowing for zeros using pseudo-absences approach (MacKenzie et al 2002);

• Presence-only:

- Consider all data treating data as presence only, while discarding information on absences
- Generally not possible to calculate probabilities of presence, only the relative likelihood (Pearce et al 2006).

Statistical models for dealing with false zeros

- Generalized Linear Models (GLM)
 - in particular Zero Inflated Bernoulli or Poisson
- Artificial Neural Networks (ANN)
 - Good at non-linear modelling, requires pseudo-absences
- Maximum Entropy (Maxent)
 - Does not require pseudo-absences

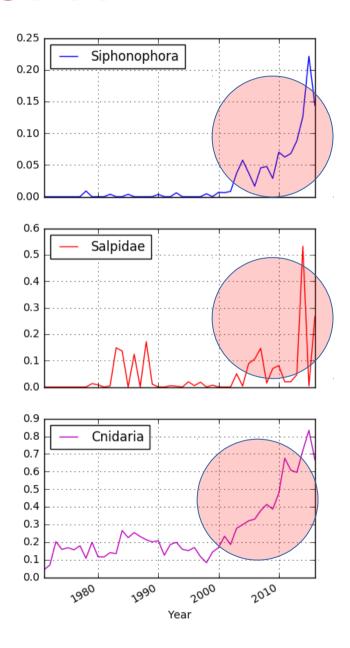
Sunfish Dataset

- Time series: 1971 2008 (2016)
- April October
- Count data, daily observations
- Zero-inflated semicontinuos data (95.81% zeros)
- Duration of observation, number of observers, sea state
- External variables: zooplankton, sst, jellies, phytoplankton, latitude of 13 degrees isotherm



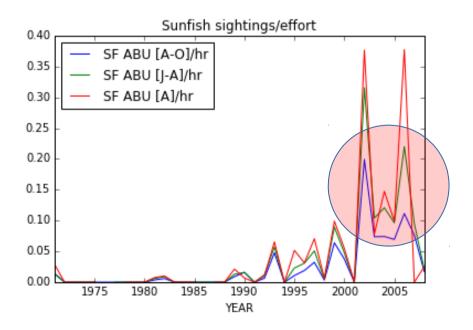
Source: Wikimedia Commons

CPR



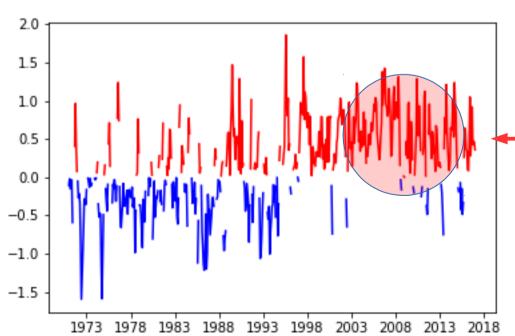


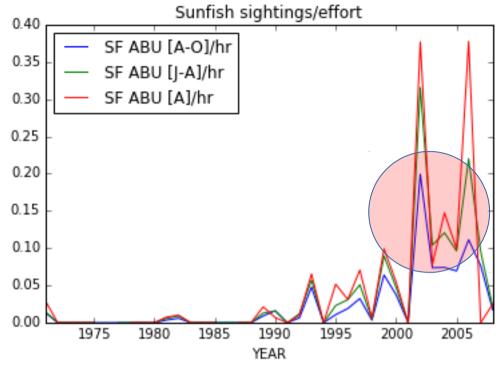
Source: Wikimedia Commons



SST anomaly

Meaningful anomaly or transient event?

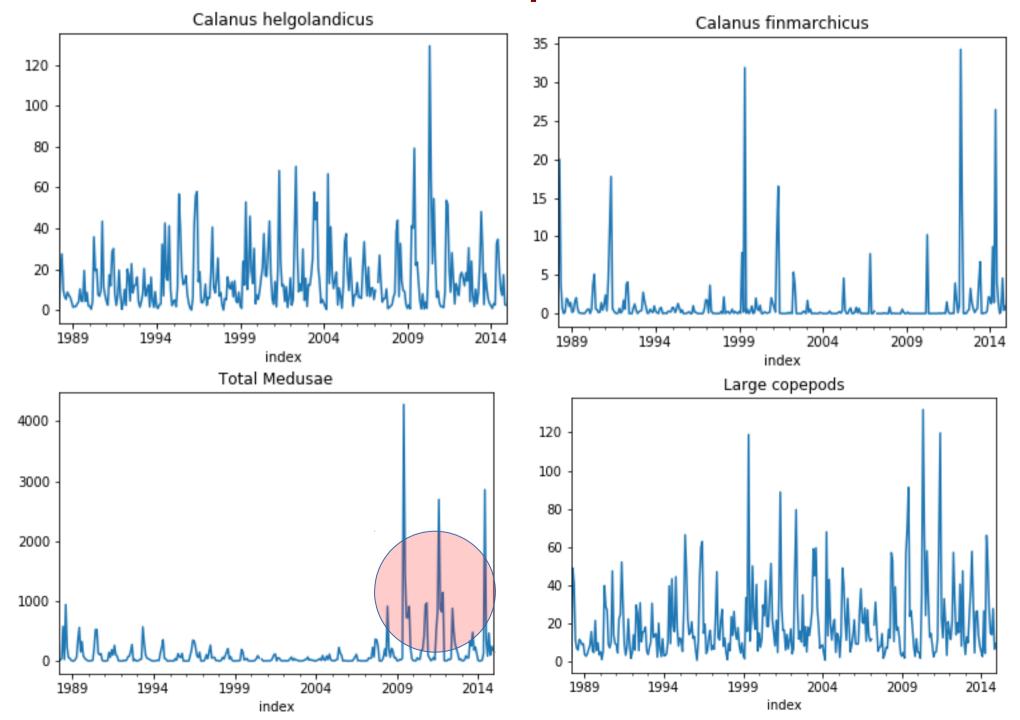




Temperatures are higher than average

How peaks in sunfish abundance relate to this?

Time series of zooplankton



Methods for dealing with zeroinflated data:

- Modelling the zero and non-zero data with one model and then modelling the truncated-at-zero data with another. This is often called a "hurdle model";
- Modelling the response variable as a mixture of a Bernoulli distribution (a point mass at zero) and a Poisson distribution (or any other count distribution supported on non-negative integers)

Hurdle model

- Hurdle models treats the zeros and non-zeros as two separate processes;
- Allow to model the zeros and non-zeros with different predictors or different roles of the same predictors

Hurdle model results (1):

Intercept only model:

Probability of non-zero value (logit model):

0.0418

Confidence intervals:

2.5 % 97.5 %

0.0355 0.0488

• Mean of distribution (log link **Gamma model**):

0.0118

Confidence intervals:

2.5 % 97.5 %

0.0097 0.0147

No better than the average value for the sample

Hurdle model results(2):

Full logit model:

```
Call:
alm(formula = non zero ~ totzoo + sst + totmedus + xmonth + totsipho +
    chel + lcop, family = binomial(link = logit), data = dat)
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.372e+01 1.417e+00 -9.677 < 2e-16
totzoo
           4.131e-05 1.153e-05 3.582 0.000341
sst
            1.017e+00 1.173e-01 8.666 < 2e-16 ***
totmedus
           -1.833e-03 1.050e-03 -1.745 0.080950
           -6.355e-01 1.510e-01 -4.209 2.56e-05
xmonth
          -8.509e-04 2.969e-04 -2.865 0.004164 **
totsipho
chel
           1.821e-01 4.823e-02 3.774 0.000160
lcop
           -1.756e-01 4.505e-02 -3.897 9.73e-05 ***
```

Probability of a positive sunfish record is positively influenced by environmental and biological variables (overall prob 1.104e-06)

Full Gamma model:

```
Call:
glm(formula = nsunfishm ~ seastate + xmonth + chel + lcop, family = Gamma(link = log),
    data = subset(dat, non_zero == 1))
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                       1.06714 -0.103
(Intercept) -0.11028
                                         0.9178
seastate
           -0.09432
                       0.05447 -1.732
                                         0.0856
xmonth
           -0.56580
                       0.13659
                               -4.142
                                          6e-05 ***
           -0.15060
                       0.06509 -2.314
                                         0.0222 *
chel
lcop
            0.13946
                       0.06084
                                2.292
                                         0.0234 *
```

Mean of distribution: 0.8956

Next steps:

- Further analyse sunfish sightings in relation to biological and environmental variables accounting for underlying trends;
- Detect annual trends that are over and above seasonal fluctuations that we would expect;
- Fit multispecies models to examine trends across species and detect potential step changes using state-space modelling