

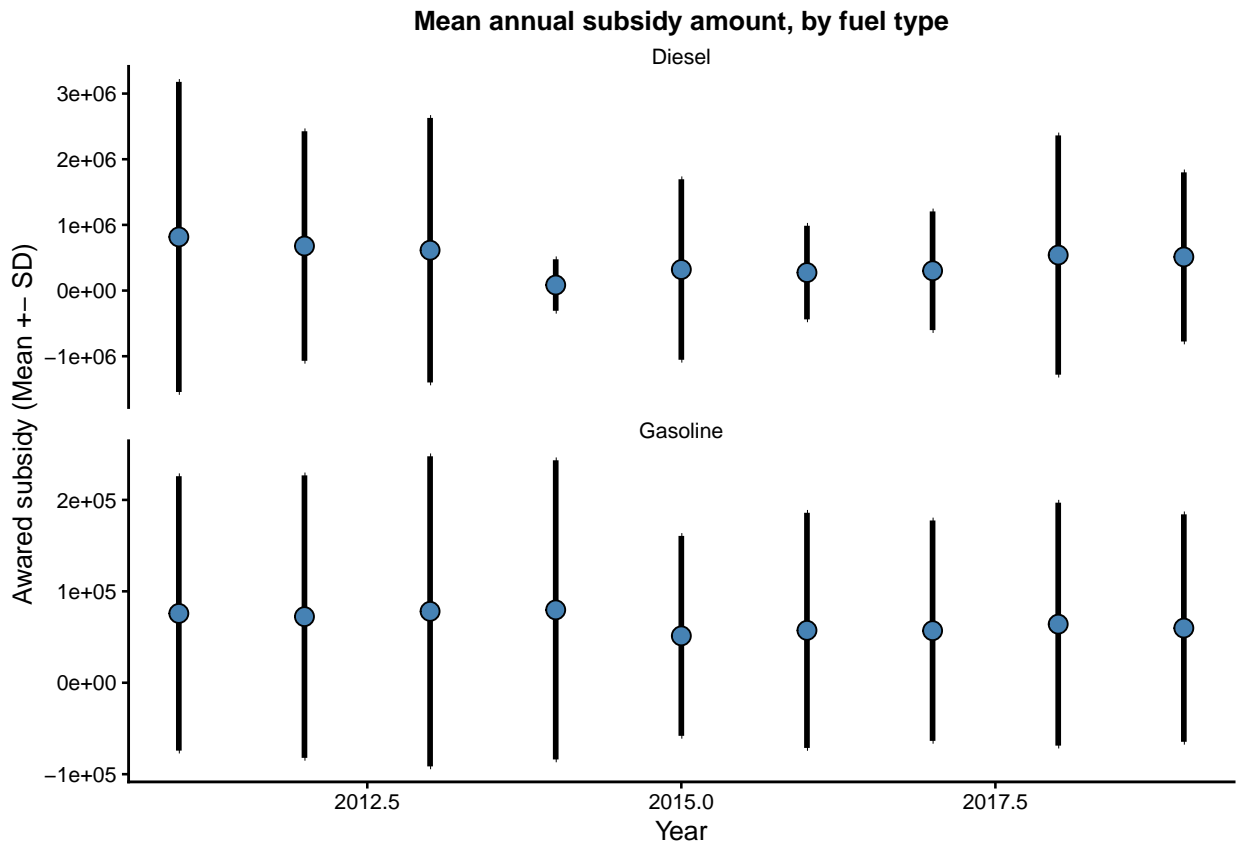
Summary stats on mexican subsidy, landings, and effort data

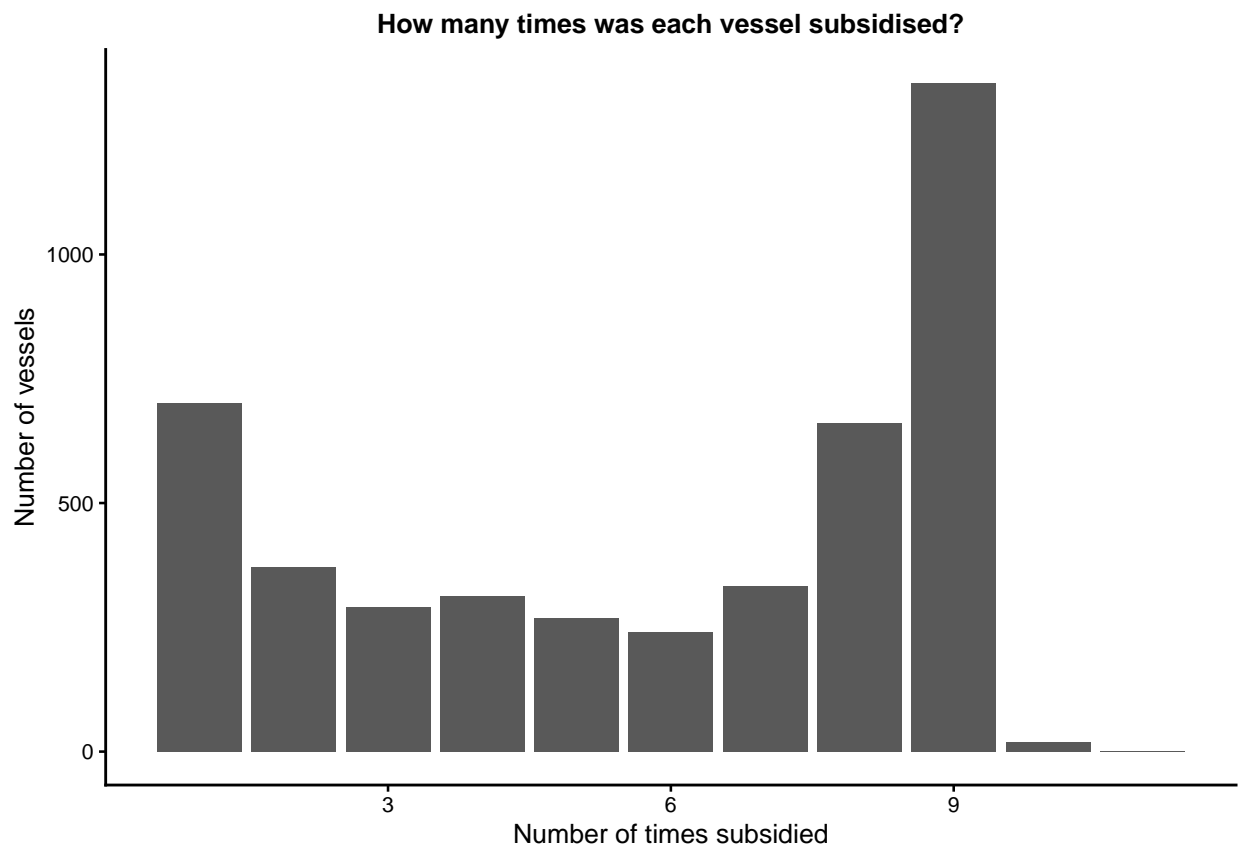
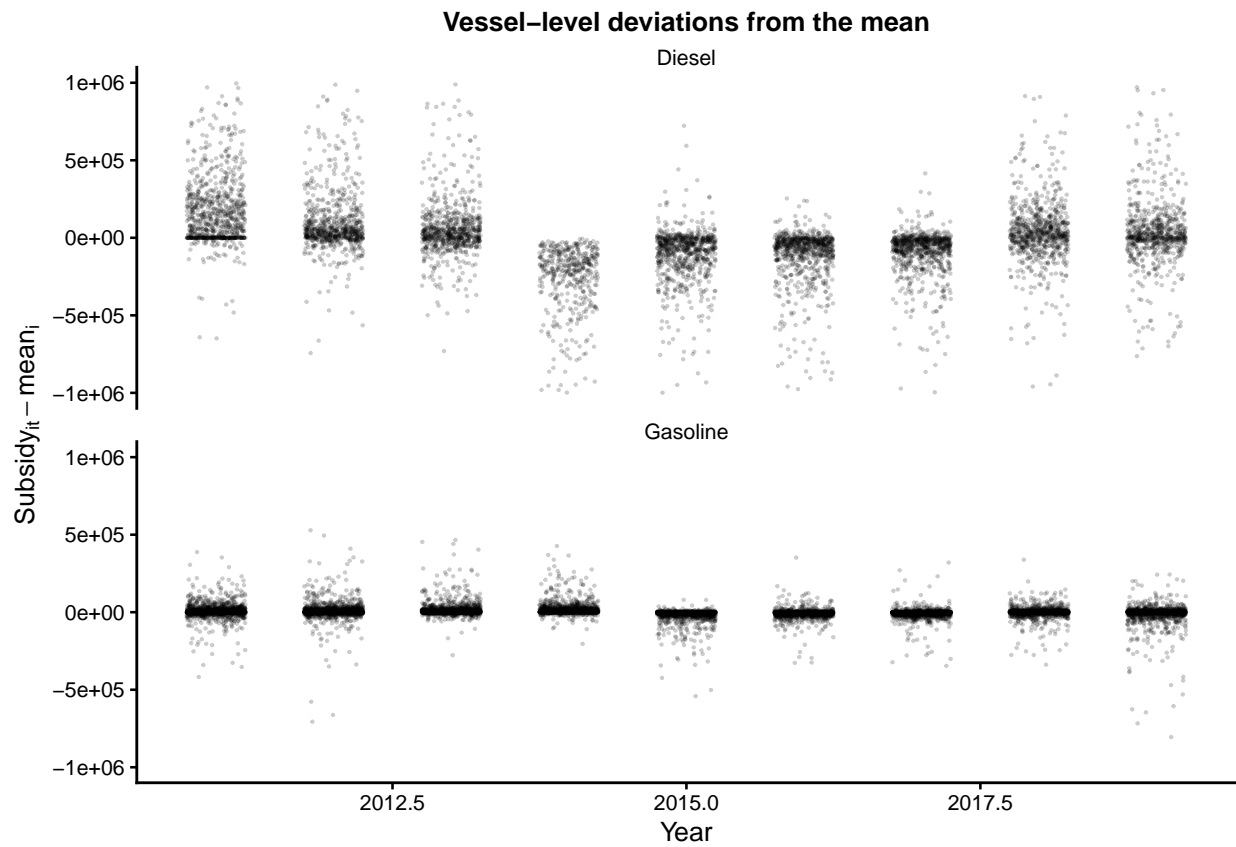
Disclaimer: I'm still getting to know and clean the data

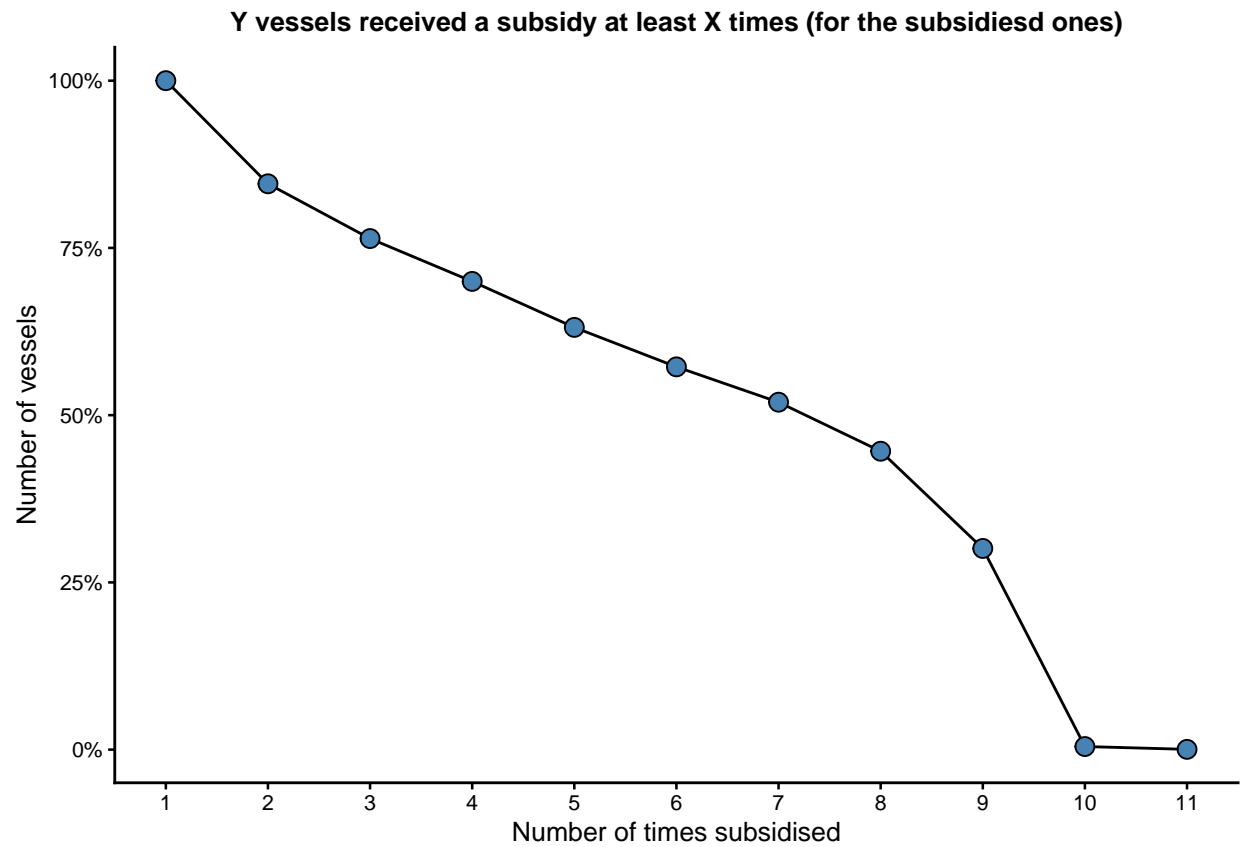
Juan Carlos Villaseñor-Derbez

Subsidy data

There are 4596 “Economic units” (cooperatives, individuals, companies) that received a fuel subsidy at least once between 2011 and 2019.

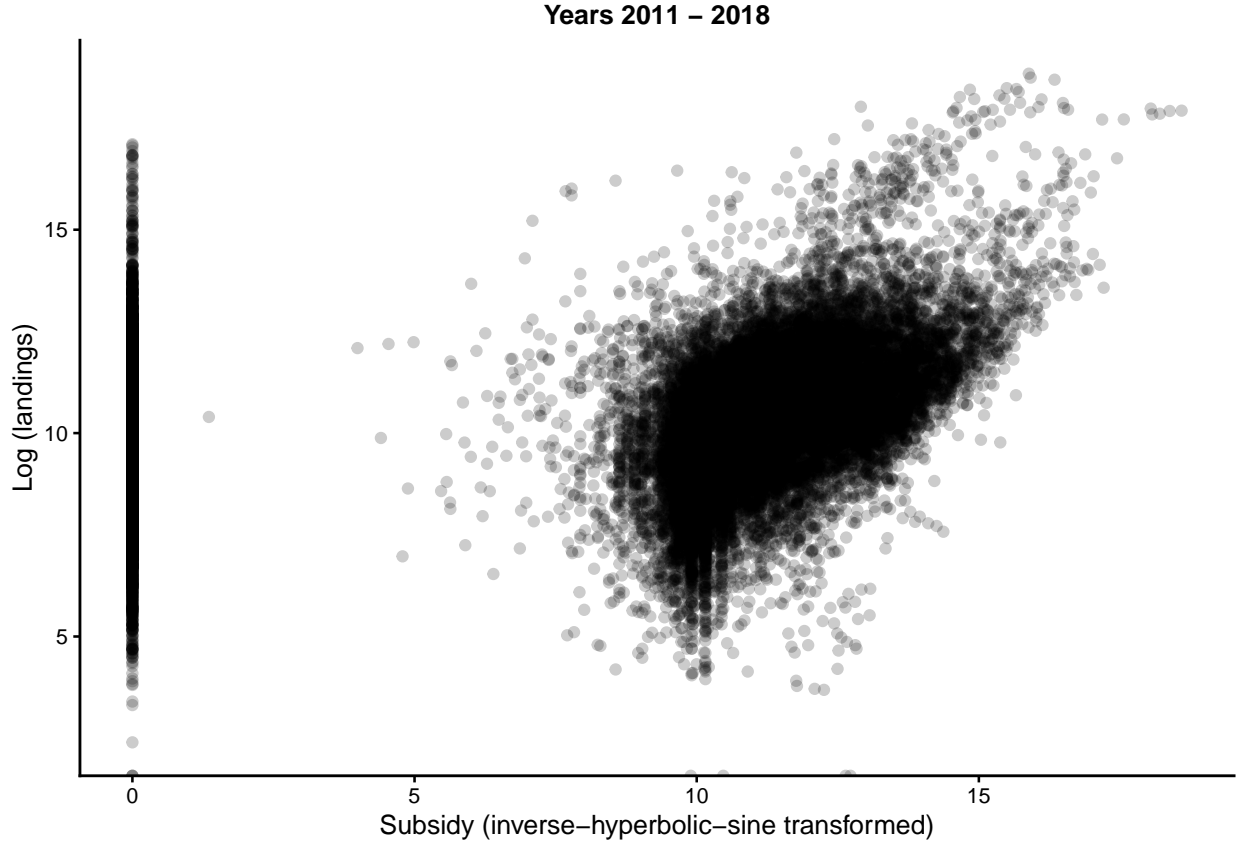






Landings data

A quick and dirty match says there are 27272 observations with a match and 14467 observations without a match. That is a match of 65.3393709%.



Some regressions

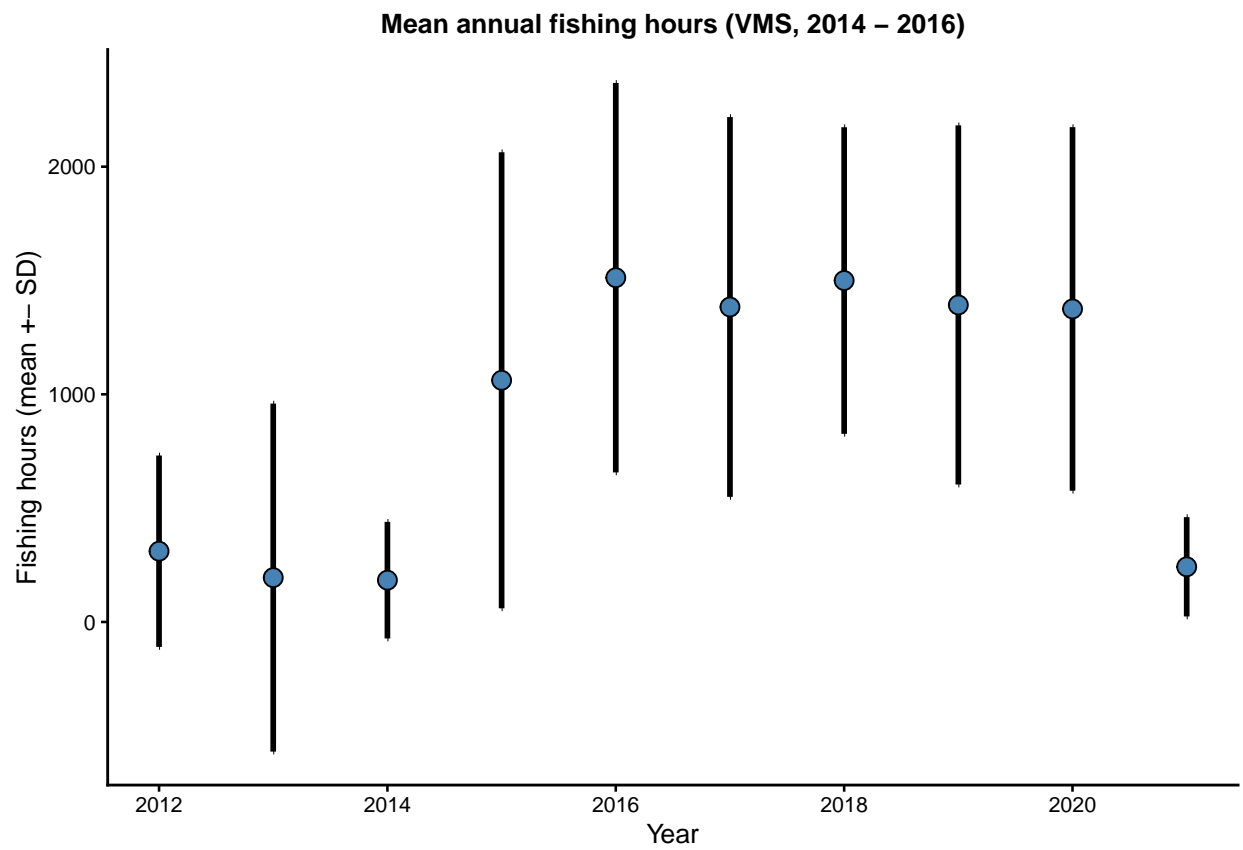
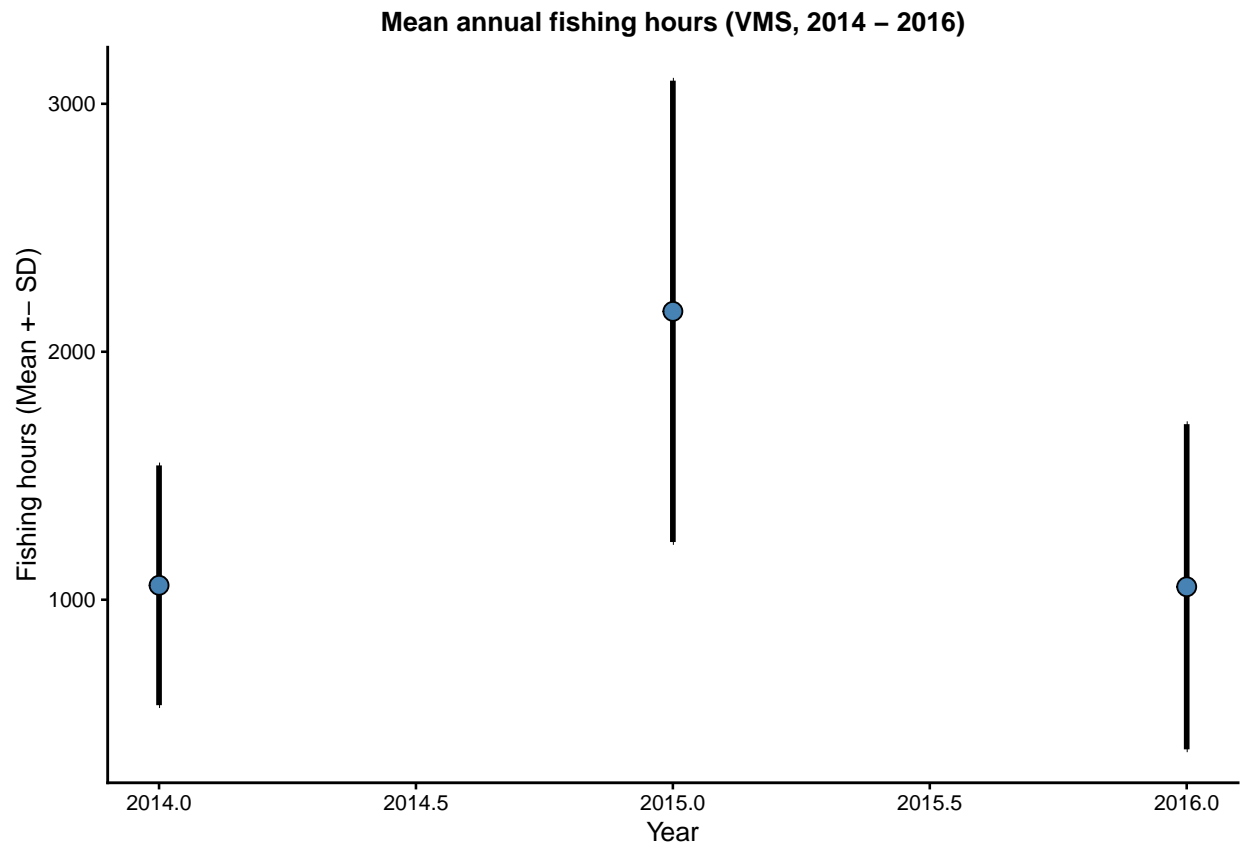
$$\log_{10}(L_{i,t}) = \alpha + \beta_1 \text{ih}(\text{subsidy}_{i,t}) + \gamma_t Y_t + \phi_i \text{RNPA} + \epsilon_{it}$$

Model 1	
hs(subsidy__amount)	0.004 (0.010)
fuel_typeGasoline	0.014 (0.021)
Num.Obs.	21227
R2	0.867
FE: as.factor(rnpa)	X
FE: as.factor(year)	X
Std. errors	Clustered (as.factor(year))

* p < 0.1, ** p < 0.05, *** p < 0.01

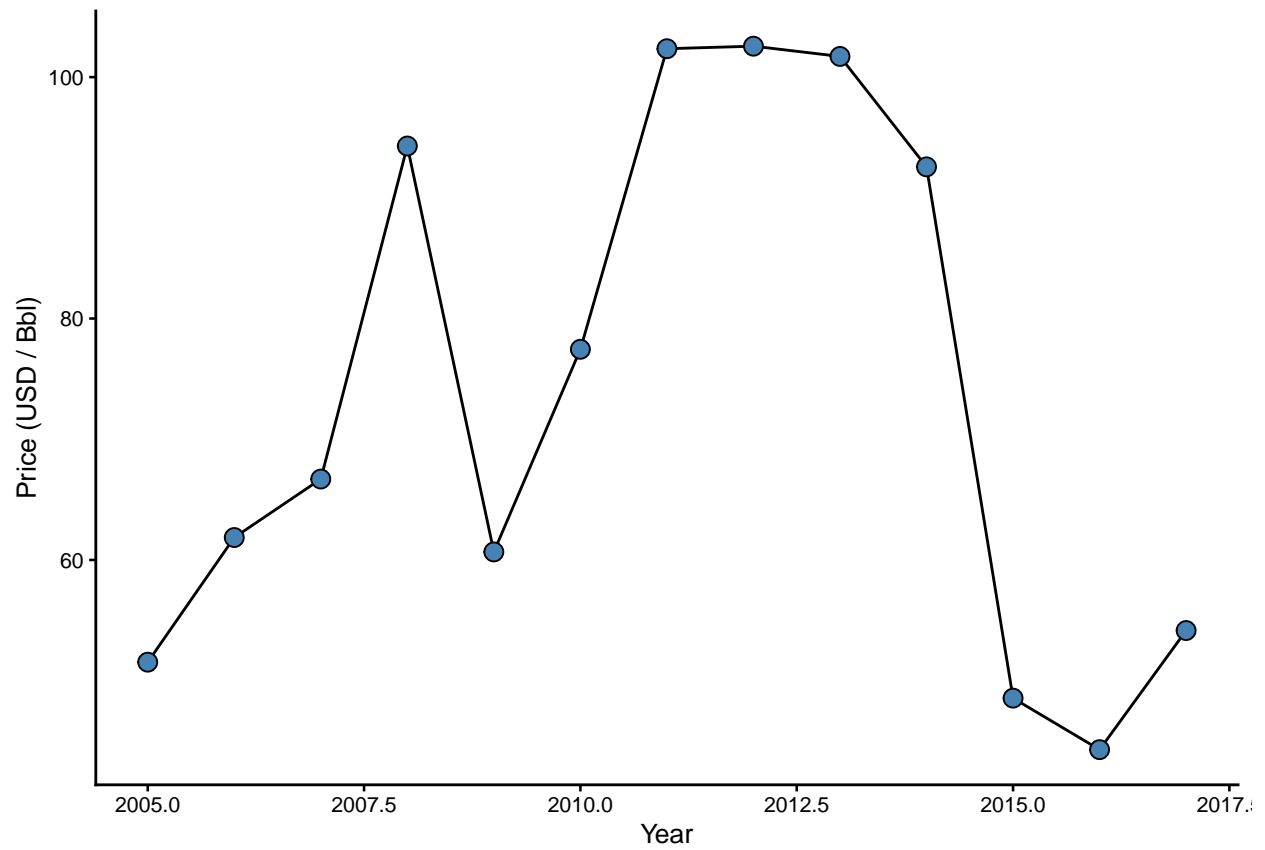
Effort data

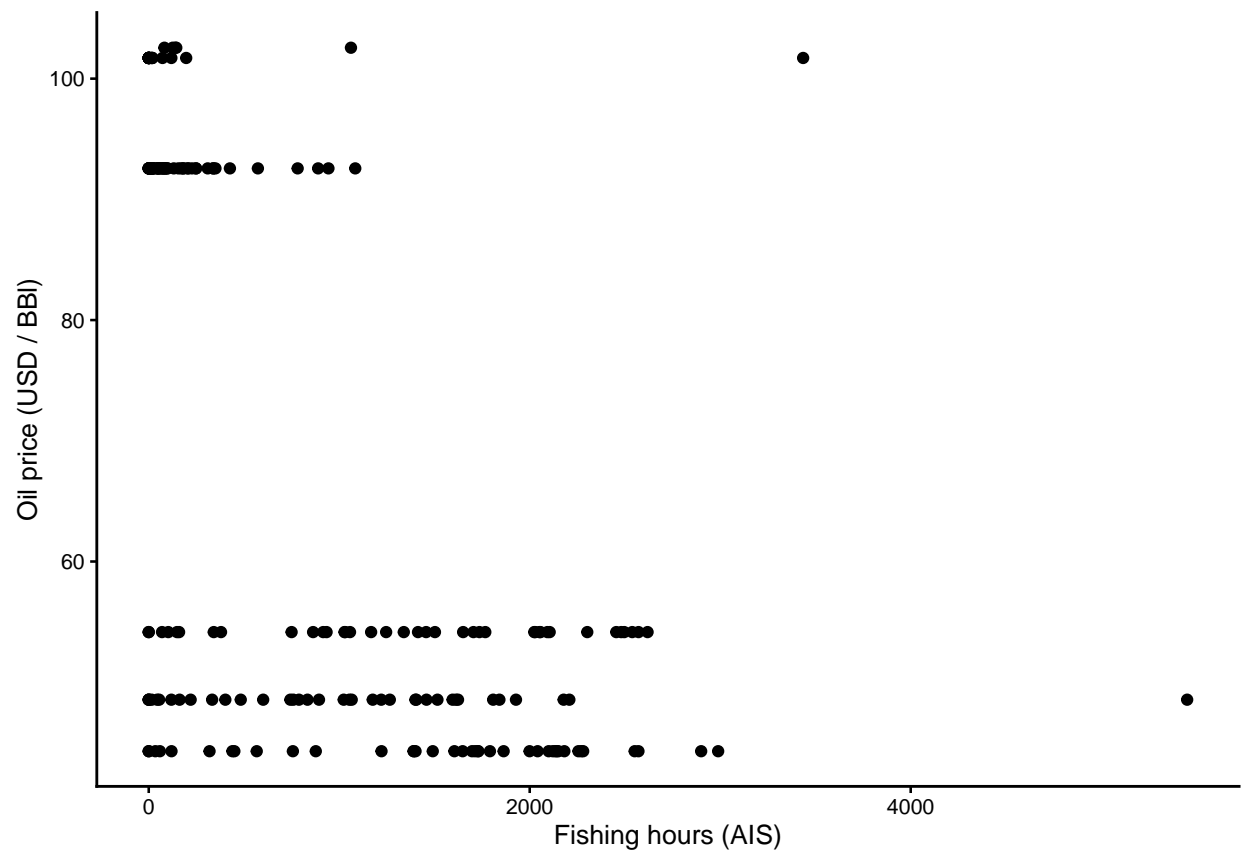
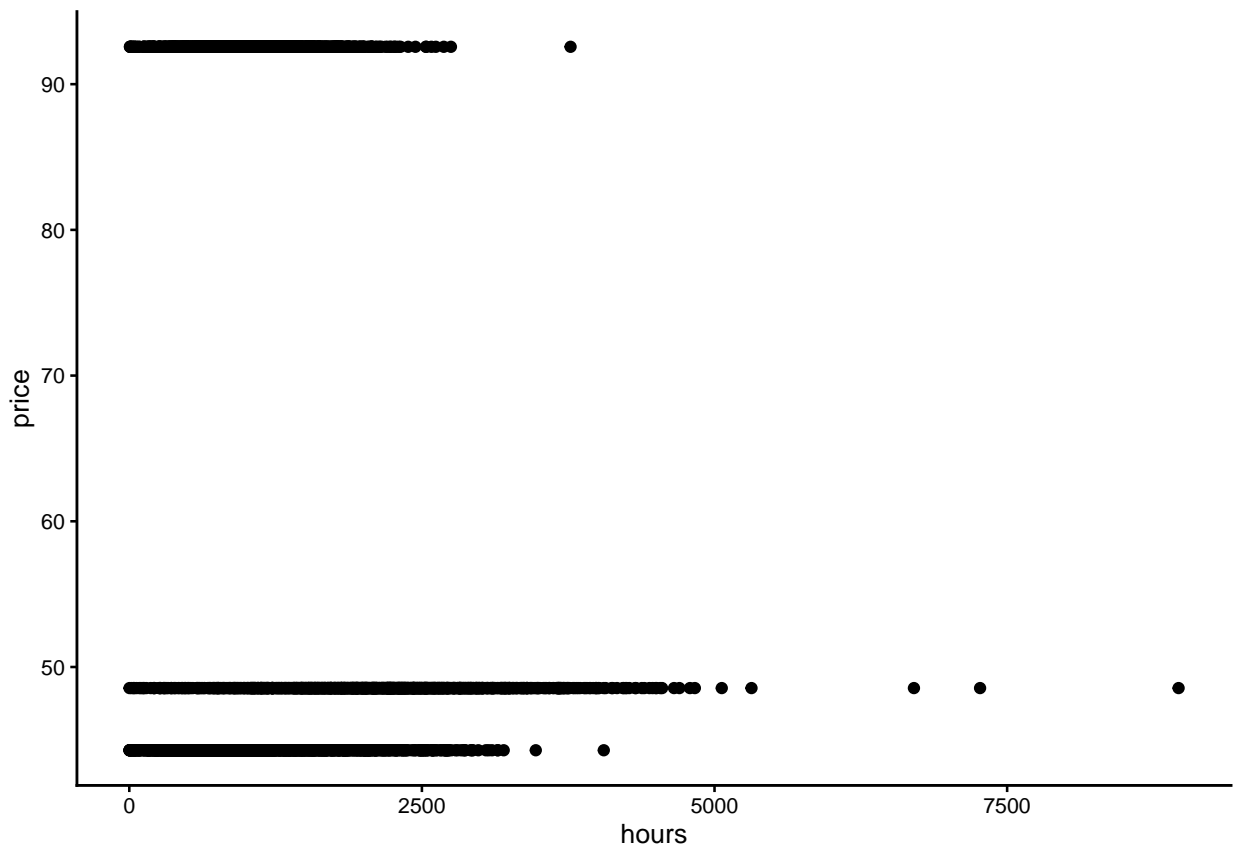
There are 0 vessels in the VMS dataset (1 ping per hour). There are 88 vessels in the AIS dataset (Many more pings).



Fuel prices

I have not been able to find fuel prices per year, so I'm using crude oil prices from Mexican oil production. That will be fixed eventually.





Preview of subsidy data

```
## # A tibble: 20 x 12
##   year  rnpa fishing_type subsidy_amount fuel_type target n_large_scale_v~
##   <int> <dbl> <chr>          <dbl> <chr>    <chr>          <int>
## 1  2011 4.08e7 Small scale          0 <NA>    <NA>          NA
## 2  2011 4.20e7 Small scale          0 <NA>    <NA>          NA
## 3  2011 2.03e8 Large scale      11780 Diesel    <NA>          3
## 4  2011 2.03e8 Small scale      11200 Gasoline  <NA>          3
## 5  2011 2.03e8 Small scale    204974 Gasoline  <NA>          0
## 6  2011 2.03e8 Large scale    480000 Diesel    <NA>          1
## 7  2011 2.03e8 Large scale    454200 Diesel    <NA>          4
## 8  2011 2.03e8 Large scale   8864176 Diesel    <NA>         NA
## 9  2011 2.03e8 Large scale    450000 Diesel    <NA>          1
## 10 2011 2.03e8 Large scale   3785320 Diesel    <NA>         NA
## 11 2011 2.03e8 Large scale    710622 Diesel    <NA>          2
## 12 2011 2.03e8 Large scale          0 <NA>    <NA>         NA
## 13 2011 2.03e8 Large scale    320000 Diesel    <NA>          1
## 14 2011 2.03e8 Large scale    451496 Diesel    <NA>          2
## 15 2011 2.03e8 Small scale     21000 Gasoline  <NA>          0
## 16 2011 2.03e8 Large scale   2400000 Diesel    <NA>         NA
## 17 2011 2.03e8 Large scale   2786870 Diesel    <NA>         NA
## 18 2011 2.03e8 Large scale     60000 Diesel    <NA>         NA
## 19 2011 2.03e8 Large scale     64660 Diesel    <NA>          2
## 20 2011 2.03e8 Large scale    400000 Diesel    <NA>          2
## # ... with 5 more variables: n_small_scale_vessels <int>, zone <chr>,
## #   state <chr>, municipality <chr>, location <chr>
```


We actually have these at the species-level but I'm combining it here. We also have the exvessel price of these landings.

```
## # A tibble: 20 x 3
##   ano_corte rnpa_unidad_economica landings
##   <int>          <dbl>      <dbl>
## 1     2000              0    758821
## 2     2003              0    74345
## 3     2005              0    84389
## 4     2007              0    26880
## 5     2014              1     4697
## 6     2015              1     1650
## 7     2014              5     3065
## 8     2015              5      2010
## 9     2015    100000000    55520
## 10    2005    100000017      458
## 11    2010    100000025    27010
## 12    2011    100000025     9320
## 13    2012    100000025     2570
## 14    2013    100000025     7420
## 15    2014    100000025     7380
## 16    2016    100000025    23900
## 17    2017    100000025    20500
## 18    2018    100000025    14750
## 19    2010    100000029     3150
## 20    2011    100000029      300
```

VMS data

```
## # A tibble: 20 x 4
##   ssvid      year rnp  hours
##   <chr>    <int> <chr> <dbl>
## 1 2804254023-7 2016 71019 1753
## 2 0402296223-2 2014 63669 1081
## 3 2301446823-9 2014 99671 1976
## 4 2804014023-3 2014 28381 1178
## 5 2804015223-8 2014 64493 2073
## 6 2804027923-9 2014 64394 1786
## 7 3001514323-1 2014 8409   748
## 8 2601022923-3 2014 50203 1878
## 9 2702327423-4 2014 69963 1402
## 10 3101134123-8 2014 33639 792
## 11 0201002523-7 2014 16105 1157
## 12 2003001423-6 2014 13201 489
## 13 2003005823-3 2014 216    594
## 14 2003045720-5 2014 2741   1315
## 15 2503052823-1 2014 15099 1354
## 16 2503056123-5 2014 60879 734
## 17 2503056623-4 2014 11817 1508
## 18 2804016523-6 2014 39792 721
## 19 0402001023-3 2015 68155 2990
## 20 2804027823-9 2015 71175 3066
```

AIS data

```
## # A tibble: 20 x 3
##   ssvid      year fishing_hours
##   <chr>    <int>         <dbl>
## 1 225983774 2015             0
## 2 345140500 2014          1084.
## 3 345080035 2018          2083.
## 4 345860501 2013           15.9
## 5 345035689 2018          1334.
## 6 345861978 2013             0
## 7 345080024 2016          1697.
## 8 345140011 2018          1909.
## 9 345080034 2016          2281.
## 10 123450020 2013           119.
## 11 345080029 2016          1999.
## 12 345080098 2020           290.
## 13 345140016 2020          1243.
## 14 123450100 2021           285.
## 15 345882880 2014           18.0
## 16 345905783 2014           16.5
## 17 345080026 2020          1324.
## 18 345080033 2014             0
## 19 345861978 2014           132.
## 20 345904851 2014           46.3
```