overview of process so far

List of predictor variables

• Forest height: FHT (m)

range: 0 - 20

year: 2019

• Length of growing season: LGS (days)

calculated by subtracting date of onset of greenness increase (15% increase in greeness) from onset greenness decrease (10% decrease in greenness)

range: 28 - 244 days

year: 2020

*there are quite a few spots without data

• Canopy cover: cc (%)

range: 0 - 100

year:

Terra Nova: 2010 with gaps filled in with 2005 (2015 had most of the park missing)

Gros Morne: 2015 with gaps filled in with 2010

- Enhanced vegetation index: EVI (NA)
 - amplitude: EVIamp, maximum minus minimum EVI during that year "interpreted as indicating the degree to which the leaf life cycles of individual plants and species are synchronized" (Gerard et al. 2020)
 - median: EVImed, median EVI during that year

range: 16 - 9600

year: 2019

Aspect: ASP (degrees)

range: 0-360

year: 2011

*when slope is 0 there can be no aspect, in these instances I put 0.01° as aspect

• Slope: SLO (percent)

range: 0 - 260

• Elevation: ELE (m)

• Dominant species: SPC (categorical) * decided to switch from dominant species to forest type because I didn't sample all dominant species

range: 8 categories (CS = coniferous shrub, DS = deciduous shrub, NS = not sufficiently restocked, bf = balsam fir, bs = black spruce, tA = trembling aspen, tL = tamarack, wB = white birch)

year: 2010

• Land cover type: CEC_LCT (categorical)

range: 5 categories (1 = temperate or sub-polar needle leaf forest, 6 = mixed forest, 8 = temperate or sub-polar shrubland, 14 = wetland, 16 = barrenland)

year: 2020

• Age class: FAC (categorical)

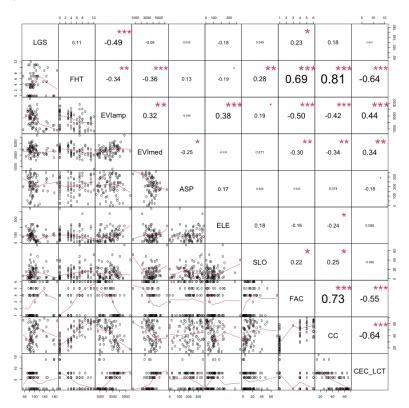
range: 5 categories (1 = 0 - 20, 2 = 21 - 40, 3 = 41 - 60, 4 = 61 - 80, 5 = 81+)

Correlation charts

at sites

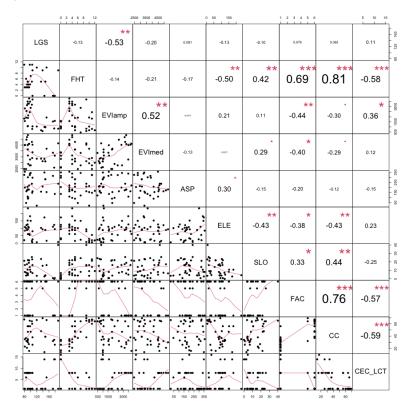
across both parks

localhost:4615 2/15



Variables <chr></chr>	VIF <dbl></dbl>
LGS	1.4141
FHT	3.1142
EVlamp	3.0911
EVImed	1.4256
ASP	1.1722
ELE	1.3655
SLO	1.3745
FAC	2.5918
CC	3.3890
CEC_LCT	2.9248

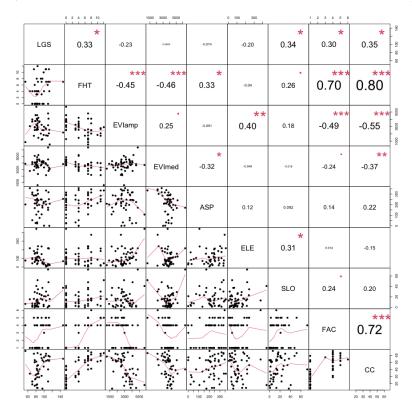
TN



Variables <chr></chr>	VIF <dbl></dbl>
LGS	1.4395
FHT	3.6099
EVlamp	1.9945
EVImed	1.3777
ASP	1.1939
ELE	1.4747
SLO	1.6086
FAC	2.7403
CC	3.6019

GM

localhost:4615 4/15

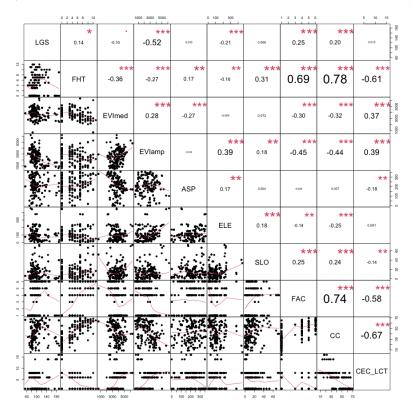


Variables <chr></chr>	VIF <dbl></dbl>
LGS	1.7306
FHT	3.5352
EVlamp	4.0959
EVImed	2.1202
ASP	2.0052
ELE	1.6722
SLO	1.7053
FAC	3.3277
CC	3.7500
CEC_LCT	4.1000

for subplots

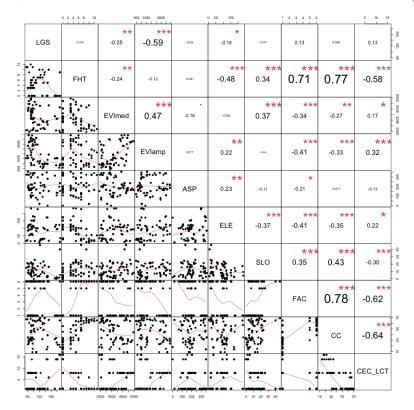
across both parks

localhost:4615 5/15



Variables <chr></chr>	VIF <dbl></dbl>
LGS	1.5367
FHT	2.9675
EVImed	1.4739
EVlamp	2.6100
ASP	1.1919
ELE	1.3763
SLO	1.3949
FAC	2.7363
CC	3.5036
CEC_LCT	2.6100

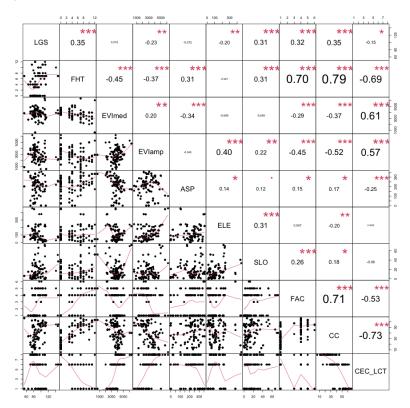
TN

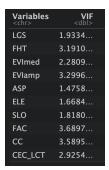


Variables <chr></chr>	VIF <dbl></dbl>
LGS	1.4528
FHT	3.9317
EVImed	1.8377
EVlamp	2.5037
ASP	1.2309
ELE	1.7139
SLO	1.6009
FAC	2.9003
CC	4.1071
CEC_LCT	4.1376

GM

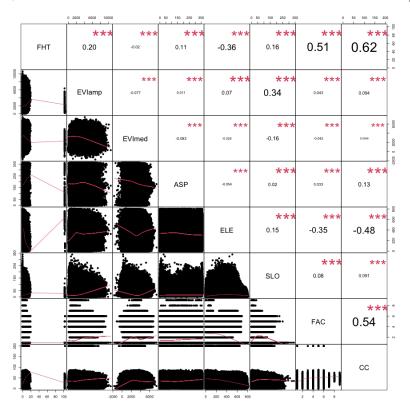
localhost:4615 7/15





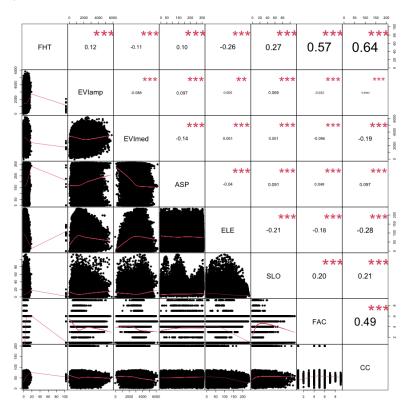
for whole parks

across both parks



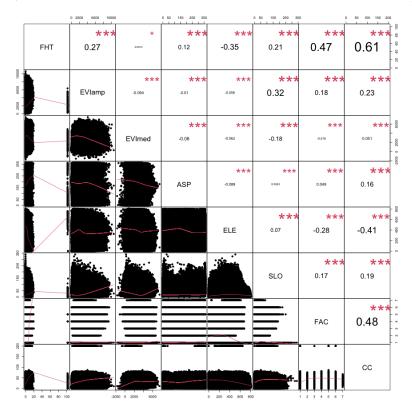
Variables <chr></chr>	VIF <dbl></dbl>
FHT	2.3911
EVlamp	1.2241
EVImed	1.0364
ASP	1.0305
ELE	1.4404
SLO	1.2438
FAC	1.6075
СС	2.2329

TN



Variables <chr></chr>	VIF <dbl></dbl>
FHT	1.8760
EVlamp	1.1895
EVImed	1.0542
ASP	1.0400
ELE	1.3228
SLO	1.2188
FAC	1.4365
CC	1.9178

GM



Variables <chr></chr>	VIF <dbl></dbl>
FHT	2.2183
EVlamp	1.0677
EVImed	1.0796
ASP	1.0346
ELE	1.1259
SLO	1.1311
FAC	1.6113
СС	2.0727

Conclusions

The patterns are quite consistent across each subset of the data.

My conclusions are predominantly based on the correlation charts, but I also am influenced a bit by the fact that none of the variance inflation factors are above 4.2, with most being below 2.

Remove CC

Canopy cover is very correlated with forest height. And although I have no biological reasoning for keeping forest height over canopy cover, the forest height data set is more complete and recent.

Remove FAC

The forest age is technically categorical, but the categories are ranges of a continuous variable. Is it ok to use it in a correlation chart?

If it is ok to use I propose removing FAC. It is very correlated with height. And I would prefer to keep height because it is actually a continuous variable.

Remove LCT

Land cover type is also actually a categorical variable so it's interesting that it is so correlated. It is probably because the higher numbers describe less vegetated land cover types (eg; 1 is temperate needle forest, 8 is shrubland and 16 is barrenland).

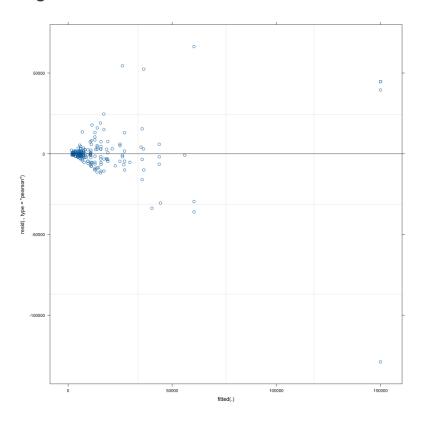
The only data we are really losing by removing it is whether the forested areas are needle leaf or mixed forest.

Residual distributions from GzLMs

for full model in Gros Morne with C per m2 of each subplot used as response variable and a random effect of site. Predictor variables have been scaled.

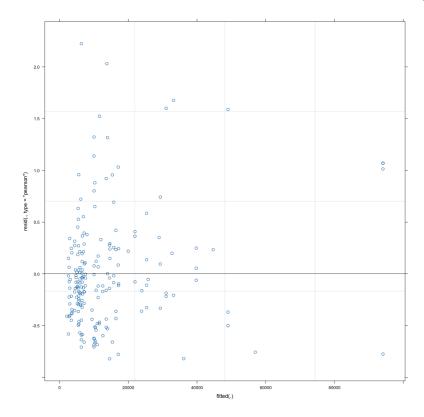
Gaussian

Log link



Gamma

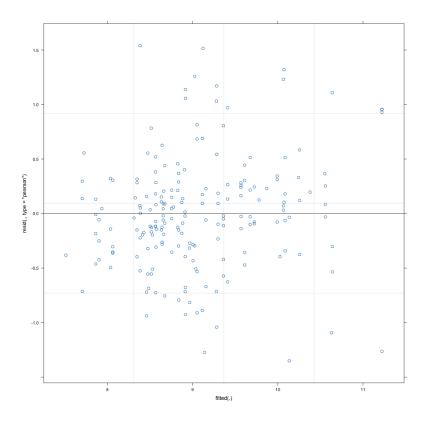
log link



Inverse and identity links don't converge

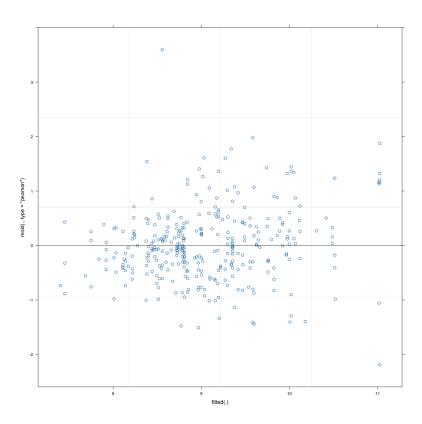
log tranformed linear model

scaled predictors



not scaled predictors: give warning

Warning: Some predictor variables are on very different scales: consider rescaling



Linear Models

subsetting data

park level

by park -> GM and TN get own models

subplot or site

sites -> average C per m2 across the 4 subplots at each site

subplots -> use C per m2 of each subplot and have random effect of site to account for pseudoreplication

structure

i.
$$log(C) \sim \epsilon$$

ii.
$$log(C) \sim (1|site) + \epsilon$$

iv.
$$log(C) \sim all env. + (1|site) + \epsilon$$

v.
$$log(C) \sim reduced env. + \epsilon$$

vi. $log(C) \sim reduced env. + (1|site) + \epsilon$

vii. $log(C) \sim univariate + \epsilon$

viii. $log(C) \sim univariate + (1|site) + \epsilon$