# Spot biomass index development

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```
# Librararies and data
knitr::opts_chunk$set(echo = TRUE, message = FALSE, warning = FALSE,
                      fig.align = "center")
library(tidyverse)
library(sf)
library(DHARMa)
library(tsibble)
library(lubridate)
library(INLA)
library(raster)
library(sdmTMB)
library(patchwork)
# load ChesMMAP data----
load(here::here("data/chesmmap.rdata"))
rm(count) # not using count data in analysis
# load harvest data----
load(here::here("data/processed_landings.rdata"))
# raster for projecting over
rast_res <- 2.5
r <- raster(extent(chesmmap_poly),</pre>
           res = rast res,
            crs = crs(ncrs)
# df from raster
gridded_cb <- chesmmap_poly %>%
  fasterize::fasterize(.,r) %>%
  dream::rst_to_tib()
# ggplotly(
  ggplot(gridded_cb)+
      qeom\_raster(aes(y = latitude, x = longitude)) +
#
      geom_sf(data = chesmmap_poly, fill = "transparent")
# )
# create prediction grid for biomass indices
create_pred_df <- F</pre>
if (create_pred_df){
  pred_df_glmm \leftarrow expand_grid(month = c(3,5,7,9,11),
```

```
year = 2002:2018,
                              gridded_cb) %>%
    mutate(month2 = ifelse(month == 3, "MAR",
                           ifelse(month == 5, "MAY",
                                  ifelse(month == 7, "JUL",
                                         ifelse(month == 9, "SEP",
                                                 ifelse(month == 11, "NOV",
                                                        NA))))),
           yseas = factor(paste(year, month2)),
           ymon = yearmonth(paste(year, month2)),
           ymon = as.numeric(ymon),
           fmonth = factor(month),
           year2 = factor(year),
           season = month2,
           metacomm = ifelse(latitude > 4200.327, "MD", "VA"))
 save(pred_df_glmm, file = here::here("data/prediction_grid.rdata"))
} else {
  load(file = here::here("data/prediction_grid.rdata"))
}
# nsims for simulation step
nsims <- 2000
```

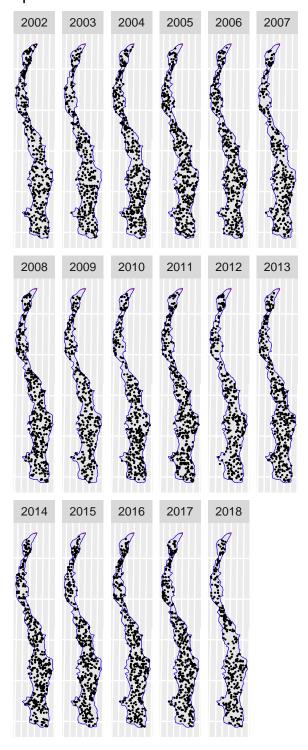
## Spot

#### Data spatial footprint

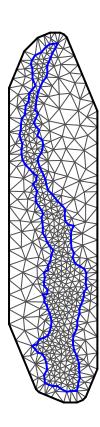
```
selected_species <- "spot"</pre>
ss <- "sp"
# data processing----
bio2 <- bio %>%
  filter(common %in% selected_species) %>%
  st_as_sf(coords = c("longitude","latitude"), crs = 4326) %>%
  st transform(crs = st crs(ncrs)) %>%
  mutate(stat_area = st_nearest_feature(.,va_mgmt)) %>%
  mutate(biomass = biomass/(areasw/1e6),
         presence = ifelse(biomass == 0, 0, 1),
         location = plyr::mapvalues(stat_area,
                                    from = 1:7,
                                     to = c("025","027","029",
                                            "CBUW", "CBUE",
                                            "CBLW", "CBLE"))) %>%
  dream::sfc_as_cols(names = c("longitude","latitude")) %>%
  st_set_geometry(NULL) %>%
  mutate(day = day(date),
         month = month(date),
         fmonth = factor(month),
         season = factor(season, levels = c("MAR", "MAY", "JUL",
                                             "SEP", "NOV")),
         season_char = as.character(season),
         cruise = factor(cruise),
```

```
year2 = factor(year),
         ymon = as.numeric(yearmonth(date)),
         yseas = factor(paste(year, season))) %>%
  mutate(metacomm = ifelse(stat_area %in% 1:3, "MD", "VA"))
# model fitting----
process <- T
p <-
  chesmmap_poly %>%
  as("sf")
ex <- extent(p)</pre>
ex[4] <- max(bio2 %>% filter(biomass > 0) %>% pull(latitude))
ex[3] <- min(bio2 %>% filter(biomass > 0) %>% pull(latitude))
pred_df_glmm_filt <- pred_df_glmm %>%
  filter(latitude <= ex[4],</pre>
         latitude >= ex[3])
chesmmap_poly2 <- st_crop(p, ex) %>% as_Spatial()
bio3 <- bio2 %>% filter(latitude <= ex[4],
                        latitude >= ex[3])
ggplot() +
  geom_sf(data = chesmmap_poly, color = "red") +
  geom_sf(data = chesmmap_poly2 %>% as("sf"),
          color = 'blue') +
  geom_point(data = bio3,
             aes(x = longitude, y = latitude),
             size = 0.1) +
  facet_wrap(~year, nrow = 3) +
  theme(axis.title = element_blank(),
        axis.text = element_blank(),
        axis.ticks = element_blank()) +
  # geom_sf(data = t, color = "blue") +
  labs(title = selected_species)
```

# spot



## Mesh



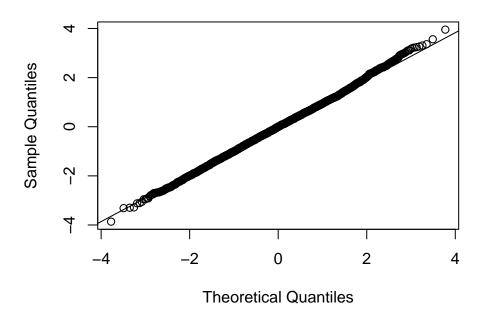
#### Fit model

#### Residuals

```
set.seed(123)
if (process){
  samps <- sdmTMBextra::predict_mle_mcmc(hurd_mod, mcmc_iter = 201, mcmc_warmup = 200)</pre>
  mcmc_res <- residuals(hurd_mod, type = "mle-mcmc", mcmc_samples = samps)</pre>
  save(samps, mcmc_res,
       file = here::here(paste0("data/model_output/mcmc_res_tw_",ss,".rdata")))
  stats::qqnorm(mcmc res, main = paste(selected species, "MCMC residuals, fam == tweedie"))
  stats::qqline(mcmc_res)
} else {
  load(here::here(paste0("data/model_output/mcmc_res_tw_",ss,".rdata")))
  stats::qqnorm(mcmc_res, main = paste(selected_species, "MCMC residuals, fam == tweedie"))
  stats::qqline(mcmc_res)
}
##
## SAMPLING FOR MODEL 'tmb_generic' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 0.119519 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 1195.19 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration: 1 / 201 [ 0%]
                                         (Warmup)
## Chain 1: Iteration: 20 / 201 [ 9%]
                                         (Warmup)
## Chain 1: Iteration: 40 / 201 [ 19%]
                                         (Warmup)
## Chain 1: Iteration: 60 / 201 [ 29%]
                                         (Warmup)
## Chain 1: Iteration: 80 / 201 [ 39%]
                                         (Warmup)
## Chain 1: Iteration: 100 / 201 [ 49%]
                                         (Warmup)
## Chain 1: Iteration: 120 / 201 [ 59%]
                                         (Warmup)
## Chain 1: Iteration: 140 / 201 [ 69%]
                                         (Warmup)
## Chain 1: Iteration: 160 / 201 [ 79%]
                                         (Warmup)
## Chain 1: Iteration: 180 / 201 [ 89%]
                                         (Warmup)
## Chain 1: Iteration: 200 / 201 [ 99%]
                                          (Warmup)
## Chain 1: Iteration: 201 / 201 [100%]
                                         (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 864.875 seconds (Warm-up)
```

```
## Chain 1: 3.044 seconds (Sampling)
## Chain 1: 867.919 seconds (Total)
## Chain 1:
```

## spot MCMC residuals, fam == tweedie

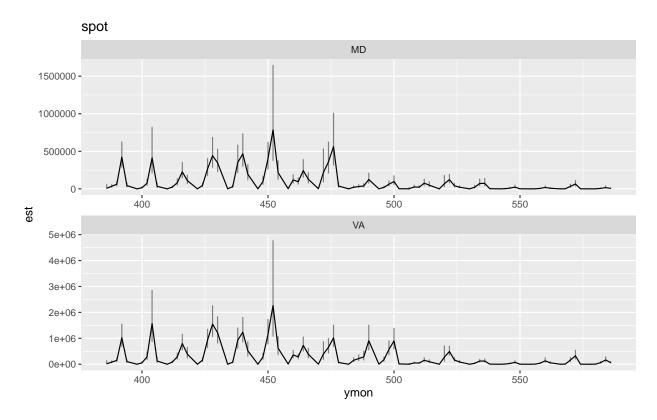


Generate predictions

```
# VA----
index_va \leftarrow lapply(c(3,5,7,9,11), \mbox{(m)} 
  nd <- pred_df_glmm_filt %>%
                  filter(month == m,
                         metacomm == "VA",
                         year %in% unique(bio3$year))
  p <- predict(hurd_mod, newdata = nd,</pre>
               return_tmb_object = TRUE)
  ii <- get_index(p, bias_correct = TRUE, area = 2.5^2)</pre>
  ii$month <- m
  ii
})
pred_out_seas_va <- do.call(rbind, index_va) %>%
  mutate(ymon = as.numeric(yearmonth(paste(year, month, sep = "-"))),
         common = selected_species,
         metacomm = "VA") %>%
  arrange(ymon)
# MD----
index_md <- lapply(c(3,5,7,9,11), \(m) {
```

```
nd <- pred_df_glmm_filt %>%
                  filter(month == m,
                         metacomm == "MD",
                         year %in% unique(bio3$year))
  p <- predict(hurd_mod, newdata = nd,</pre>
               return_tmb_object = TRUE)
  ii <- get_index(p, bias_correct = T, area = 2.5^2)</pre>
  ii$month <- m</pre>
  ii
})
pred_out_seas_md <- do.call(rbind, index_md) %>%
  mutate(ymon = as.numeric(yearmonth(paste(year, month, sep = "-"))),
         common = selected_species,
         metacomm = "MD") %>%
  arrange(ymon)
# combine----
pred_out_seas <- bind_rows(pred_out_seas_va,</pre>
                            pred_out_seas_md)
```

#### Visualize predictions



```
## save outputs

pred_out_seas_sp <- pred_out_seas %>%
    as_tibble()

save(pred_out_seas_sp,
    file = here::here(paste0("data/model_output/pred_out_seas_",ss,"_sim.rdata")))
```