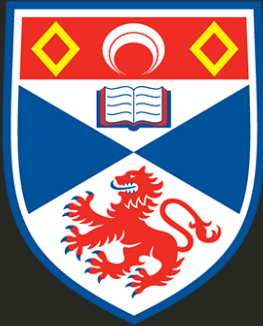


# Extras and advanced topics



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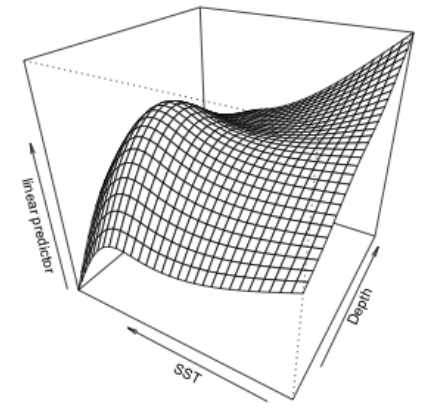
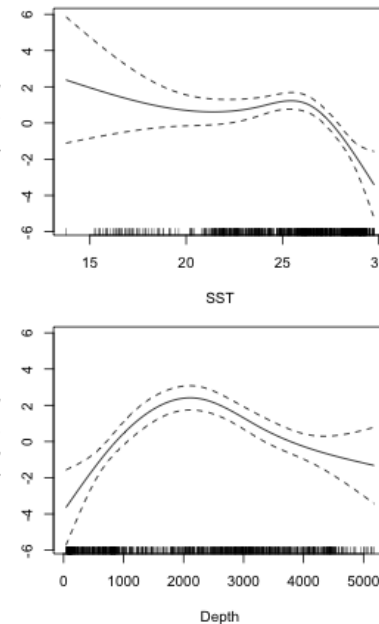
More complicated effects

# $s(x, y)$ doesn't always work

- Only works for  $bs="tp"$  or  $bs="ts"$
- Covariates are isotropic
- What if we wanted to use lat/long?
- Or, more generally: interactions between covariates?

# Enter `te()`

- We can build interactions using `te()`
- Construct 2D basis from 2 1D bases
- Biomass as a function of temperature and depth?
  - `te(temp_bottom, depth)`
- 🧠 "marginal 1Ds, join them up"



# Using `te()`

Just like `s()`:

```
dsm_te <- dsm(count ~ te(Depth, SST),  
              ddf.obj=df_hr,  
              observation.data=obs, segment.data=segs,  
              family=tw())
```

# summary

```
##
## Family: Tweedie(p=1.282)
## Link function: log
##
## Formula:
## count ~ te(Depth, SST) + offset(off.set)
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -20.3862      0.2831  -72.02   <2e-16 ***
## ---
## Signif. codes:
## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##              edf Ref.df      F p-value
## te(Depth,SST) 11.79  14.03  7.104   <2e-16 ***
## ---
## Signif. codes:
## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) =  0.117   Deviance explained = 36.6%
## -REML = 387.64   Scale est. = 4.5541      n = 949
```

# Things to fiddle with

- Setting  $k = 2$  ways:
  - $k=5$ : 5 for all covariates (total  $5 * 5 = 25$ )
  - $k=c(3, 5)$ : per basis, in order (total  $3 * 5 = 15$ )
- Setting  $bs = 2$  ways:
  - $bs="tp"$ : tprs for all bases
  - $bs=c("tp", "tp")$ : tprs per basis

# Pulling $te()$ apart: $ti()$

- Can we look at the components of the  $te()$
- $te(x, y) = ti(x, y) + ti(x) + ti(y)$

```
dsm_ti <- dsm(count ~ ti(Depth, SST) + ti(Depth) + ti(SST),  
              ddf.obj=df_hr,  
              observation.data=obs, segment.data=segs,  
              family=tw())
```



# summary

```
##
## Family: Tweedie(p=1.281)
## Link function: log
##
## Formula:
## count ~ ti(Depth, SST) + ti(Depth) + ti(SST) + offset(off.set)
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -20.4337      0.2868  -71.25   <2e-16 ***
## ---
## Signif. codes:
## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##              edf Ref.df      F  p-value
## ti(Depth,SST) 2.295  2.794  2.068    0.124
## ti(Depth)      3.477  3.817 16.905   < 2e-16 ***
## ti(SST)        3.175  3.505  8.492 4.08e-06 ***
## ---
## Signif. codes:
## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) =  0.114   Deviance explained =   36%
```

# Space x time

- We had a 2d spatial model, add time?
  - `te(x, y, year)` ?
- `d=` groups covariates
  - `te(x, y, year, d=c(2, 1))` gives `x`, `y` smooth and `year` smooth tensor
- (Assuming default `k=` and `bs=` for bases above)

# Fiddling

- Often fewer temporal replicates
  - Fewer years than unique locations
  - $k$  = smaller for temporal covariate?
- Use cubic spline basis for time?
  - simpler basis, even knot placement
- When using `ti()` everything needs to match up!

Other effects

# Random effects

- "Simple" random slope/random intercept models
- `s(..., bs="re")`
- **think** about what these models mean

# Factor-smooth interactions

- What if we only have a few "years"?
- What if we don't think the "years" are smooth?
  - (Before/after?)
- Terms like `s(Depth, by=year)` change the smooth by year
- also `s(Depth, year, bs="fs")` (lots of ways to specify)
- see [Pedersen et al. \(2019\)](#) for more on these models

# Availability

# Availability

- Is an animal *available* to be detected
- e.g., diving marine mammals
- Primitive way to do this in dsm
- `availability`= for each segment (only for count models)
- Active research area!



$g(0)$ , MRDS etc

# Mark-recapture distance sampling

- Will be able to include these models in next dsm release
- Only independent observer ("io") and trial ("trial") modes supported
- [Example here](#)

# Combining multiple surveys

# Combining multiple surveys

- What about combining aerial/shipboard data?
- Different detection functions
- Again, next dsm release allows this
- **Fitting complicated models** example

Finally...

# Recent developments

- New dsm out in the next few weeks!
- Fitting DSMs in JAGS/Nimble
- DenMod project has produced lots of methodology
- Society for Marine Mammalogy meeting December

# Extra bits

Deviance explained, explained



# Deviance explained, explained

- Avoid  $R^2$  (see [these notes](#) for more info)
- But what about deviance explained?
- First, what is it?

$$D = -2(l_s - l)$$

where  $\mathcal{L}_s$  is the *saturated* log likelihood and  $\mathcal{L}$  is the likelihood of our model.

- Saturated means the "best" model we can get, one parameter per data point.
- So meaning is it's relative to the best we can do *for this model*

# Deviance explained, explained

- mgcv reports "Deviance explained" as a percentage

$$D_{\%} = 100(l_s - l)/l_s$$

- Problem: for different models (with different numbers of parameters)  $l_s$  is different
- So are we making fair comparisons?
- AIC is simpler and easier to think about!

More info on deviance for GAMs

# More difficulties with explanatory power

- Low (<60%) deviance is common. But why?
- Sampling a temporally variable system
- Revisiting the same place multiple times, we might get zero counts twice and then one large count.
- What should the model make of this?
- Without explicit temporal model, it tries to average
- So prediction will be a "medium" count, bad prediction for the zeros and the large counts
- No one is happy!
- See observed vs. expected diagnostics etc

That's all folks!