# **Supplementary material: ICES Journal of Marine Science, 75**

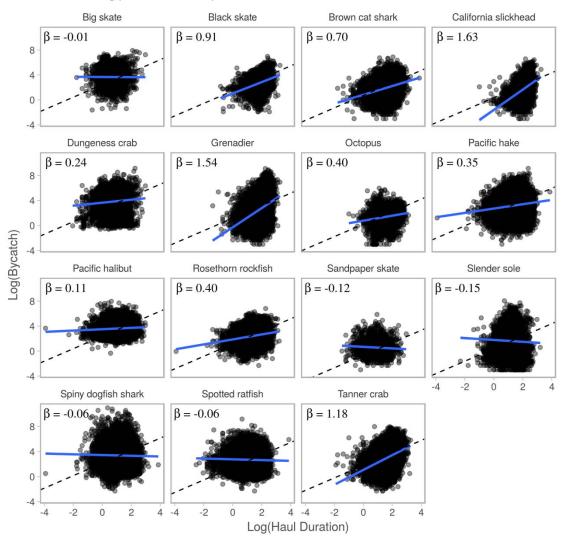
# The utility of spatial model-based estimators of unobserved bycatch

Brian C. Stock, Eric J. Ward, James T. Thorson, Jason E. Jannot, and Brice X. Semmens

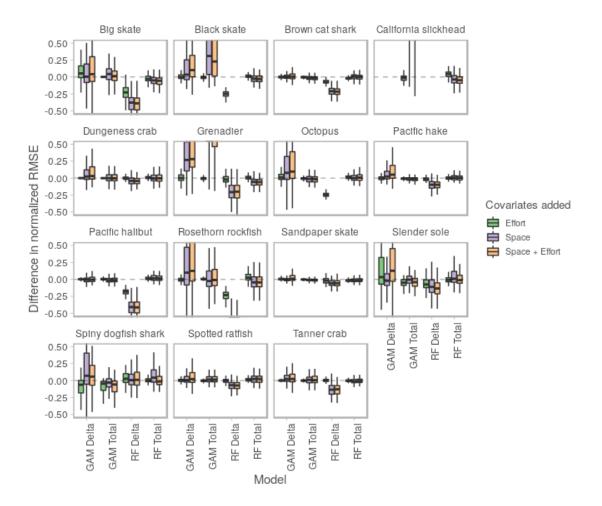
**Table S1.** Annual bycatch (tonnes) and bycatch rate (percent of hauls) for species selected from the US West Coast Groundfish Observer Program (WCGOP) dataset. All selected species are exclusively discarded. The summarized data are 35 440 post-IFQ hauls (4007 trips) observed in 2011–2015 in the area north of Cape Falcon, Oregon (45.77°N).

	20	11	20	12	20	13	20:	14	20	15
Species	Catch (t)	% Hauls								
Big skate	25.2	10.2	33.9	10.8	24.1	9.1	68.2	17.9	34.0	18.5
Black skate	18.5	17.3	15.3	14.4	14.0	15.2	13.7	15.3	10.5	13.3
Brown catshark	19.3	45.6	21.5	43.5	24.3	45.4	25.4	45.4	22.9	45.8
California slickhead	9.3	12.3	6.3	8.1	6.4	9.0	5.3	9.3	4.7	6.7
Dungeness crab	120.1	27.6	137.8	32.7	98.2	25.3	105.0	31.9	86.8	30.7
Grenadier	116.8	34.0	121.9	29.8	108.1	29.8	64.0	26.0	42.0	22.5
Octopus	3.7	15.9	2.8	13.2	4.7	15.4	3.4	13.2	2.4	10.9
Pacific hake	147.6	55.1	165.8	58.2	148.0	54.2	122.7	56.2	143.8	60.7
Pacific halibut	61.0	29.3	62.3	30.3	63.7	27.1	53.8	33.9	65.9	36.2
Rosethorn rockfish	0.7	3.3	0.7	4.5	0.9	5.9	0.8	4.2	0.1	2.5
Sandpaper skate	25.9	44.9	33.0	48.4	35.0	51.8	33.9	53.9	34.3	55.4
Slender sole	18.7	20.7	35.2	23.6	46.7	26.9	31.7	31.3	28.2	31.2
Spiny dogfish shark	268.7	42.5	261.4	46.5	258.0	39.2	262.9	46.9	165.5	42.2
Spotted ratfish	50.7	37.5	58.7	42.3	69.0	41.9	57.3	44.4	59.4	48.8
Tanner crab	136.3	46.3	85.1	38.6	104.2	39.7	84.3	39.4	84.9	34.4

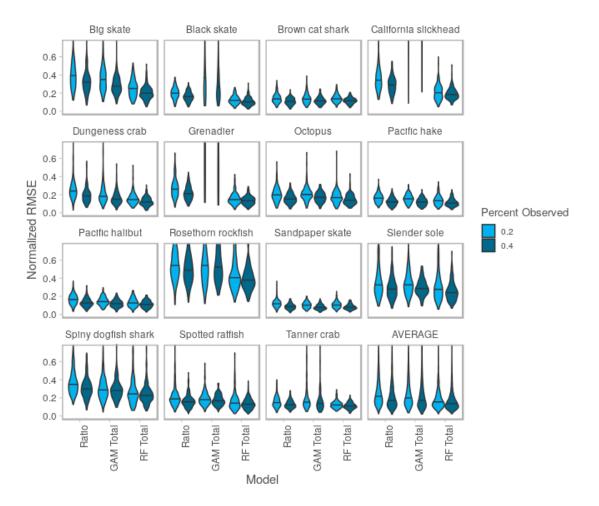
#### Effort = Log(Haul Duration)



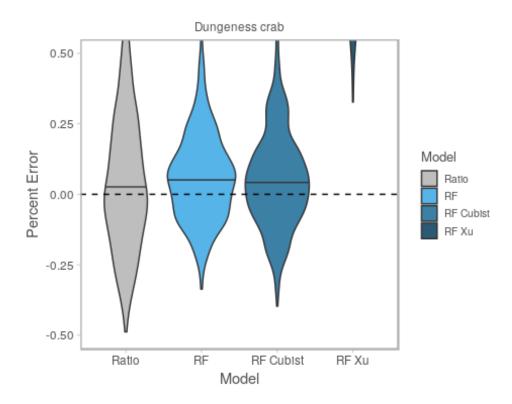
**Figure S1.** Estimated relationships between fishing effort (haul duration in hours) and bycatch (kg) for 15 species analyzed in the west coast groundfish trawl fishery. The slope terms  $\beta$  of loglog linear models are exponents of an assumed power law fit to each species, Bycatch =  $\alpha$ Effort<sup> $\beta$ </sup>. Most  $\beta$  are much less than 1, indicating the relationship between bycatch and effort is either weak or not linear. Data ( $n = 35\,440$ ) consist of observed hauls from the West Coast Groundfish Observer Program recorded in 2011–2015 in the area north of Cape Falcon, Oregon (45.77°N).



**Figure S2.** Change in predictive performance (normalized RMSE) when adding fishing effort and spatial location as covariates in each model. For many species, adding space to the GAM-Delta and GAM-Total models led to worse predictions (positive change in RMSE, above dashed line). On the other hand, adding space to the RF-Delta model consistently improved predictions (negative change in RMSE, below dashed line). For RF-Total, including space had either slightly improved predictions or had no effect. Adding effort had little effect for nearly all species and models and never had a larger effect than adding space.



**Figure S3.** Predictive performance (normalized RMSE) for different levels of simulated observer coverage. Averaged across species, RF-Total had lower median RMSE than the ratio estimator, even at half the observer coverage (RF-Total at 20%: 0.155, Ratio at 40%: 0.180). GAM-Total failed to converge for 3/15 species.



**Figure S4.** Performance of RF bias correction methods applied to Dungeness crab bycatch (percent error, PE, averaged across years 2011-2015). The ratio estimator is unbiased (median PE = 0.002). RF is positively biased (median PE = 0.055) and Cubist is less positively biased (median PE = 0.043). Cubist reduces bias by fitting a linear model in regression tree terminal nodes instead of using the data mean (Quinlan, 1992, 1993). The second method (Xu, 2013) fits a second RF model to the residuals of the original RF, but this method performed poorly (median PE = 1.107, off chart).

#### **Random forest and GAM models**

For each (of 15) species, we fit the models described below. **fac(x)** is shorthand for treating covariate *x* as a factor. All random forest models were fit with **ntree=500** and default **mtry**. All GAMs were fit with **family=Tweedie(p=tweedie\_p[i])**, **method="REML"**. All data are at the haul-level.

- *Y*: response, bycatch/discards (in kg)
- *year*: year (5 levels: 2011–2015)
- *depth\_interval*: depth of haul (3 levels: < 125, 126–250, > 250 fathoms)
- *season*: season (2 levels: summer, winter)
- bimonth: bimonthly period (6 levels: Jan–Feb, Mar–Apr, . . . , Nov–Dec)
- *logret*: log(retained target species catch in kg)
- *lat*: degrees latitude
- *long*: degrees longitude

For random forest models, we report the percent variance explained [pseudo-R2 = 1-mse Var(y)]. Random forest covariate effect plots were created using the **forestFloor R** package and custom code.

For GAM models, we report the output of **mgcv::summary.gam**, including percent deviance explained. GAM covariate effect plots were created using the **visreg R** package.

#### **RF-Total**

Designed to *mimic the stratified ratio estimator* by treating **year**, **season**, and **depth\_interval** as factors. **bimonth** is included as linear and quadratic terms to avoid confounding with fac(season):

```
randomForest[Y \sim fac(year) + fac(season) + fac(depth\_interval) + bimonth + I(bimonth^2) + logret + lat + long + I(lat^2) + I(long^2)]
```

#### **GAM-Total**

Designed to *mimic the stratified ratio estimator* by treating **year**, **season**, and **depth\_interval** as factors. **bimonth** is included as linear and quadratic terms to avoid confounding with **fac(season)**:

```
gam[Y \sim fac(year) + fac(season) + fac(depth interval) + bimonth + I(bimonth^2) + logret + s(lat, long, k=50)]
```

#### **RF-Nonlinear**

Allow random forest to fit covariates with *full non-linear flexibility* (not as factors):

```
randomForest(Y \sim year + julian_day + time + depth + gear + logret + lat + long)
```

New covariates introduced:

- *julian\_day*: Julian day of year
- *time*: time of day in hours
- *depth*: depth of haul in fathoms

• *gear*: gear type (3 levels: groundfish trawl w/ footrope < 8 inches, groundfish trawl w/ footrope > 8 inches, pineapple trawl)

## **GAM-Nonlinear**

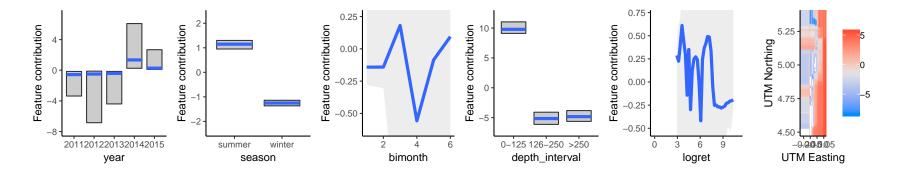
Allow GAM to fit covariates with *full non-linear flexibility* (not as factors):

```
gam[Y \sim s(year, k=5) + s(julian\_day, k=5) + s(time, k=5) + s(depth, k=5) + gear + logret + s(lat, long, k=50)]
```

## Species 1: Big Skate

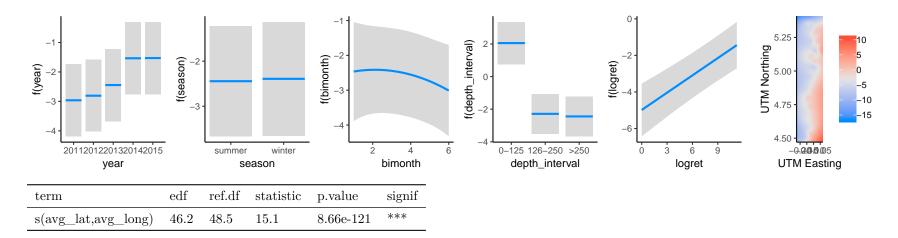
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 1989.101 % Var explained: 47.8



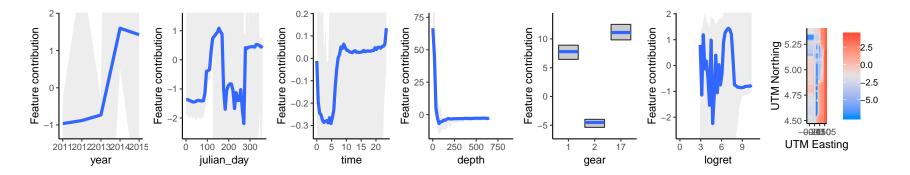
#### $GAM_{Total}$

 $R-sq.(adj) = 0.233 \ Deviance \ explained = 65.8\% \ -REML = 32567 \ Scale \ est. \ = 102.63 \ n = 35440$ 



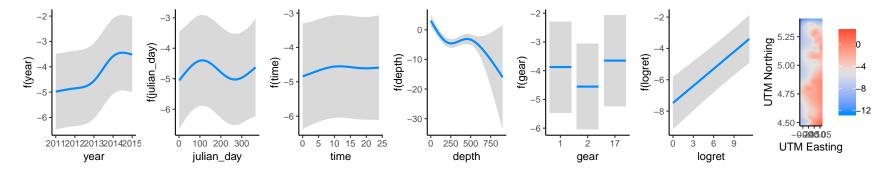
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-6.02	0.806	-7.47	8.29e-14	***
seasonwinter	0.0525	0.248	0.212	0.832	
bimonth	0.169	0.37	0.457	0.648	
I(bimonth^2)	-0.0398	0.0486	-0.819	0.413	
year2012	0.156	0.158	0.987	0.324	
year2013	0.518	0.164	3.15	0.00163	**
year2014	1.43	0.156	9.17	4.96e-20	***
year2015	1.43	0.16	8.97	3.2e-19	***
$depth\_interval0-125$	4.5	0.348	12.9	3.13e-38	***
$depth\_interval 126-250$	0.157	0.262	0.598	0.55	
logret	0.315	0.05	6.32	2.73e-10	***

Mean of squared residuals: 1784.603 % Var explained: 53.17



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = 0.26 Deviance explained = 67.6% -REML = 32318 Scale est. = 95.592 n = 35440



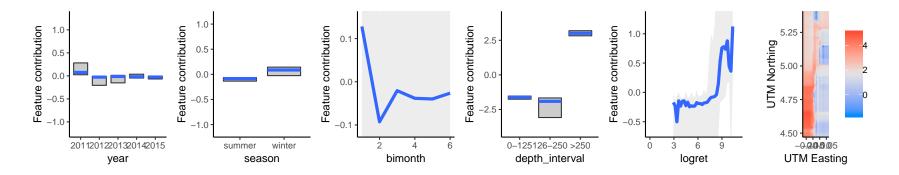
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	45.6	48.4	9.01	5.26e-63	***
s(year)	3.91	3.99	41.7	7.89e-35	***
s(depth)	3.89	3.99	31.5	3.1e-26	***
$s(julian\_day)$	3.75	3.96	5.17	0.000357	***
s(time)	2.63	3.17	0.455	0.708	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-3.88	0.458	-8.49	2.23e-17	***
gear2	-0.681	0.289	-2.35	0.0186	*
gear17	0.224	0.109	2.06	0.0399	*
logret	0.364	0.0502	7.25	4.38e-13	***

## Species 2: Black Skate

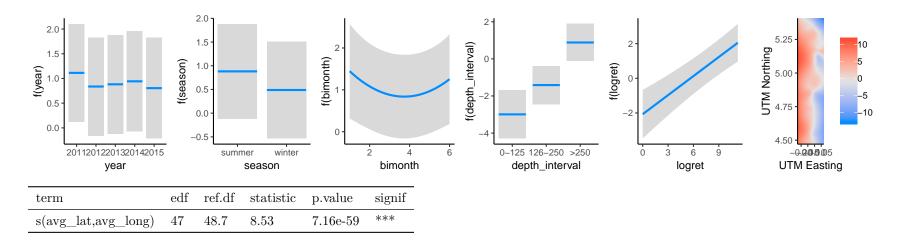
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 253.1007 % Var explained: 48.13



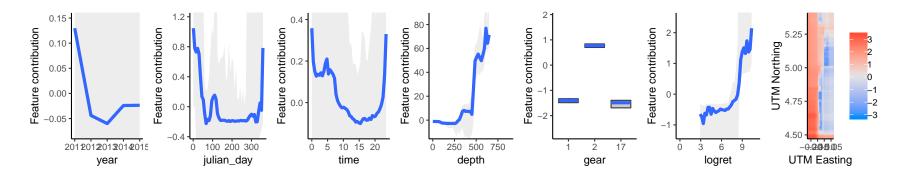
#### $GAM_{Total}$

R-sq.(adj) = -1.54 Deviance explained = 56% -REML = 33123 Scale est. = 139.75 n = 35440



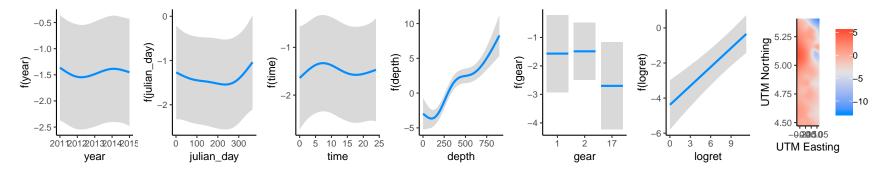
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-0.848	0.71	-1.19	0.232	
seasonwinter	-0.393	0.203	-1.94	0.0529	
bimonth	-0.606	0.272	-2.23	0.0257	*
I(bimonth^2)	0.0812	0.0368	2.21	0.0274	*
year2012	-0.276	0.172	-1.61	0.108	
year2013	-0.231	0.169	-1.37	0.171	
year2014	-0.171	0.18	-0.947	0.344	
year2015	-0.309	0.191	-1.61	0.107	
$depth\_interval0-125$	-3.87	0.461	-8.4	4.8e-17	***
$depth\_interval 126-250$	-2.29	0.193	-11.9	2.2e-32	***
logret	0.366	0.0634	5.77	7.85e-09	***

Mean of squared residuals: 226.5094 % Var explained: 53.58



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = -0.416 Deviance explained = 62.3% -REML = 32187 Scale est. = 122.09 n = 35440



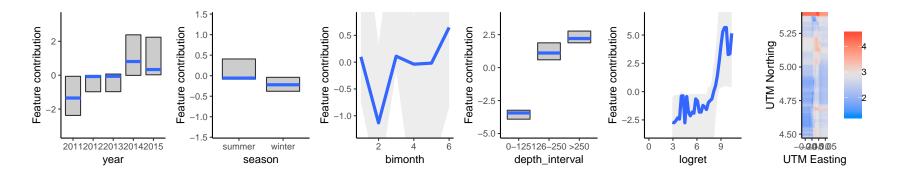
term	edf	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	46.2	48.5	3.19	6.26 e-13	***
s(year)	3.32	3.75	0.453	0.764	
s(depth)	3.99	4	92.9	1.66e-78	***
$s(julian\_day)$	3.45	3.83	0.877	0.486	
s(time)	3.5	3.86	0.702	0.569	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-4.02	0.614	-6.55	5.71e-11	***
gear2	0.0803	0.474	0.17	0.865	
gear17	-1.13	0.496	-2.28	0.0223	*
logret	0.359	0.0623	5.77	8.02e-09	***

## Species 3: Brown Cat Shark

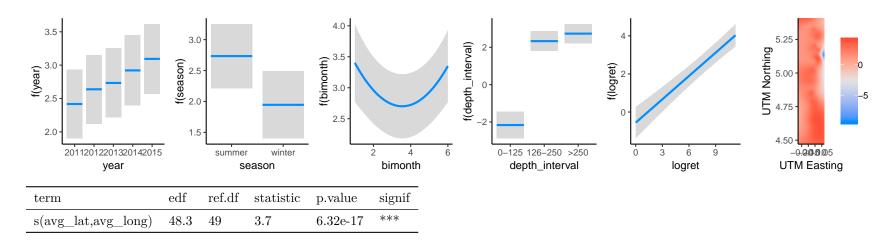
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 356.8372 % Var explained: 20.8



#### $GAM_{Total}$

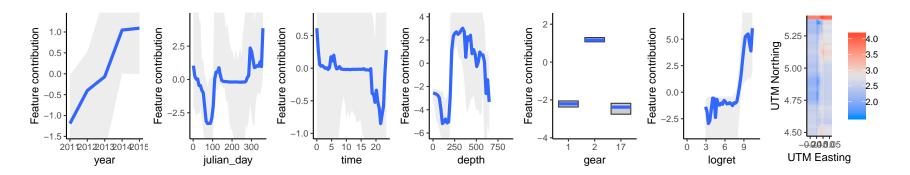
R-sq.(adj) = 0.121 Deviance explained = 45.4% -REML = 74672 Scale est. = 88.94 n = 35440



term	estimate	std.error	statistic	p.value	signif
(Intercept)	-0.18	0.479	-0.376	0.707	
seasonwinter	-0.788	0.135	-5.82	5.99e-09	***
bimonth	-0.766	0.184	-4.16	3.13e-05	***
I(bimonth^2)	0.108	0.0249	4.34	1.46 e - 05	***
year2012	0.221	0.119	1.86	0.0631	•
year2013	0.316	0.115	2.75	0.00591	**
year2014	0.503	0.121	4.14	3.44 e - 05	***
year2015	0.676	0.122	5.55	2.9e-08	***
$depth\_interval0-125$	-4.9	0.271	-18.1	1.6e-72	***
$depth\_interval126-250$	-0.402	0.0936	-4.3	1.73e-05	***
logret	0.408	0.0423	9.65	5.37e-22	***

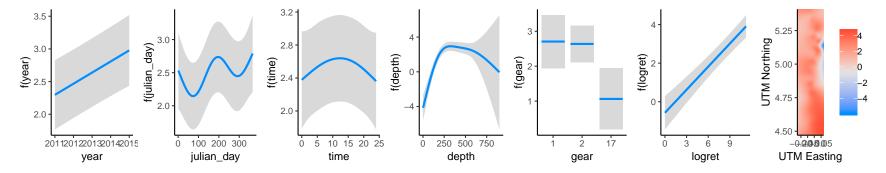
# $\mathbf{RF}_{\mathbf{Nonlinear}}$

Mean of squared residuals: 344.8937 % Var explained: 23.45



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = 0.132 Deviance explained = 46.7% -REML = 74340 Scale est. = 91.971 n = 35440



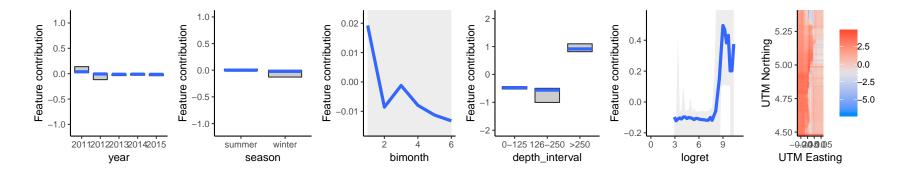
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	3.06	4.15e-12	***
s(year)	1	1.01	35.4	2.48e-09	***
s(depth)	3.98	4	35.1	2.62e-29	***
$s(julian\_day)$	3.98	4	5.69	0.00014	***
s(time)	3.32	3.75	1.27	0.279	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-2.69	0.412	-6.52	7.02e-11	***
gear2	-0.0681	0.286	-0.238	0.812	
gear17	-1.64	0.329	-4.99	6.13e-07	***
logret	0.399	0.0432	9.22	3.04e-20	***

## Species 4: California Slickhead

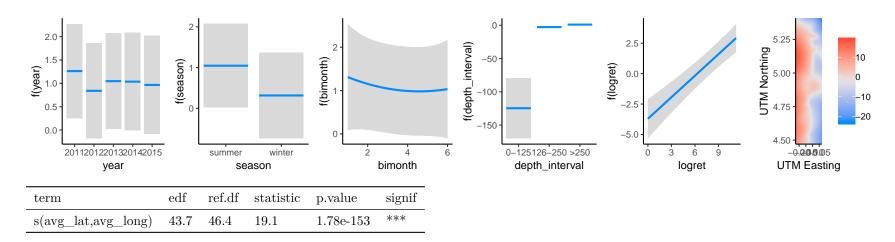
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 145.0971 % Var explained: 42.62



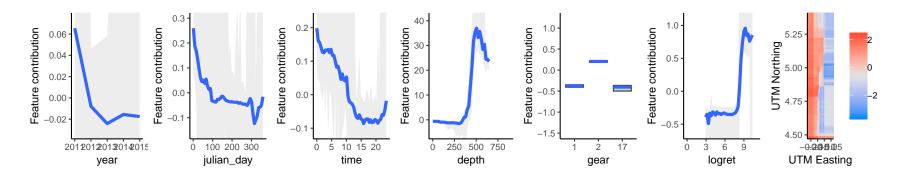
#### $GAM_{Total}$

 $R-sq.(adj) = -9.19e + 03 \ Deviance \ explained = 65.7\% \ -REML = 18895 \ Scale \ est. = 97.632 \ n = 35440$ 



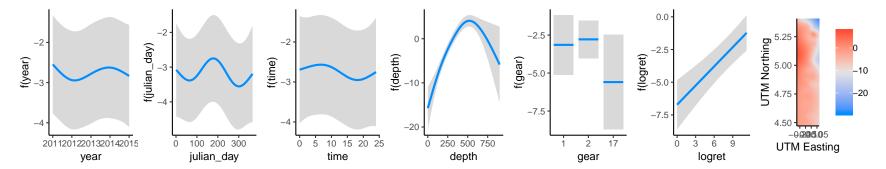
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-5.31	0.979	-5.42	5.87e-08	***
seasonwinter	-0.731	0.252	-2.9	0.00375	**
bimonth	-0.235	0.343	-0.686	0.492	
I(bimonth^2)	0.0256	0.0463	0.553	0.58	
year2012	-0.423	0.214	-1.98	0.048	*
year2013	-0.216	0.211	-1.03	0.304	
year2014	-0.227	0.23	-0.986	0.324	
year2015	-0.297	0.242	-1.23	0.219	
$depth\_interval0-125$	-126	23.1	-5.43	5.6e-08	***
$depth\_interval126-250$	-3.74	0.251	-14.9	2.68e-50	***
logret	0.59	0.0808	7.31	2.83e-13	***

Mean of squared residuals: 127.1781 % Var explained: 49.7



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = -14.3 Deviance explained = 76.2% -REML = 17595 Scale est. = 103.48 n = 35440



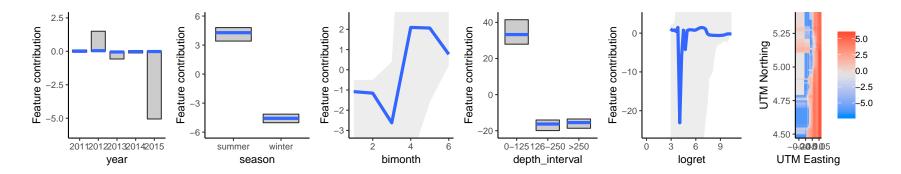
term	edf	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	43	46.3	3.73	2e-16	***
s(year)	3.56	3.89	0.853	0.476	
s(depth)	3.94	4	120	4.75e-102	***
s(julian_day)	3.9	3.99	1.84	0.13	
s(time)	3.27	3.71	0.864	0.508	

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-11.5	1.07	-10.7	8.35e-27	***
gear2	0.362	0.792	0.457	0.647	
gear17	-2.45	1.53	-1.6	0.109	
logret	0.489	0.0915	5.34	9.4e-08	***

## Species 5: Dungeness Crab

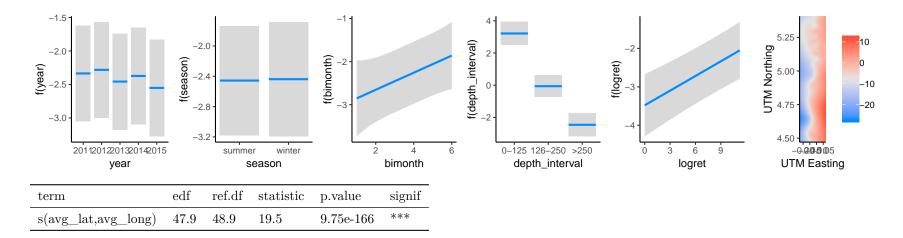
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 11727.1 % Var explained: 28.81



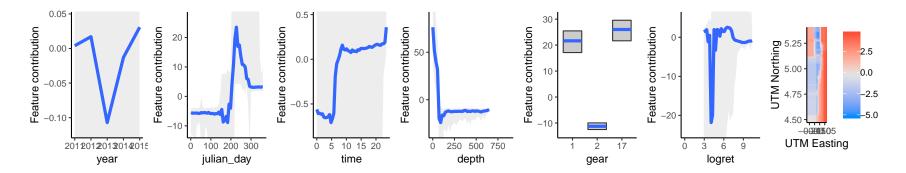
#### $GAM_{Total}$

R-sq.(adj) = 0.213 Deviance explained = 75.9% -REML = 65965 Scale est. = 55.075 n = 35440



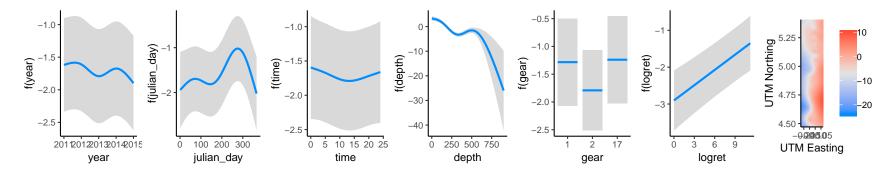
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-5.35	0.542	-9.87	6.04e-23	***
seasonwinter	0.0186	0.162	0.115	0.909	
bimonth	0.198	0.26	0.763	0.445	
I(bimonth <sup>2</sup> )	6.62e-05	0.0336	0.00197	0.998	
year2012	0.0538	0.0808	0.666	0.506	
year2013	-0.122	0.0853	-1.42	0.154	
year2014	-0.0376	0.0877	-0.429	0.668	
year2015	-0.216	0.0901	-2.4	0.0165	*
$depth\_interval0-125$	5.67	0.234	24.3	3.16e-129	***
$depth\_interval126-250$	2.4	0.195	12.3	1.03e-34	***
logret	0.127	0.0246	5.16	2.5e-07	***

Mean of squared residuals: 10642.64 % Var explained: 35.39



# ${\rm GAM}_{\rm Nonlinear}$

 $R-sq.(adj) = 0.231 \ Deviance \ explained = 77.3\% \ -REML = 65494 \ Scale \ est. \ = 50.523 \ n = 35440$ 



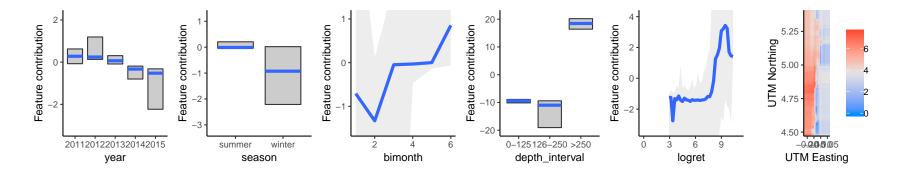
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.8	48.9	13.9	6.66e-111	***
s(year)	3.89	3.99	4.3	0.00226	**
s(depth)	3.97	4	104	1.55e-88	***
$s(julian\_day)$	3.97	4	29.3	2.62e-24	***
s(time)	2.87	3.39	1.01	0.395	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-2.05	0.238	-8.63	6.21e-18	***
gear2	-0.508	0.16	-3.18	0.00148	**
gear17	0.0437	0.0594	0.735	0.462	
logret	0.138	0.0241	5.73	1.02e-08	***

## Species 6: Grenadier

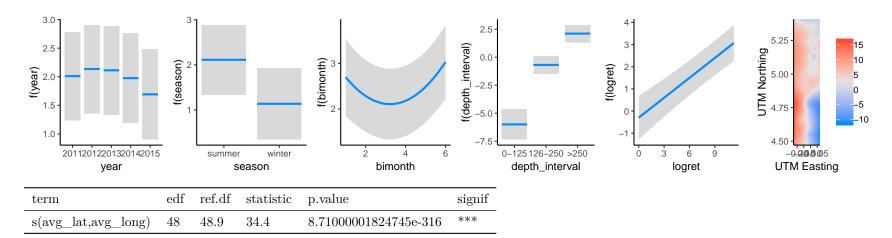
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 18071.56 % Var explained: 49.32



#### $GAM_{Total}$

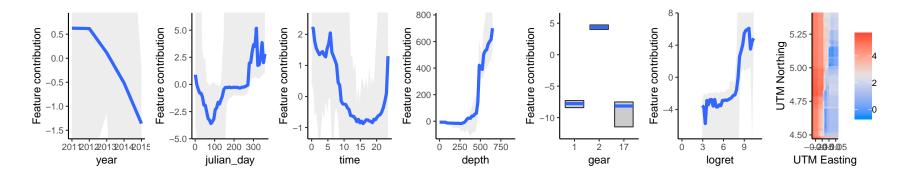
 $R\text{-sq.}(\mathrm{adj}) = \text{-}241$  Deviance explained = 66.8% -REML = 63079 Scale est. = 133.84 n = 35440



term	estimate	std.error	statistic	p.value	signif
(Intercept)	1.6	0.462	3.47	0.000516	***
seasonwinter	-0.977	0.131	-7.48	7.64e-14	***
bimonth	-0.771	0.181	-4.27	1.95 e-05	***
I(bimonth^2)	0.12	0.0244	4.91	9.18e-07	***
year2012	0.125	0.105	1.19	0.235	
year2013	0.102	0.106	0.964	0.335	
year2014	-0.0354	0.118	-0.299	0.765	
year2015	-0.318	0.13	-2.46	0.014	*
$depth\_interval0-125$	-8.11	0.58	-14	2.96e-44	***
$depth\_interval 126-250$	-2.8	0.137	-20.5	7.95e-93	***
logret	0.299	0.0394	7.6	2.95e-14	***

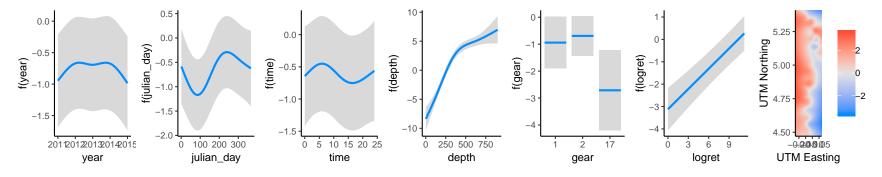
# $\mathbf{RF}_{\mathbf{Nonlinear}}$

Mean of squared residuals: 15794.11 % Var explained: 55.71



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = 0.397 Deviance explained = 78% -REML = 58132 Scale est. = 102.36 n = 35440



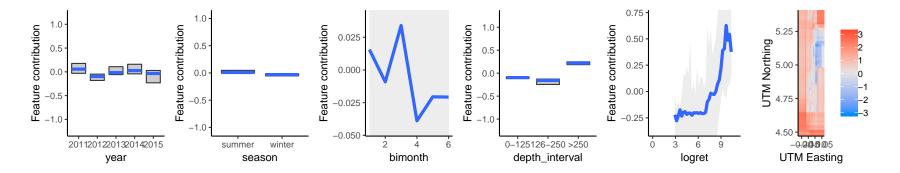
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	45.7	48.1	11.6	3.84e-87	***
s(year)	3.85	3.99	3.98	0.00323	**
s(depth)	3.97	4	371	$2.52000000822904 \mathrm{e}\text{-}316$	***
$s(julian\_day)$	3.97	4	18.2	5.15e-15	***
s(time)	3.79	3.97	2.32	0.0554	

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-3.54	0.435	-8.14	3.97e-16	***
gear2	0.25	0.322	0.776	0.438	
gear17	-1.77	0.685	-2.58	0.00975	**
logret	0.301	0.0382	7.89	3.01e-15	***

## Species 7: Octopus Unid

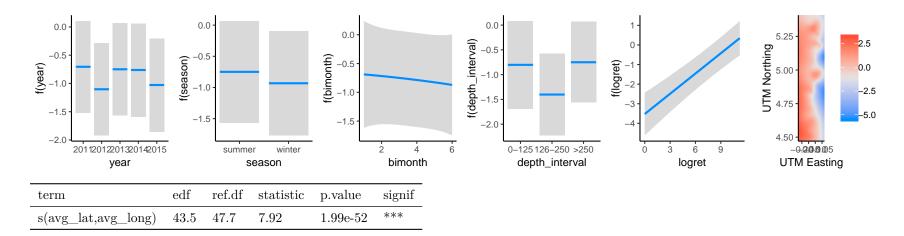
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 36.92363 % Var explained: 2.19



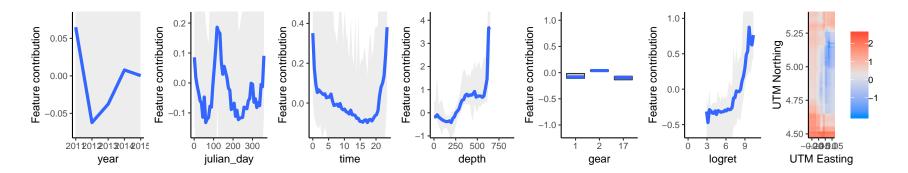
#### $GAM_{Total}$

R-sq.(adj) = 0.0433 Deviance explained = 20.2% -REML = 27714 Scale est. = 56.927 n = 35440



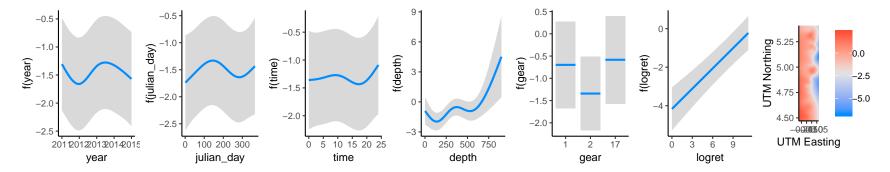
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-2.76	0.559	-4.95	7.61e-07	***
seasonwinter	-0.183	0.164	-1.12	0.264	
bimonth	-0.0225	0.226	-0.0993	0.921	
I(bimonth^2)	-0.002	0.0304	-0.0658	0.948	
year2012	-0.398	0.137	-2.91	0.00363	**
year2013	-0.0443	0.126	-0.35	0.726	
year2014	-0.0568	0.136	-0.418	0.676	
year2015	-0.323	0.143	-2.25	0.0242	*
$depth\_interval0-125$	-0.0528	0.219	-0.241	0.809	
$depth\_interval126-250$	-0.652	0.125	-5.24	1.65e-07	***
logret	0.345	0.0474	7.27	3.76e-13	***

Mean of squared residuals: 36.05839 % Var explained: 4.48



# ${\rm GAM}_{\rm Nonlinear}$

 $R\text{-sq.}(\mathrm{adj}) = 0.0407$  Deviance explained = 21% -REML = 27672 Scale est. = 55.784 n = 35440



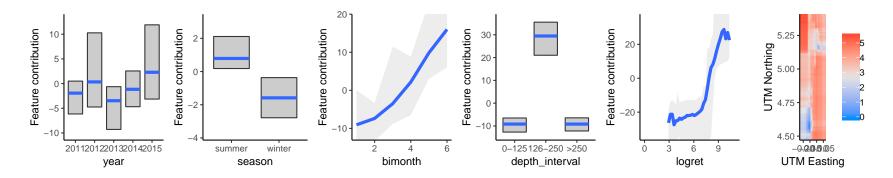
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	44.4	48	7.53	2.21e-49	***
s(year)	3.84	3.98	2.7	0.022	*
s(depth)	3.96	4	12.4	7.68e-10	***
$s(julian\_day)$	3.6	3.91	1.61	0.134	
s(time)	3.52	3.87	0.981	0.513	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-2.96	0.427	-6.95	3.71e-12	***
gear2	-0.644	0.272	-2.37	0.018	*
gear17	0.115	0.183	0.628	0.53	
logret	0.351	0.0475	7.39	1.49e-13	***

## Species 8: Pacific Hake

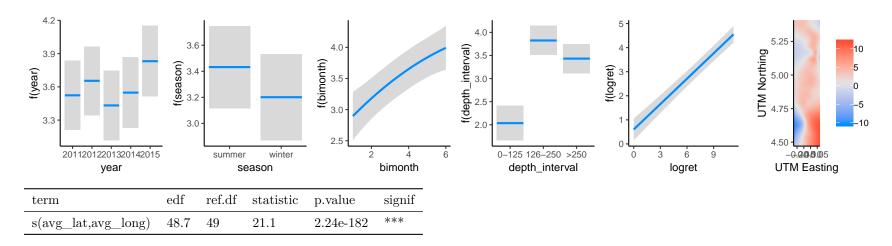
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 18788.2 % Var explained: 20.61



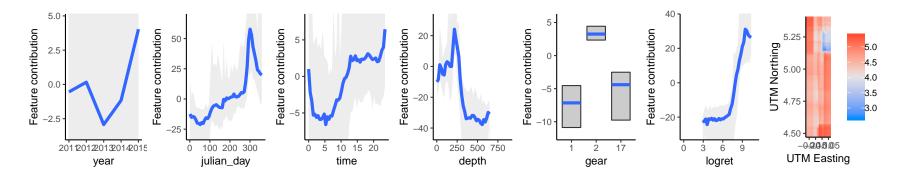
#### $GAM_{Total}$

 $R-sq.(adj) = 0.0796 \ Deviance \ explained = 22.8\% \ -REML = 1.2473e + 05 \ Scale \ est. = 49.425 \ n = 35440$ 



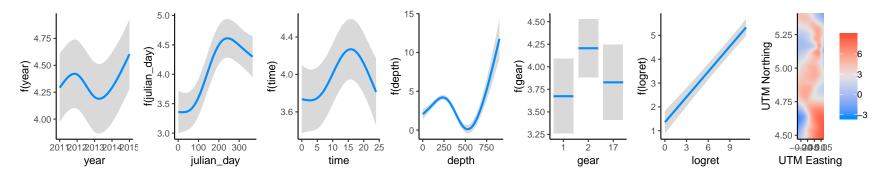
term	estimate	std.error	statistic	p.value	signif
(Intercept)	0.00278	0.277	0.01	0.992	
seasonwinter	-0.231	0.0825	-2.8	0.00511	**
bimonth	0.334	0.119	2.81	0.00496	**
I(bimonth^2)	-0.0164	0.0159	-1.03	0.301	
year2012	0.131	0.0643	2.03	0.0419	*
year2013	-0.0913	0.0637	-1.43	0.152	
year2014	0.0242	0.0684	0.354	0.723	
year2015	0.308	0.0679	4.54	5.78e-06	***
$depth\_interval0-125$	-1.4	0.117	-12	5.58e-33	***
$depth\_interval 126-250$	0.393	0.0567	6.93	4.36e-12	***
logret	0.352	0.0216	16.3	2.56e-59	***

Mean of squared residuals: 17934.13 % Var explained: 24.22



# ${\rm GAM_{Nonlinear}}$

 $R-sq.(adj) = -0.0354 \ Deviance \ explained = 27.4\% \ -REML = 1.2361e + 05 \ Scale \ est. \ = 48.638 \ n = 35440$ 



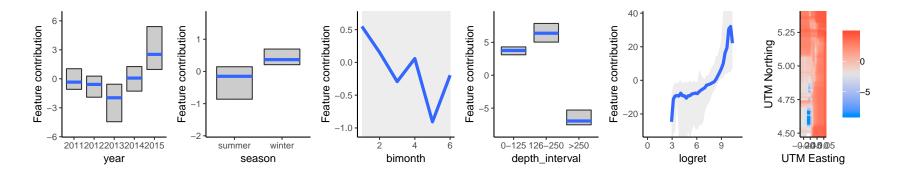
term	edf	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	48.5	49	15.9	2.96e-130	***
s(year)	3.89	3.99	10.5	1.19e-08	***
s(depth)	3.99	4	138	3.94e-117	***
$s(julian\_day)$	3.95	4	92.4	2.8e-78	***
s(time)	3.88	3.99	20.4	8.62e-17	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-0.0724	0.199	-0.363	0.716	
gear2	0.532	0.133	3.98	6.82 e-05	***
gear17	0.154	0.0848	1.81	0.0697	•
logret	0.353	0.0218	16.2	8.02e-59	***

## Species 9: Pacific Halibut

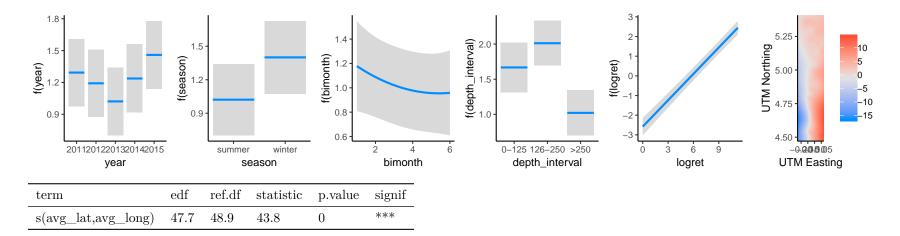
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 3041.369 % Var explained: 18.98



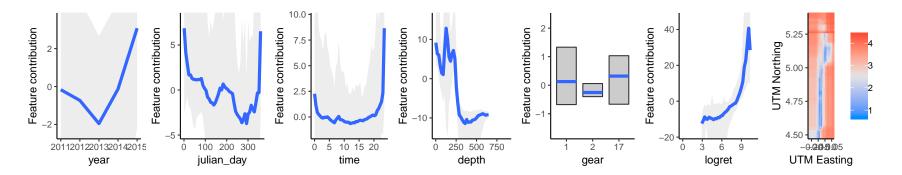
#### $GAM_{Total}$

R-sq.(adj) = 0.103 Deviance explained = 27.3% -REML = 74265 Scale est. = 44.927 n = 35440



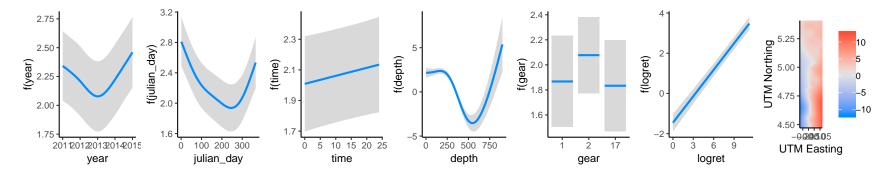
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.62	0.239	-6.78	1.2e-11	***
seasonwinter	0.379	0.0713	5.31	1.08e-07	***
bimonth	-0.124	0.0959	-1.3	0.195	
I(bimonth^2)	0.0115	0.0129	0.891	0.373	
year2012	-0.101	0.0565	-1.79	0.0742	
year2013	-0.271	0.0563	-4.82	1.47e-06	***
year 2014	-0.0556	0.0595	-0.935	0.35	
year2015	0.166	0.0576	2.89	0.00386	**
$depth\_interval0-125$	0.646	0.0996	6.49	8.95e-11	***
depth_interval126-250	0.99	0.0524	18.9	3.35e-79	***
logret	0.447	0.0203	22.1	3.85e-107	***

Mean of squared residuals: 2970.634 % Var explained: 20.86



# $\mathbf{GAM}_{\mathbf{Nonlinear}}$

R-sq.(adj) = 0.109 Deviance explained = 29.3% -REML = 73942 Scale est. = 41.378 n = 35440



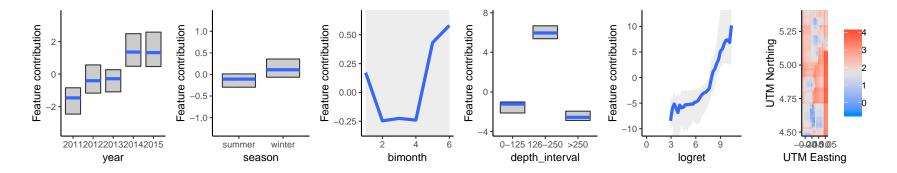
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.2	48.8	41.1	0	***
s(year)	3.59	3.9	15.1	5.61e-12	***
s(depth)	3.98	4	140	2.4e-119	***
$s(julian\_day)$	3.83	3.98	26.3	6.75 e-22	***
s(time)	1.01	1.02	3.54	0.0592	

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.55	0.178	-8.73	2.76e-18	***
gear2	0.21	0.105	2	0.0453	*
gear17	-0.0341	0.0602	-0.566	0.571	
logret	0.438	0.0196	22.4	3.17e-110	***

## Species 10: Sandpaper Skate

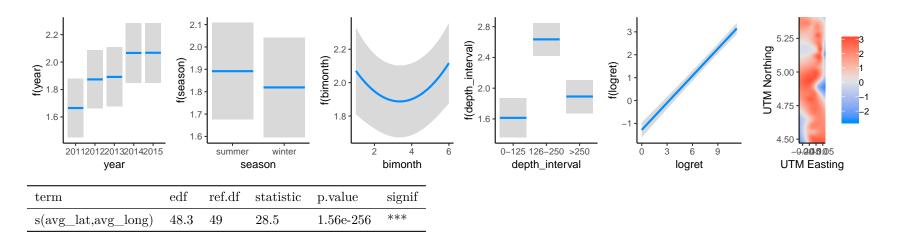
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 385.3107 % Var explained: 22.37



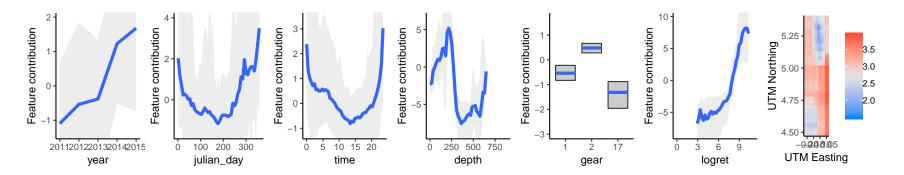
#### $GAM_{Total}$

R-sq.(adj) = 0.125 Deviance explained = 20% -REML = 92531 Scale est. = 20.414 n = 35440



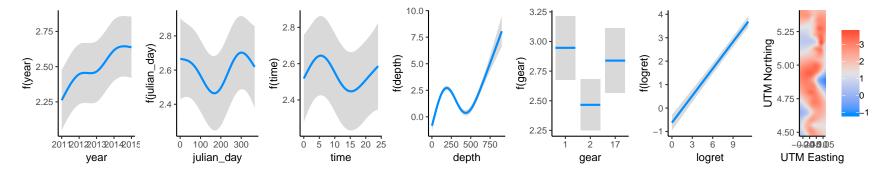
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.17	0.19	-6.13	8.83e-10	***
seasonwinter	-0.0725	0.0552	-1.31	0.189	
bimonth	-0.221	0.0756	-2.92	0.00345	**
I(bimonth <sup>2</sup> )	0.0329	0.0102	3.24	0.00121	**
year2012	0.209	0.0462	4.52	6.3e-06	***
year2013	0.227	0.0449	5.05	4.55e-07	***
year2014	0.401	0.0471	8.52	1.64e-17	***
year2015	0.403	0.0476	8.47	2.65e-17	***
$depth\_interval0-125$	-0.278	0.0803	-3.46	0.00054	***
depth_interval126-250	0.744	0.0387	19.2	8.38e-82	***
logret	0.394	0.0162	24.3	2.52e-129	***

Mean of squared residuals: 371.2104 % Var explained: 25.21



# ${\rm GAM}_{\rm Nonlinear}$

 $R\text{-sq.}(\mathrm{adj}) = 0.0892$  Deviance explained = 23.4% -REML = 91802 Scale est. = 19.723 n = 35440



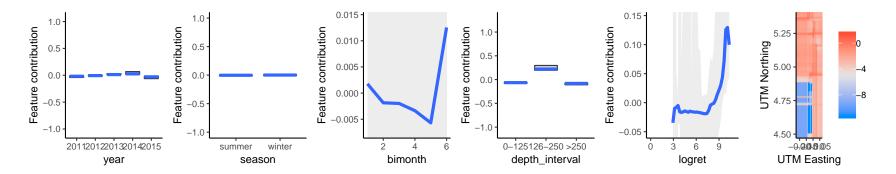
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	26.7	4.88e-238	***
s(year)	3.76	3.97	20.7	6.69e-17	***
s(depth)	4	4	277	8.25e-236	***
$s(julian\_day)$	3.83	3.98	6.43	2.58e-05	***
s(time)	3.73	3.96	5.01	0.000366	***

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-0.803	0.142	-5.67	1.41e-08	***
gear2	-0.481	0.0834	-5.77	8.02e-09	***
gear17	-0.108	0.0563	-1.92	0.0551	•
logret	0.383	0.016	23.9	1.8e-125	***

### Species 11: Rosethorn Rockfish

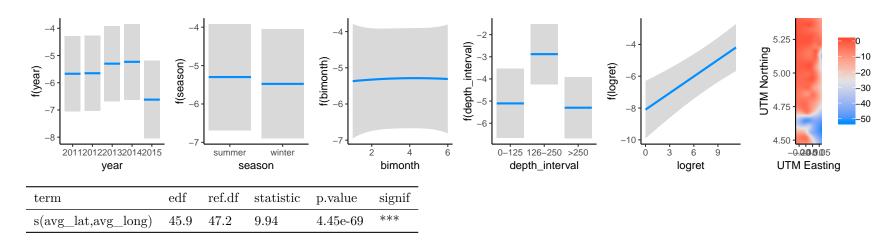
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 7.098818 % Var explained: 17.73



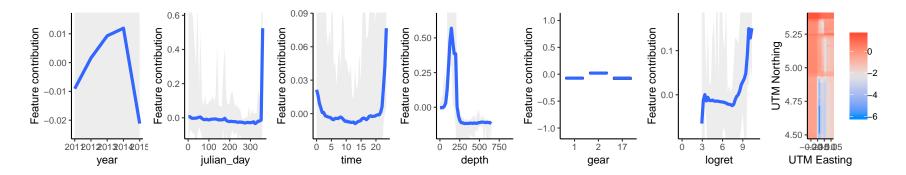
#### $GAM_{Total}$

R-sq.(adj) = 0.00397 Deviance explained = 52.8% -REML = 8403.7 Scale est. = 63.336 n = 35440



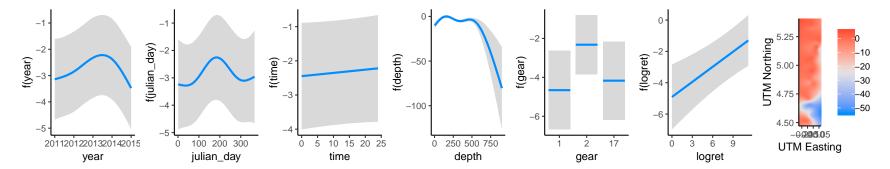
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-11.1	1.46	-7.61	2.86e-14	***
seasonwinter	-0.178	0.291	-0.611	0.541	
bimonth	0.0664	0.394	0.168	0.866	
I(bimonth^2)	-0.00776	0.0533	-0.146	0.884	
year2012	0.0185	0.253	0.0731	0.942	
year2013	0.371	0.234	1.58	0.114	
year2014	0.439	0.257	1.71	0.0873	
year2015	-0.951	0.301	-3.16	0.00157	**
$depth\_interval0-125$	0.195	0.443	0.44	0.66	
$depth\_interval 126-250$	2.42	0.211	11.4	2.85e-30	***
logret	0.348	0.0767	4.54	5.74e-06	***

Mean of squared residuals: 6.679766 % Var explained: 22.59



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = 0.0729 Deviance explained = 59.5% -REML = 8114.9 Scale est. = 77.733 n = 35440



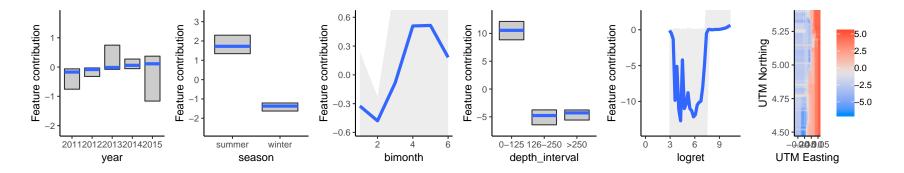
term	edf	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	45.6	47.1	6.42	2.83e-38	***
s(year)	3.68	3.94	5.16	0.000429	***
s(depth)	3.97	4	44.7	1.72e-37	***
$s(julian\_day)$	3.77	3.97	2.44	0.0378	*
s(time)	1.01	1.01	0.461	0.499	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-12.6	1.58	-7.99	1.44e-15	***
gear2	2.33	0.684	3.41	0.00064	***
gear17	0.487	0.487	0.999	0.318	
logret	0.321	0.0889	3.61	0.000306	***

### Species 12: Slender Sole

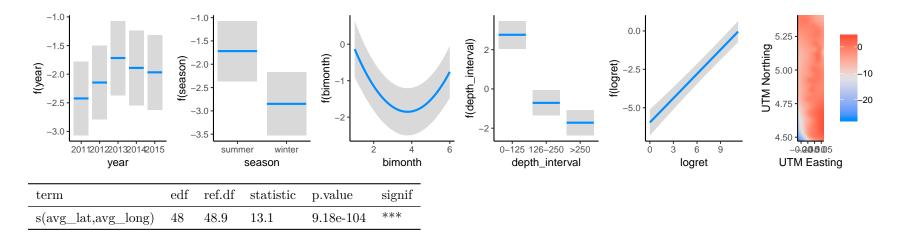
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 2193.15 % Var explained: 28.43



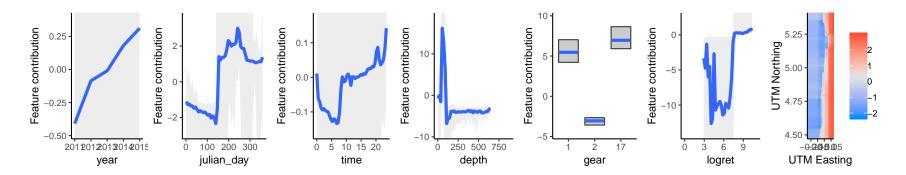
#### $GAM_{Total}$

 $R-sq.(adj) = 0.129 \ Deviance \ explained = 47.3\% \ -REML = 51504 \ Scale \ est. = 47.418 \ n = 35440$ 



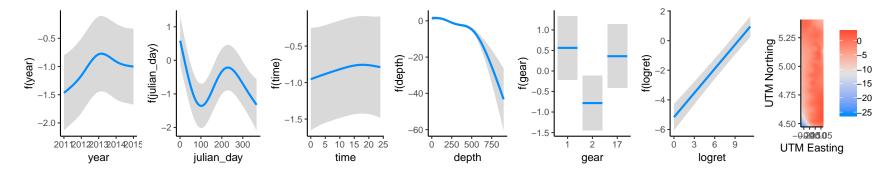
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-3.53	0.495	-7.12	1.08e-12	***
seasonwinter	-1.13	0.159	-7.11	1.14e-12	***
bimonth	-1.68	0.218	-7.69	1.47e-14	***
I(bimonth^2)	0.222	0.0292	7.62	2.52e-14	***
year2012	0.278	0.114	2.44	0.0147	*
year2013	0.705	0.11	6.44	1.23e-10	***
year2014	0.533	0.116	4.57	4.78e-06	***
year2015	0.456	0.117	3.88	0.000103	***
$depth\_interval0-125$	4.47	0.199	22.5	4.96e-111	***
depth_interval126-250	1.01	0.122	8.28	1.31e-16	***
logret	0.526	0.0359	14.7	1.51e-48	***

Mean of squared residuals: 2006.376 % Var explained: 34.53



# ${\rm GAM}_{\rm Nonlinear}$

 $R-sq.(adj) = 0.144 \ Deviance \ explained = 48.6\% \ -REML = 51284 \ Scale \ est. \ = 51.074 \ n = 35440$ 



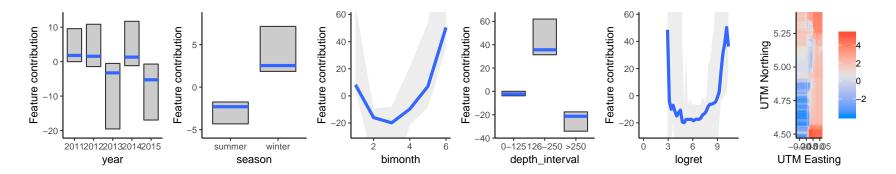
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	10.2	1.47e-75	***
s(year)	3.73	3.96	10.1	4.15e-08	***
s(depth)	3.95	4	59.9	2.91e-50	***
$s(julian\_day)$	3.98	4	29.4	1.56e-24	***
s(time)	2.05	2.53	0.809	0.434	

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-3.89	0.341	-11.4	4.17e-30	***
gear2	-1.34	0.211	-6.37	1.93e-10	***
gear17	-0.203	0.101	-2	0.0452	*
logret	0.544	0.038	14.3	2.5e-46	***

## Species 13: Spiny Dogfish Shark

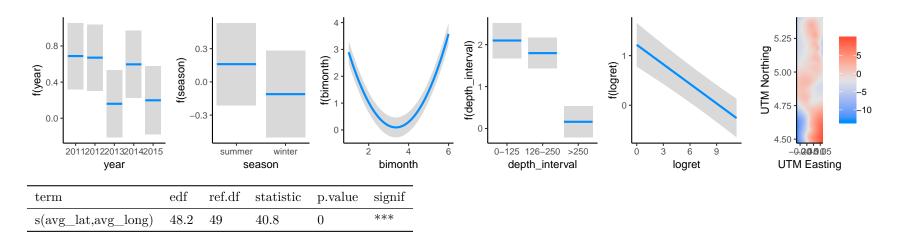
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 444124.3 % Var explained: 25.06



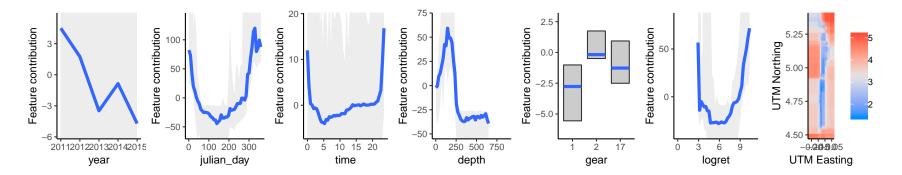
#### $GAM_{Total}$

 $R-sq.(adj) = 0.0248 \ Deviance \ explained = 40.6\% \ -REML = 1.0483e + 05 \ Scale \ est. = 64.451 \ n = 35440$ 



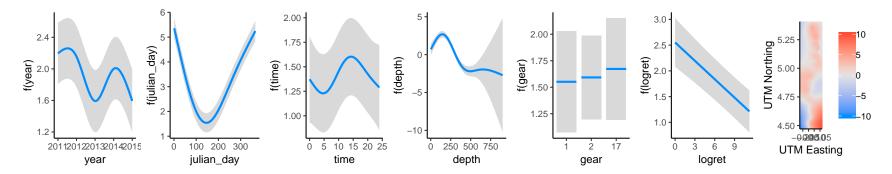
term	estimate	std.error	statistic	p.value	signif
(Intercept)	7.67	0.255	30.1	2.23e-196	***
seasonwinter	-0.271	0.0944	-2.87	0.00412	**
bimonth	-3.37	0.124	-27.2	1.46e-161	***
I(bimonth^2)	0.501	0.0169	29.7	5.82e-192	***
year2012	-0.0176	0.0692	-0.255	0.799	
year2013	-0.528	0.069	-7.65	1.98e-14	***
year2014	-0.0918	0.0724	-1.27	0.204	
year2015	-0.49	0.0753	-6.5	8.04e-11	***
$depth\_interval0-125$	1.94	0.124	15.6	8.99e-55	***
$depth\_interval 126-250$	1.64	0.0654	25	6.35e-137	***
logret	-0.131	0.0158	-8.31	9.88e-17	***

Mean of squared residuals: 468651 % Var explained: 20.92



# ${\rm GAM}_{\rm Nonlinear}$

R-sq.(adj) = 0.0454 Deviance explained = 43.8% -REML = 1.0398e+05 Scale est. = 69.251 n = 35440



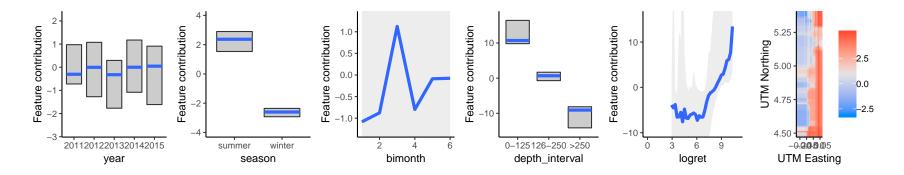
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.8	48.9	30.2	4.04e-273	***
s(year)	3.98	4	30.3	3.05e-25	***
s(depth)	3.96	4	218	2.39e-185	***
$s(julian\_day)$	3.99	4	514	0	***
s(time)	3.84	3.98	6.37	3.26 e - 05	***

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	3.65	0.168	21.8	2.35e-104	***
gear2	0.0409	0.142	0.288	0.773	
gear17	0.121	0.0833	1.45	0.147	
logret	-0.119	0.0167	-7.15	8.76e-13	***

### Species 14: Spotted Ratfish

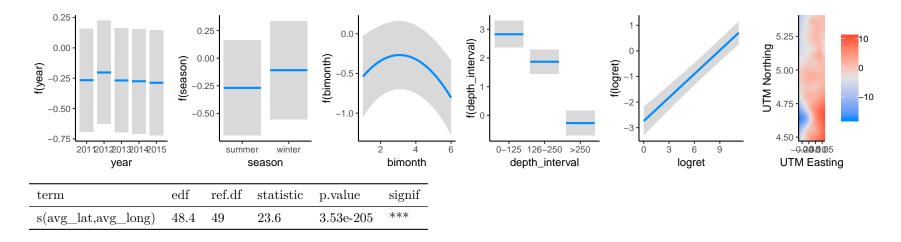
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 6941.341 % Var explained: 7.86



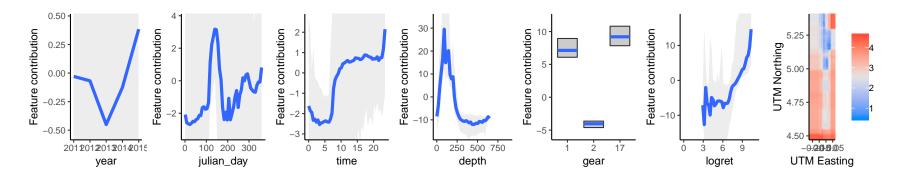
#### $GAM_{Total}$

 $R-sq.(adj) = 0.0237 \; Deviance \; explained = 34.4\% \; -REML = 88372 \; Scale \; est. \; = 46.446 \; n = 35440 \; respectively.$ 



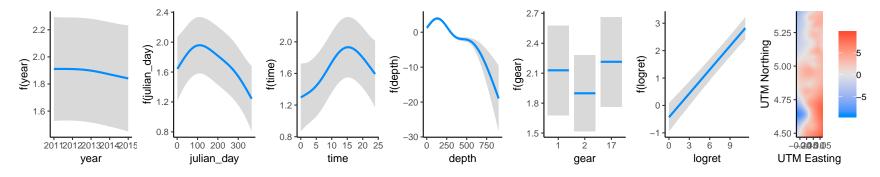
term	estimate	std.error	statistic	p.value	signif
(Intercept)	-2.69	0.332	-8.09	5.94e-16	***
seasonwinter	0.16	0.103	1.56	0.118	
bimonth	0.388	0.146	2.66	0.00777	**
I(bimonth <sup>2</sup> )	-0.063	0.0194	-3.25	0.00115	**
year2012	0.0635	0.0759	0.836	0.403	
year2013	-0.00229	0.0746	-0.0306	0.976	
year2014	-0.00898	0.0789	-0.114	0.909	
year2015	-0.0214	0.0793	-0.27	0.787	
$depth\_interval0-125$	3.1	0.131	23.6	5.54e-122	***
$depth\_interval126-250$	2.14	0.0778	27.5	2.39e-164	***
logret	0.307	0.024	12.8	1.76e-37	***

Mean of squared residuals: 6596.819 % Var explained: 12.44



### ${\rm GAM_{Nonlinear}}$

 $R\text{-sq.}(\mathrm{adj}) = 0.0547$  Deviance explained = 39.8% -REML = 87121 Scale est. = 38.584 n = 35440



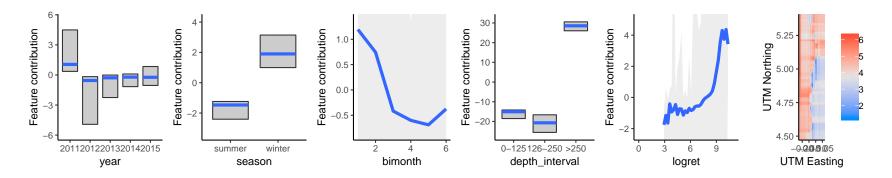
term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	48.1	49	22.1	7.34e-190	***
s(year)	1.7	2.09	0.693	0.495	
s(depth)	3.98	4	375	5.5399580868179e-320	***
$s(julian\_day)$	3.65	3.93	12.4	5.92e-10	***
s(time)	3.73	3.96	16.3	3.7e-13	***

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-0.557	0.198	-2.81	0.00489	**
gear2	-0.23	0.125	-1.84	0.0657	•
gear17	0.0856	0.0664	1.29	0.197	
logret	0.289	0.022	13.1	2.52e-39	***

## Species 15: Tanneri Tanner Crab

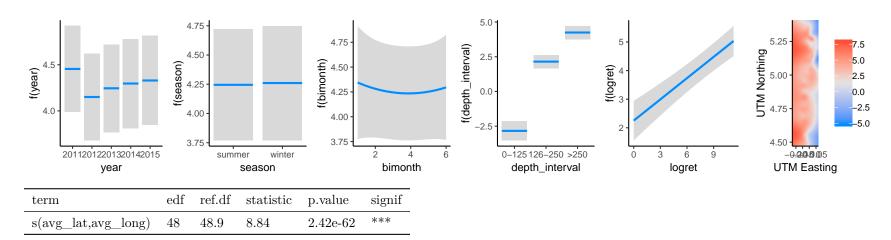
 $\mathbf{RF}_{\mathbf{Total}}$ 

Mean of squared residuals: 7438.715 % Var explained: 34.52



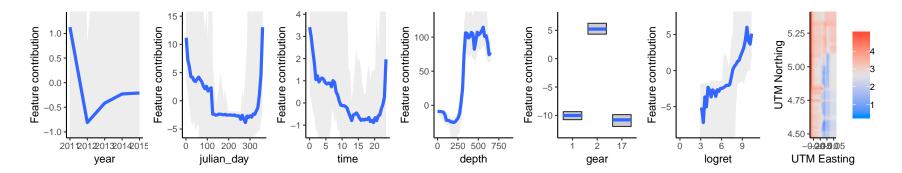
#### $GAM_{Total}$

 $R\text{-sq.}(\mathrm{adj}) = 0.176$  Deviance explained = 53.9% -REML = 85417 Scale est. = 85.96 n = 35440



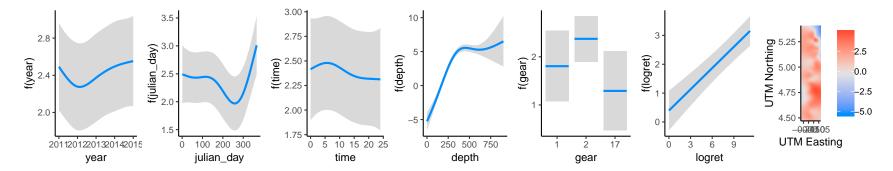
term	estimate	std.error	statistic	p.value	signif
(Intercept)	1.73	0.398	4.35	1.38e-05	***
seasonwinter	0.0155	0.121	0.128	0.898	
bimonth	-0.103	0.162	-0.637	0.524	
I(bimonth <sup>2</sup> )	0.0134	0.022	0.609	0.543	
year2012	-0.304	0.105	-2.9	0.00378	**
year2013	-0.21	0.101	-2.07	0.0386	*
year2014	-0.158	0.11	-1.43	0.153	
year2015	-0.125	0.113	-1.1	0.271	
$depth\_interval0-125$	-7.08	0.273	-26	1.97e-147	***
$depth\_interval126-250$	-2.1	0.0956	-21.9	7.42e-106	***
logret	0.248	0.0332	7.46	9.18e-14	***

Mean of squared residuals: 6860.649 % Var explained: 39.61



# ${\rm GAM}_{\rm Nonlinear}$

 $R-sq.(adj) = 0.245 \ Deviance \ explained = 60.1\% \ -REML = 83562 \ Scale \ est. = 75.678 \ n = 35440$ 



term	$\operatorname{edf}$	$\operatorname{ref.df}$	statistic	p.value	signif
s(avg_lat,avg_long)	47.6	48.9	7.14	1.41e-46	***
s(year)	3.65	3.93	2.09	0.0743	
s(depth)	3.98	4	267	2.71e-227	***
$s(julian\_day)$	3.95	4	7.12	1.18e-05	***
s(time)	3.07	3.56	0.968	0.389	

term	estimate	$\operatorname{std.error}$	statistic	p.value	signif
(Intercept)	-2.08	0.347	-5.99	2.1e-09	***
gear2	0.568	0.291	1.95	0.0507	•
gear17	-0.514	0.33	-1.56	0.12	
logret	0.246	0.0322	7.63	2.35e-14	***