RF and GAM model summaries

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Random forest and GAM models

For each (of 15) species we fit the following models. `fac(x)` is shorthand for treating covariate x as a factor. All random forest models were fit with `ntree=500`. 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

$\mathbf{RF_{strat}}$

Designed to *mimic the stratified ratio estimator* by treating `year`, `season`, and `depth_interval` as factors.

randomForest(Y ~ fac(year) + fac(season) + fac(depth_interval) + bimonth + I(bimonth^2) + logret + lat + long + I(lat^2) + I(long^2))

$\mathbf{GAM_{strat}}$

Designed to *mimic the stratified ratio estimator* by treating `year`, `season`, and `depth_interval` as factors.

gam(Y ~ fac(year) + fac(season) + fac(depth_interval) + bimonth + I(bimonth^2) + logret + s(lat, long, k=50))

$\mathbf{RF_{strat2}}$

Also designed to mimic the stratified ratio estimator, but now *treat* ***`bimonth`*** *as factor and remove* ***`season`*** to be consistent with ratio estimator in paper.

randomForest(Y ~ fac(year) + fac(bimonth) + fac(depth_interval) + logret + lat + long + I(lat^2) + I(long^2))

$\mathbf{GAM_{strat2}}$

Also designed to mimic the stratified ratio estimator, but now *treat* `bimonth` *as factor and remove* `season` to be consistent with ratio estimator in paper.

gam(Y ~ fac(year) + fac(bimonth) + fac(depth_interval) + logret + s(lat, long, k=50))

$\mathbf{RF_{nonlinear}}$

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

randomForest(Y ~ 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)

$\mathbf{GAM_{nonlinear}}$

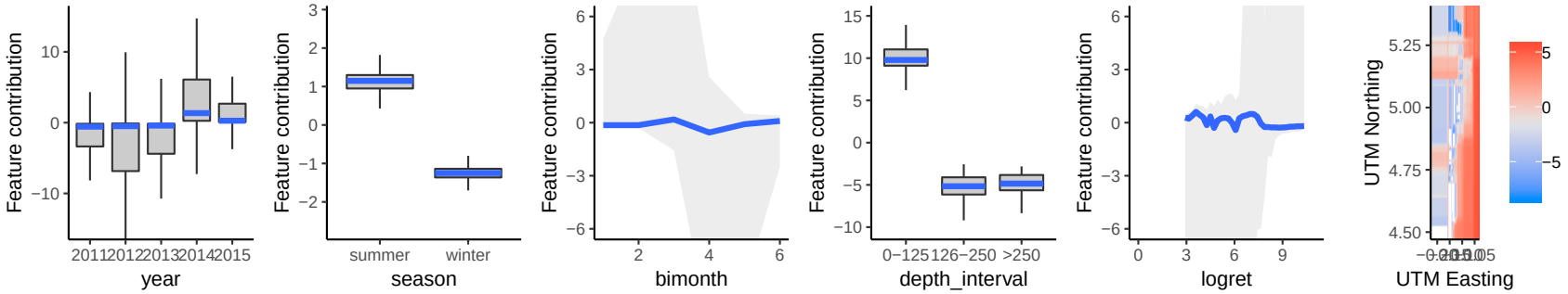
Allow GAM to fit covariates with *full non-linear flexibility* (not as factors):

gam(Y ~ 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

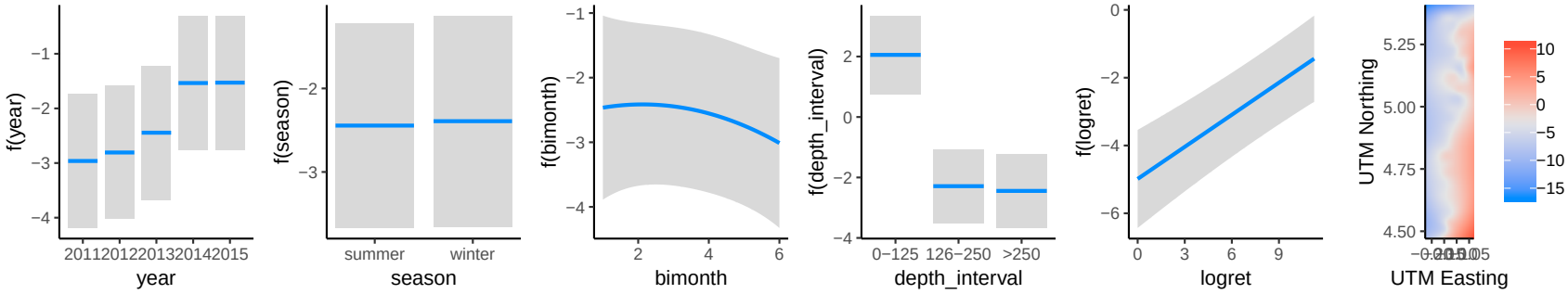
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 1989.101
          % Var explained: 47.8
```



$\mathbf{GAM_{strat1}}$

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

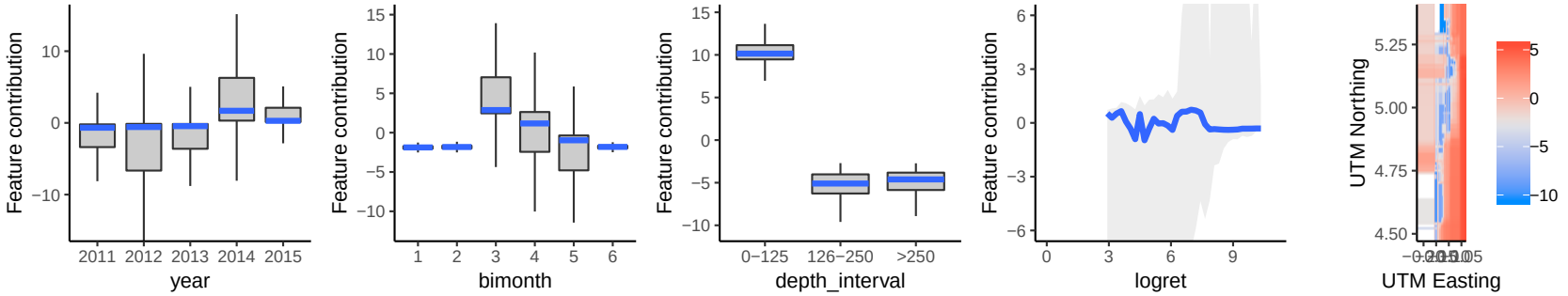


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	46.2	48.5	15.1	8.66e-121	***

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_interval126-250	0.157	0.262	0.598	0.55	
logret	0.315	0.05	6.32	2.73e-10	***

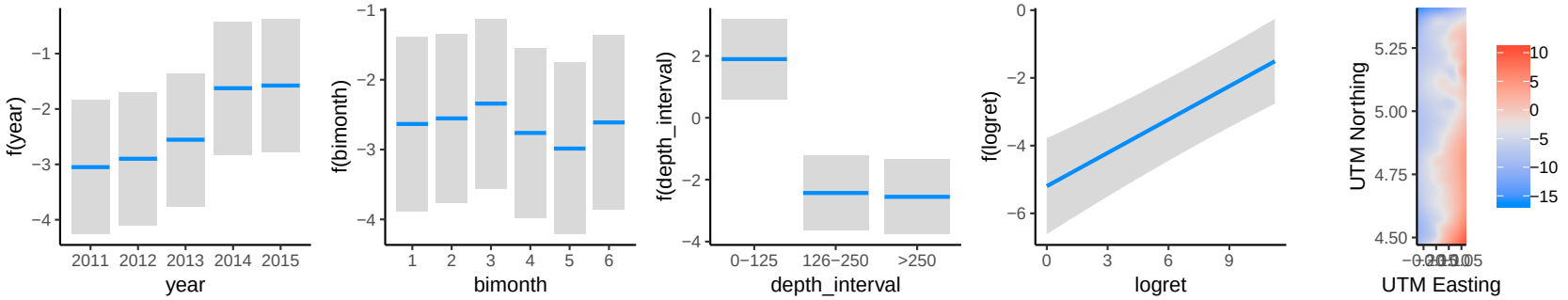
RF_{strat2}

```
Mean of squared residuals: 1998.454
          % Var explained: 47.56
```



GAM_{strat2}

R-sq.(adj) = 0.23 Deviance explained = 66% -REML = 32535 Scale est. = 97.865 n = 35440

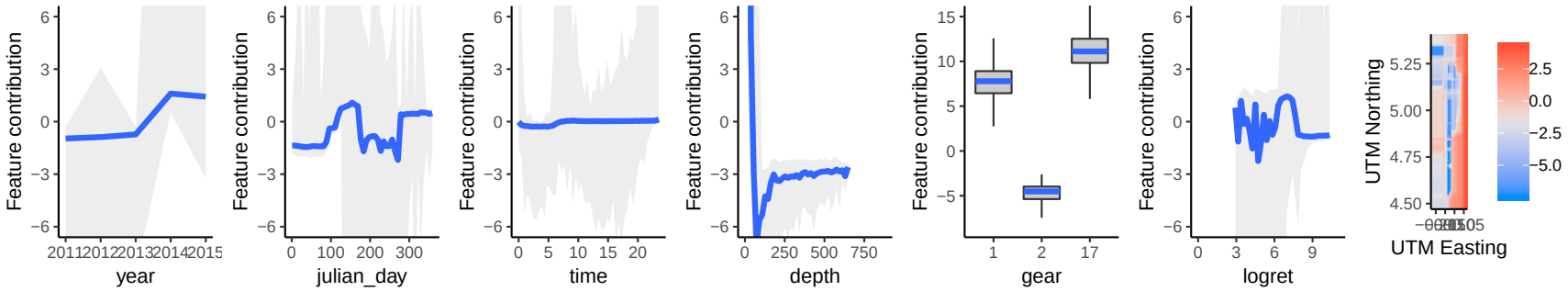


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	46	48.5	15.5	3.95e-125	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-6.05	0.472	-12.8	1.31e-37	***
bimonth2	0.0799	0.257	0.311	0.756	
bimonth3	0.292	0.238	1.23	0.22	
bimonth4	-0.128	0.242	-0.53	0.596	
bimonth5	-0.352	0.251	-1.4	0.16	
bimonth6	0.0223	0.29	0.077	0.939	
year2012	0.152	0.155	0.981	0.327	
year2013	0.496	0.162	3.07	0.00217	**
year2014	1.43	0.152	9.35	9.08e-21	***
year2015	1.47	0.157	9.41	5.09e-21	***
depth_interval0-125	4.45	0.339	13.1	2.98e-39	***
depth_interval126-250	0.124	0.257	0.481	0.631	
logret	0.328	0.0491	6.68	2.48e-11	***

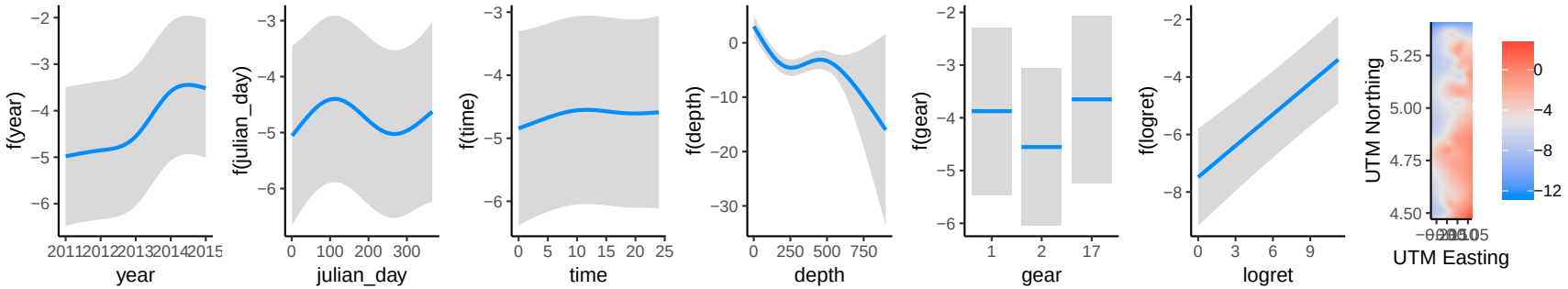
$RF_{nonlinear}$

```
Mean of squared residuals: 1784.603
          % Var explained: 53.17
```



GAM$_{\text{nonlinear}}$

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



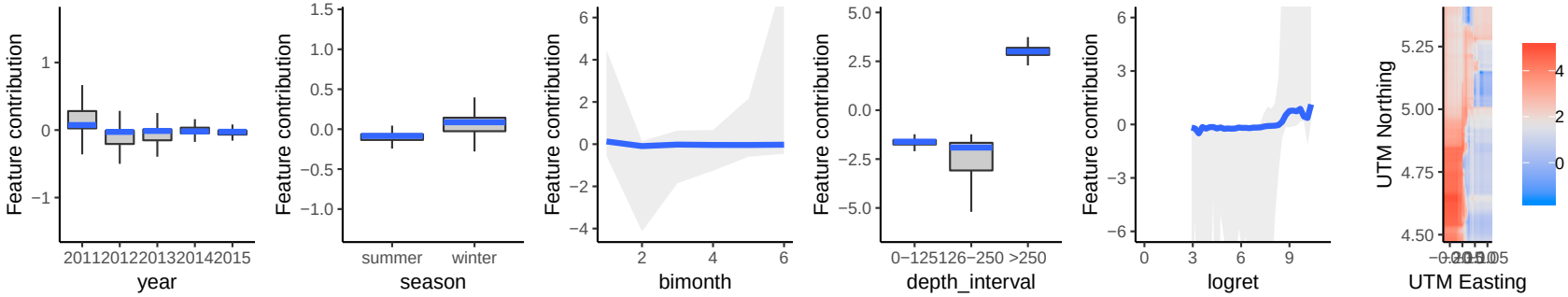
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	45.6	48.4	9.01	5.26e-63	***
s(year)	3.91	3.99	41.7	7.89e-35	***
s(depth)	3.89	3.99	31.5	3.1e-26	***
s(julian_day)	3.75	3.96	5.17	0.000357	***
s(time)	2.63	3.17	0.455	0.708	

term	estimate	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

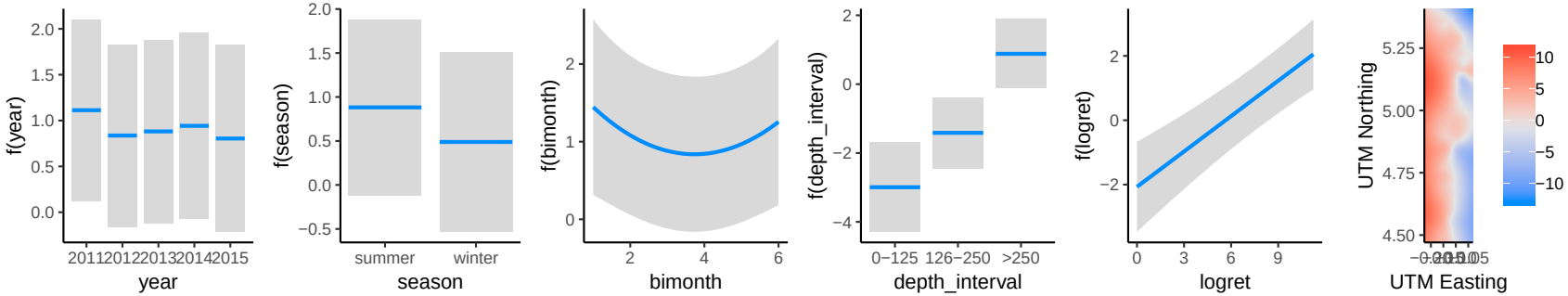
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 253.1007
          % Var explained: 48.13
```



$\mathbf{GAM_{strat1}}$

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

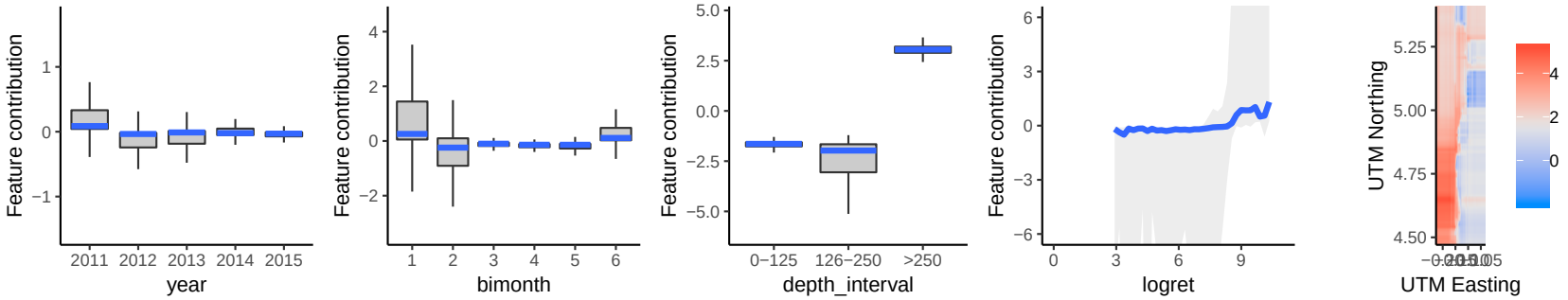


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47	48.7	8.53	7.16e-59	***

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_interval126-250	-2.29	0.193	-11.9	2.2e-32	***
logret	0.366	0.0634	5.77	7.85e-09	***

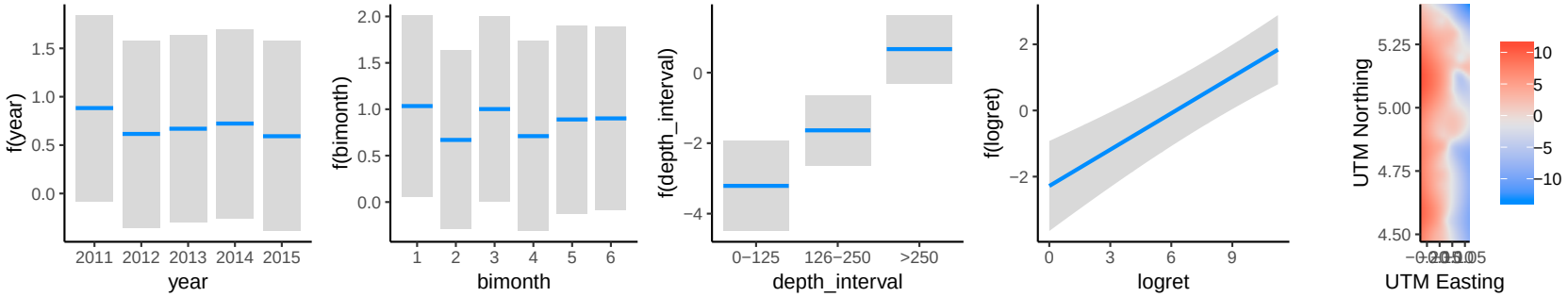
RF_{strat2}

```
Mean of squared residuals: 252.1983
          % Var explained: 48.32
```



GAM_{strat2}

R-sq.(adj) = -1.51 Deviance explained = 56.1% -REML = 33116 Scale est. = 132.49 n = 35440

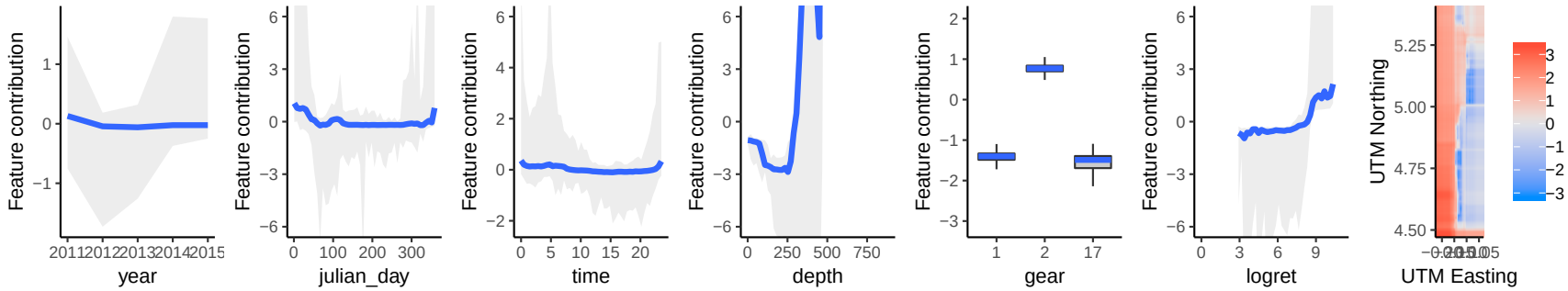


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47	48.7	9.02	2e-63	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.79	0.53	-3.37	0.000741	***
bimonth2	-0.365	0.163	-2.24	0.0251	*
bimonth3	-0.0326	0.188	-0.174	0.862	
bimonth4	-0.324	0.225	-1.44	0.15	
bimonth5	-0.145	0.214	-0.677	0.498	
bimonth6	-0.133	0.197	-0.676	0.499	
year2012	-0.267	0.167	-1.6	0.11	
year2013	-0.213	0.165	-1.3	0.195	
year2014	-0.16	0.176	-0.911	0.362	
year2015	-0.291	0.187	-1.56	0.119	
depth_interval0-125	-3.88	0.452	-8.59	9.37e-18	***
depth_interval126-250	-2.31	0.189	-12.2	2.69e-34	***
logret	0.366	0.0618	5.93	3.12e-09	***

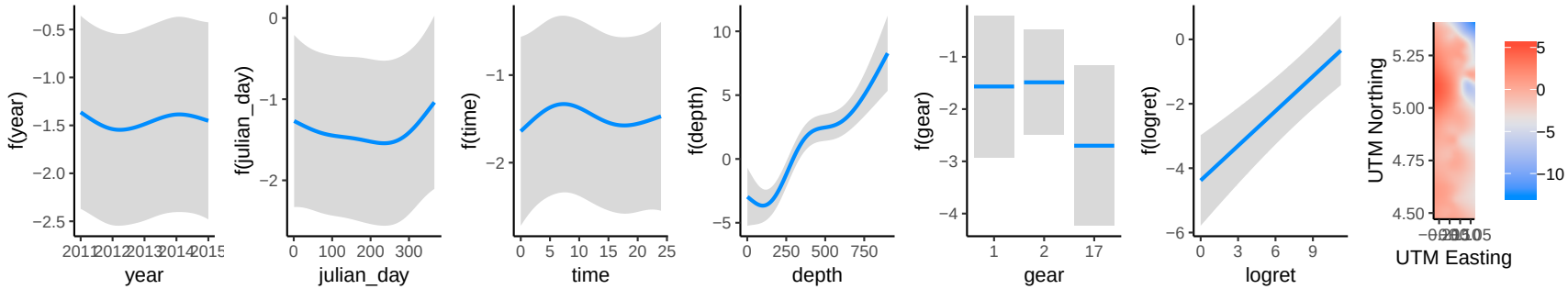
$RF_{nonlinear}$

```
Mean of squared residuals: 226.5094
          % Var explained: 53.58
```



GAM$_{\text{nonlinear}}$

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



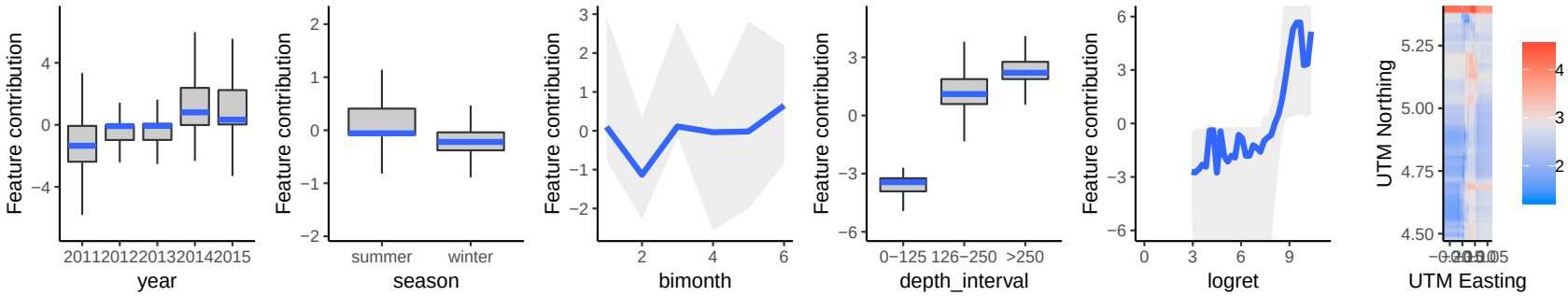
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	46.2	48.5	3.19	6.26e-13	***
s(year)	3.32	3.75	0.453	0.764	
s(depth)	3.99	4	92.9	1.66e-78	***
s(julian_day)	3.45	3.83	0.877	0.486	
s(time)	3.5	3.86	0.702	0.569	

term	estimate	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

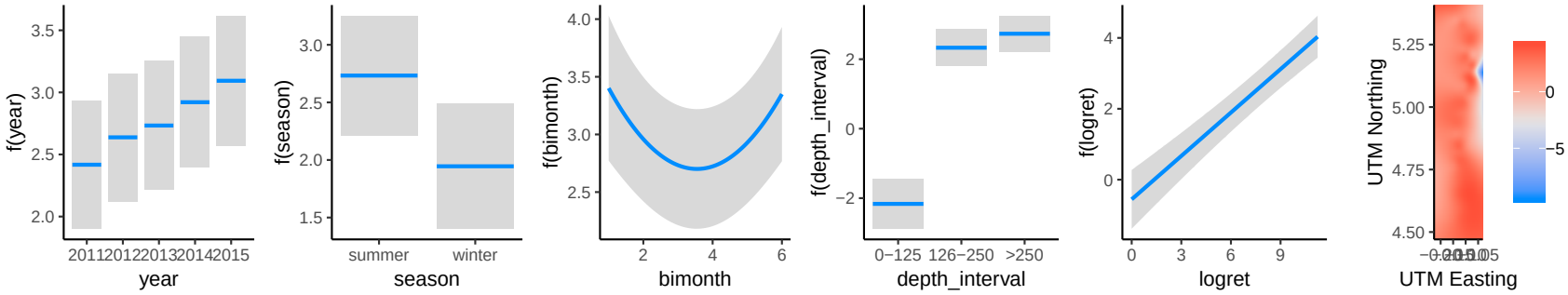
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 356.8372
          % Var explained: 20.8
```



$\mathbf{GAM_{strat1}}$

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

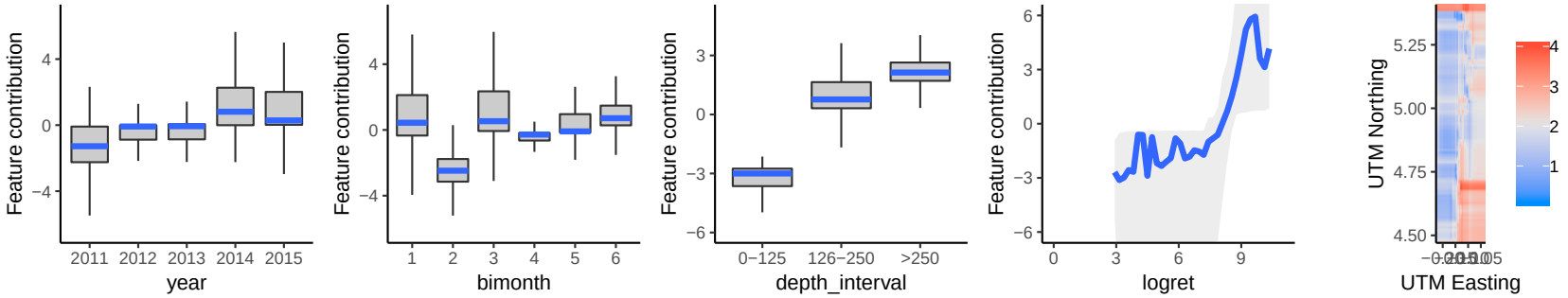


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.3	49	3.7	6.32e-17	***

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	1.46e-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	3.44e-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	***

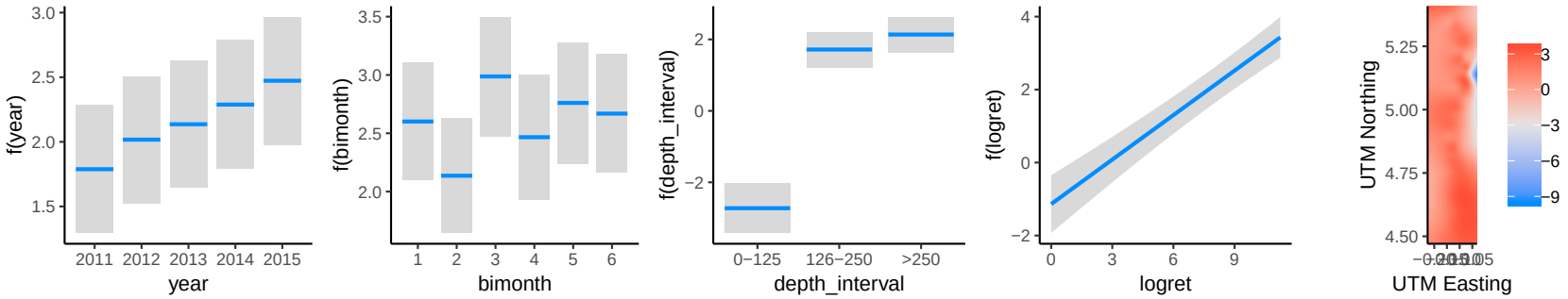
RF_{strat2}

```
Mean of squared residuals: 356.0772
          % Var explained: 20.97
```



GAM_{strat2}

R-sq.(adj) = 0.129 Deviance explained = 45.9% -REML = 74548 Scale est. = 80.654 n = 35440

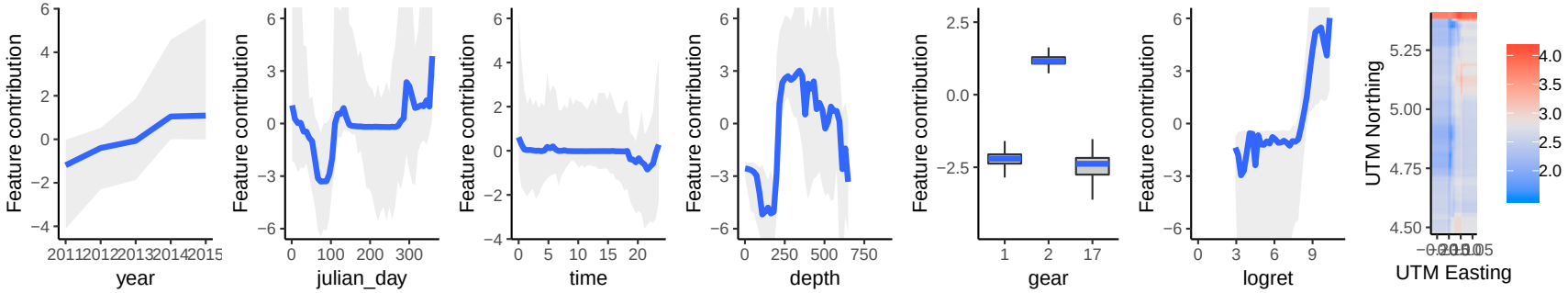


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.2	48.9	4.17	1.11e-20	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.63	0.351	-4.65	3.28e-06	***
bimonth2	-0.465	0.109	-4.26	2.02e-05	***
bimonth3	0.386	0.117	3.29	0.00101	**
bimonth4	-0.135	0.146	-0.921	0.357	
bimonth5	0.159	0.133	1.2	0.23	
bimonth6	0.0675	0.123	0.547	0.584	
year2012	0.229	0.113	2.02	0.0432	*
year2013	0.348	0.11	3.18	0.00149	**
year2014	0.5	0.116	4.32	1.59e-05	***
year2015	0.684	0.117	5.87	4.42e-09	***
depth_interval0-125	-4.86	0.26	-18.7	2.17e-77	***
depth_interval126-250	-0.419	0.0894	-4.69	2.71e-06	***
logret	0.406	0.0403	10.1	7.34e-24	***

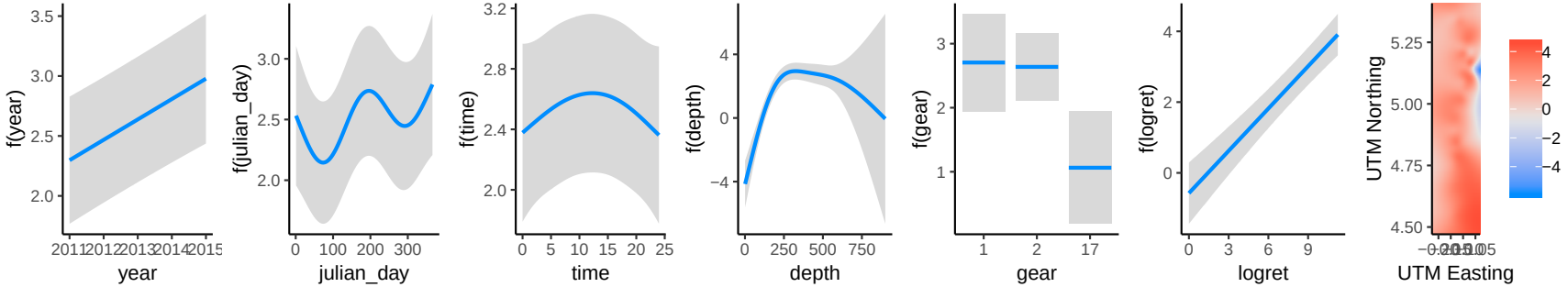
$RF_{nonlinear}$

```
Mean of squared residuals: 344.8937
          % Var explained: 23.45
```



GAM$_{\text{nonlinear}}$

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



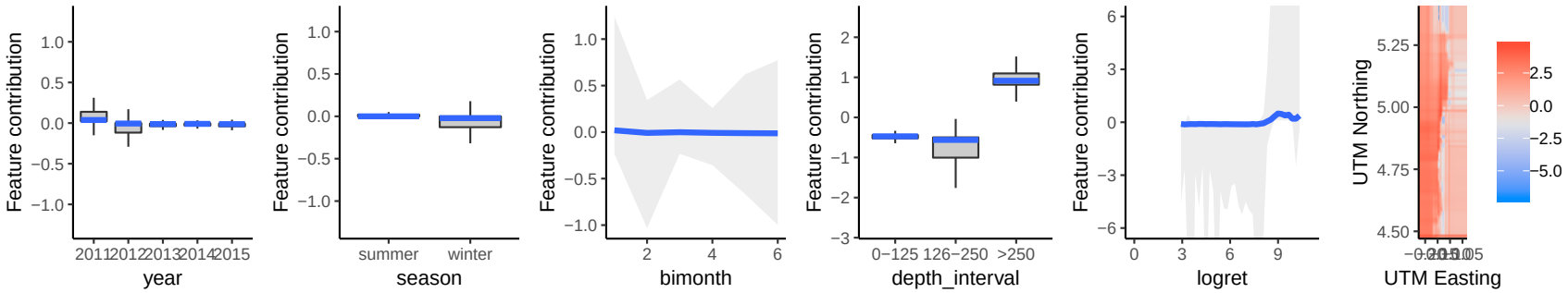
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	3.06	4.15e-12	***
s(year)	1	1.01	35.4	2.48e-09	***
s(depth)	3.98	4	35.1	2.62e-29	***
s(julian_day)	3.98	4	5.69	0.00014	***
s(time)	3.32	3.75	1.27	0.279	

term	estimate	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

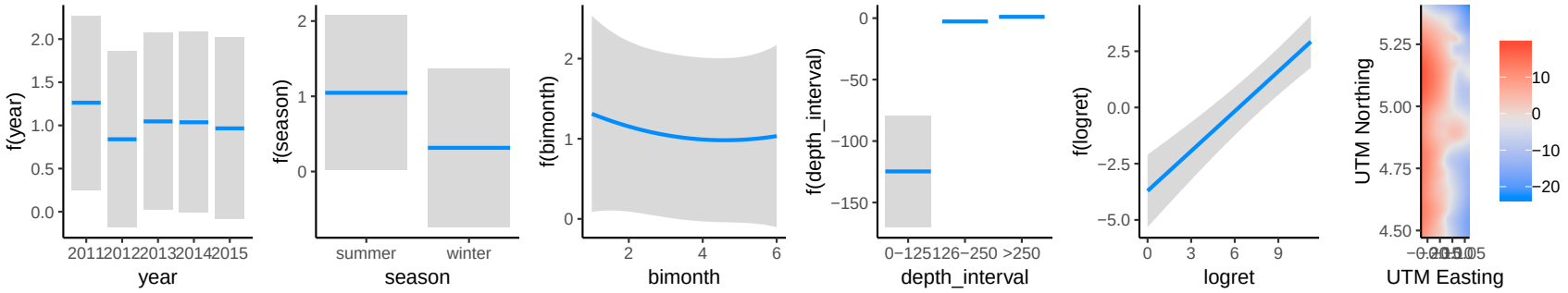
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 145.0971
          % Var explained: 42.62
```



$\mathbf{GAM_{strat1}}$

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

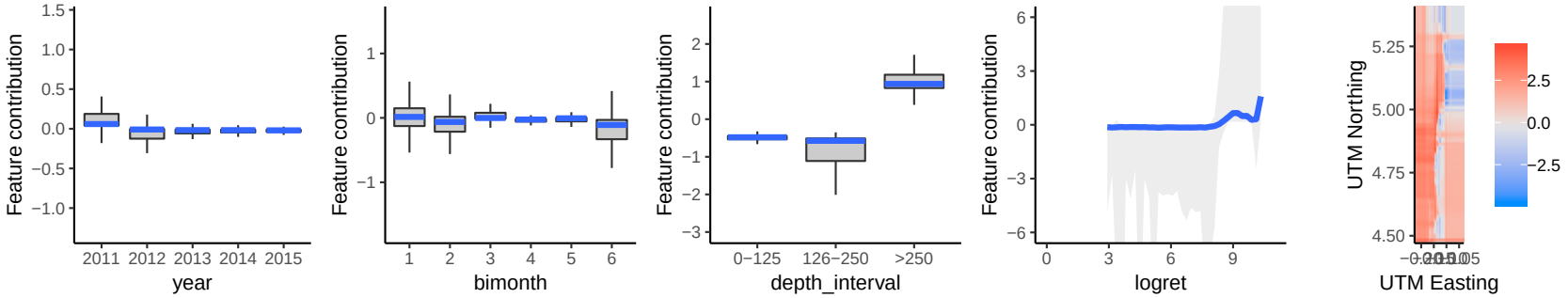


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	43.7	46.4	19.1	1.78e-153	***

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	***

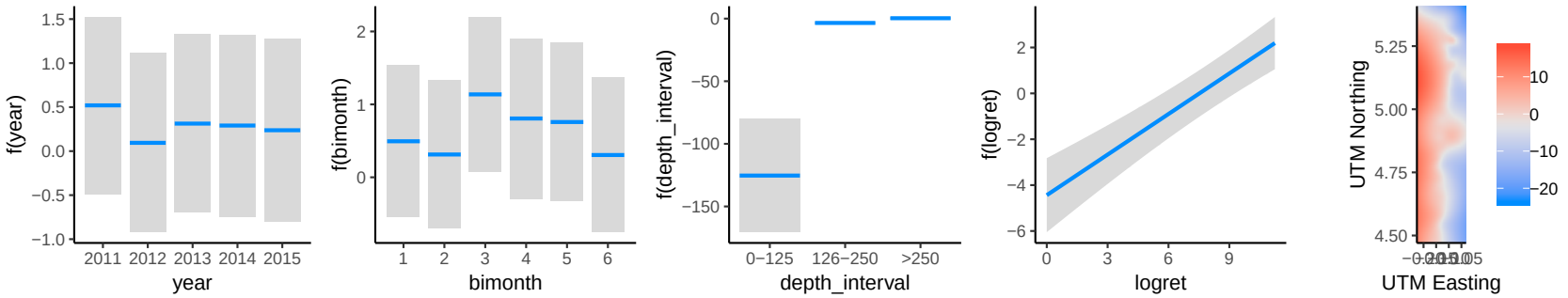
RF_{strat2}

```
Mean of squared residuals: 146.0836
          % Var explained: 42.23
```



GAM_{strat2}

R-sq.(adj) = -8.79e+03 Deviance explained = 65.8% -REML = 18890 Scale est. = 96.934 n = 35440

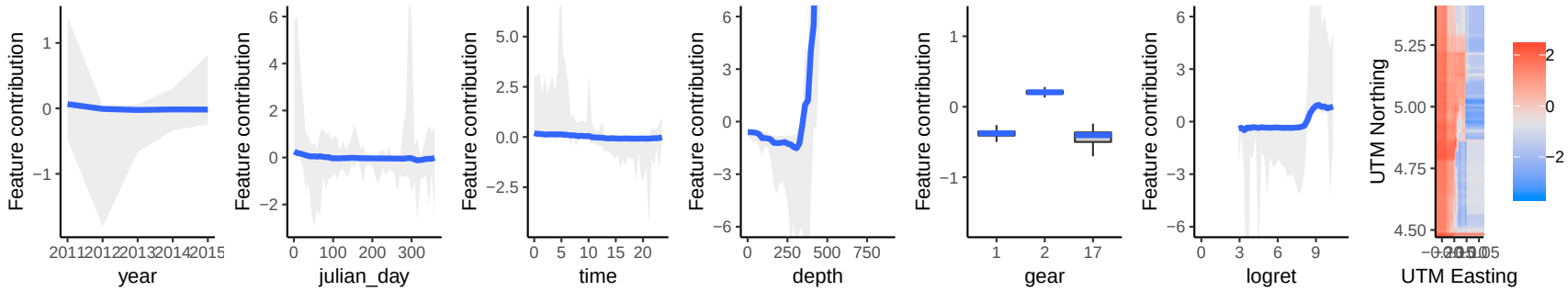


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	43.8	46.4	19.2	7.8e-155	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-6.27	0.792	-7.92	2.46e-15	***
bimonth2	-0.181	0.21	-0.863	0.388	
bimonth3	0.643	0.242	2.66	0.00786	**
bimonth4	0.312	0.284	1.1	0.272	
bimonth5	0.264	0.277	0.952	0.341	
bimonth6	-0.189	0.26	-0.725	0.468	
year2012	-0.426	0.214	-1.99	0.0464	*
year2013	-0.207	0.211	-0.984	0.325	
year2014	-0.229	0.229	-0.999	0.318	
year2015	-0.282	0.242	-1.17	0.243	
depth_interval0-125	-126	23.1	-5.45	5.1e-08	***
depth_interval126-250	-3.76	0.25	-15	6.4e-51	***
logret	0.59	0.0806	7.32	2.57e-13	***

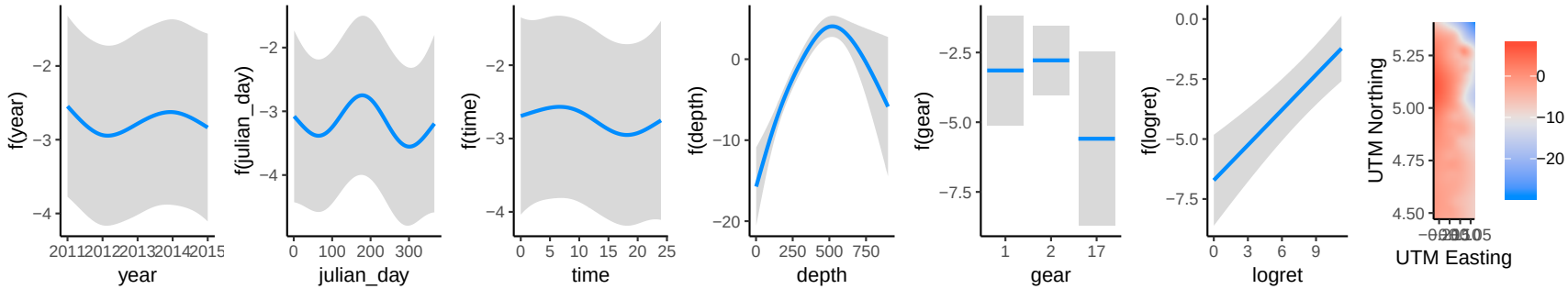
$RF_{nonlinear}$

```
Mean of squared residuals: 127.1781
          % Var explained: 49.7
```



GAM$_{\text{nonlinear}}$

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



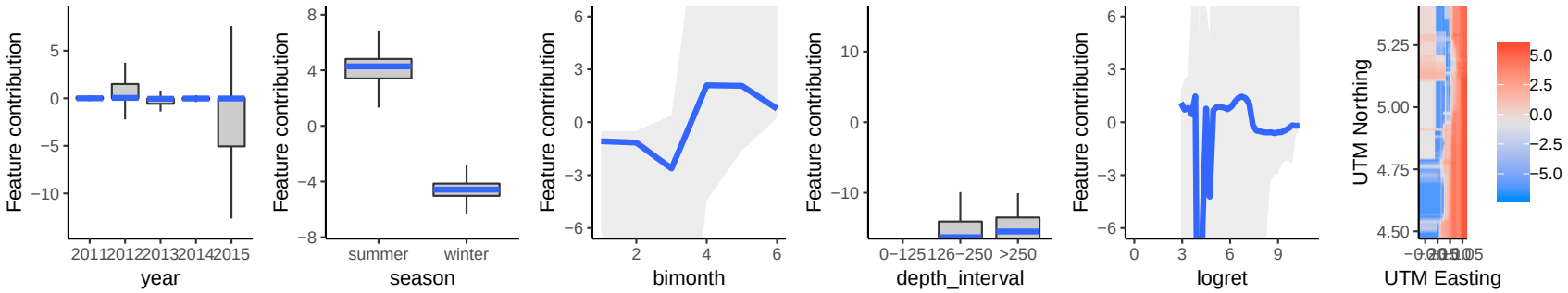
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	43	46.3	3.73	2e-16	***
s(year)	3.56	3.89	0.853	0.476	
s(depth)	3.94	4	120	4.75e-102	***
s(julian_day)	3.9	3.99	1.84	0.13	
s(time)	3.27	3.71	0.864	0.508	

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

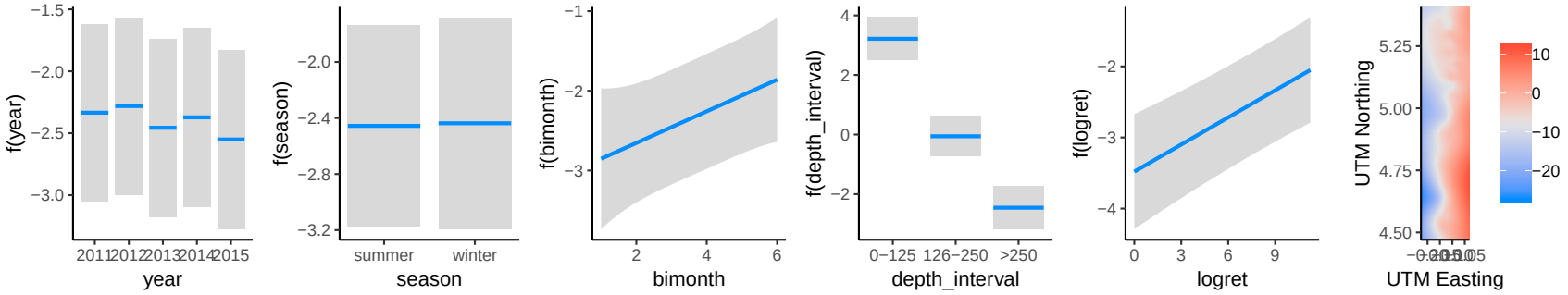
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 11727.1
          % Var explained: 28.81
```



$\mathbf{GAM_{strat1}}$

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

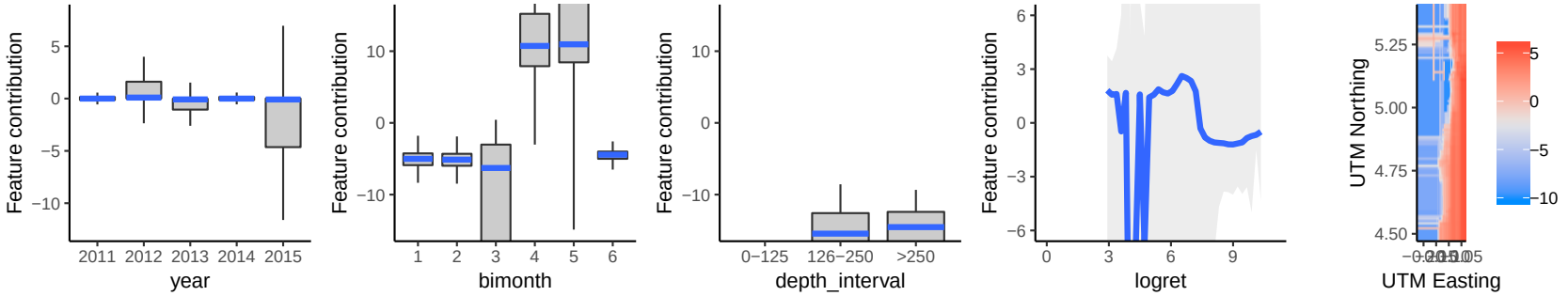


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	19.5	9.75e-166	***

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^2)	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	***

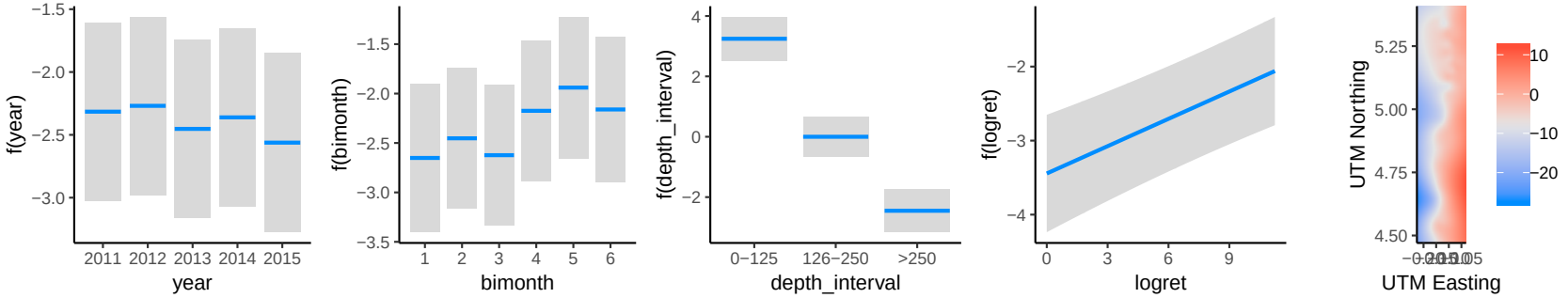
RF_{strat2}

```
Mean of squared residuals: 11658.05
          % Var explained: 29.23
```



GAM_{strat2}

R-sq.(adj) = 0.216 Deviance explained = 76.2% -REML = 65891 Scale est. = 53.774 n = 35440

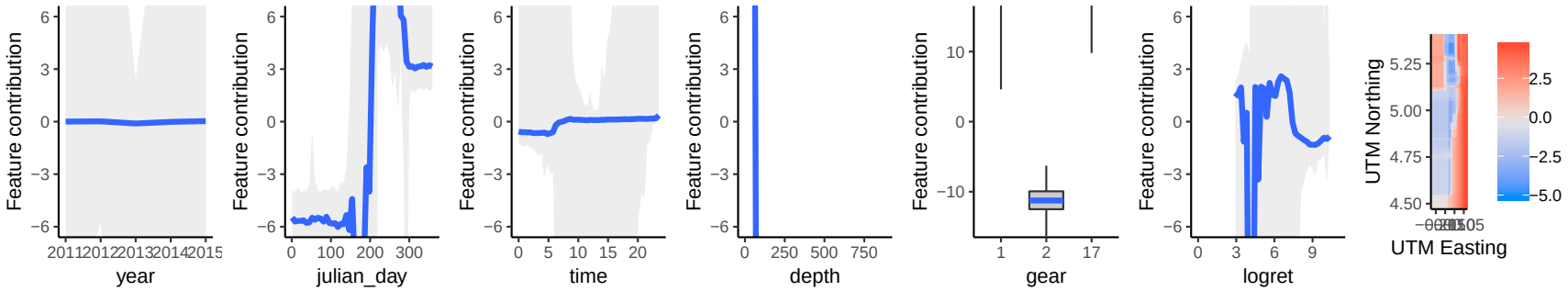


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	20.1	7.83e-172	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-4.96	0.287	-17.3	9.7e-67	***
bimonth2	0.199	0.19	1.04	0.296	
bimonth3	0.0281	0.175	0.161	0.872	
bimonth4	0.477	0.174	2.74	0.00607	**
bimonth5	0.712	0.176	4.05	5.08e-05	***
bimonth6	0.491	0.2	2.45	0.0143	*
year2012	0.0464	0.0799	0.581	0.562	
year2013	-0.137	0.0845	-1.62	0.104	
year2014	-0.0455	0.0868	-0.524	0.6	
year2015	-0.247	0.0893	-2.76	0.00575	**
depth_interval0-125	5.7	0.231	24.6	8.85e-133	***
depth_interval126-250	2.45	0.193	12.7	6.86e-37	***
logret	0.123	0.0243	5.06	4.24e-07	***

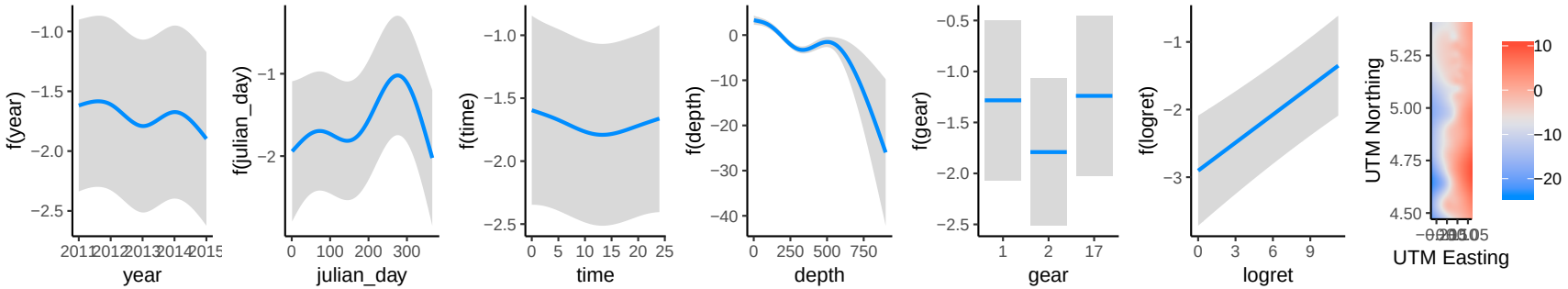
$RF_{nonlinear}$

```
Mean of squared residuals: 10642.64
          % Var explained: 35.39
```



GAM$_{\text{nonlinear}}$

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



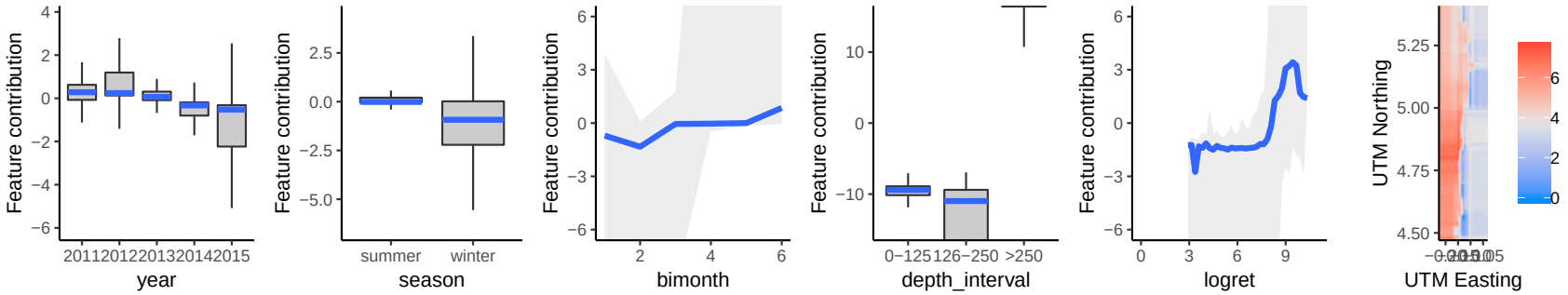
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.8	48.9	13.9	6.66e-111	***
s(year)	3.89	3.99	4.3	0.00226	**
s(depth)	3.97	4	104	1.55e-88	***
s(julian_day)	3.97	4	29.3	2.62e-24	***
s(time)	2.87	3.39	1.01	0.395	

term	estimate	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

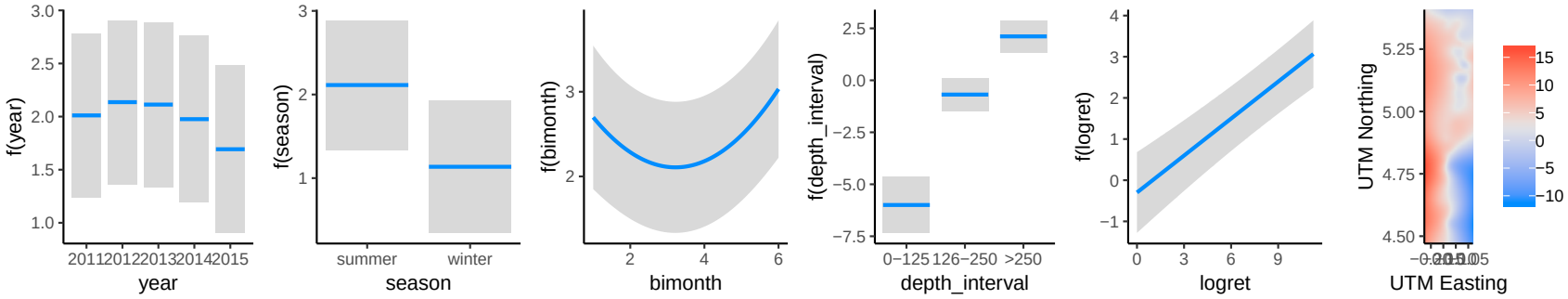
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 18071.56
          % Var explained: 49.32
```



$\mathbf{GAM_{strat1}}$

R-sq.(adj) = -241 Deviance explained = 66.8% -REML = 63079 Scale est. = 133.84 n = 35440

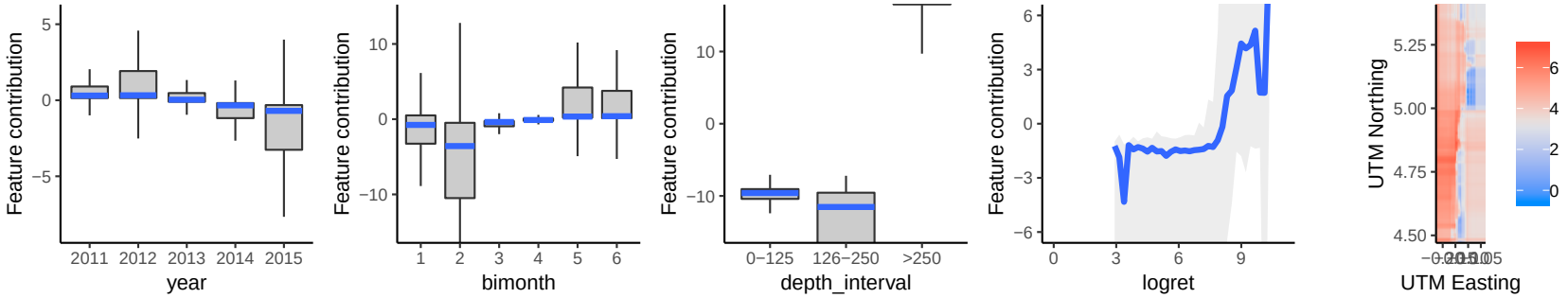


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	34.4	8.71000001824745e-316	***

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	1.95e-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_interval126-250	-2.8	0.137	-20.5	7.95e-93	***
logret	0.299	0.0394	7.6	2.95e-14	***

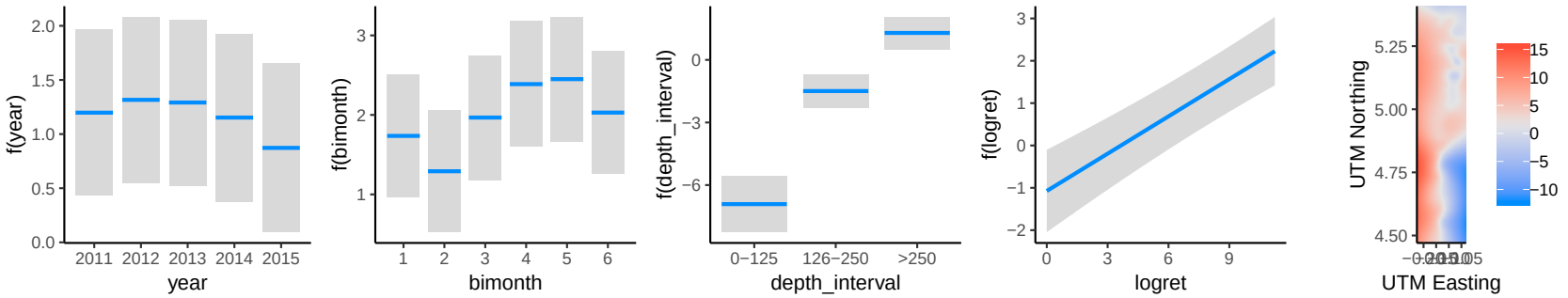
RF_{strat2}

```
Mean of squared residuals: 17874.49
          % Var explained: 49.87
```



GAM_{strat2}

R-sq.(adj) = -305 Deviance explained = 66.8% -REML = 63048 Scale est. = 131.17 n = 35440

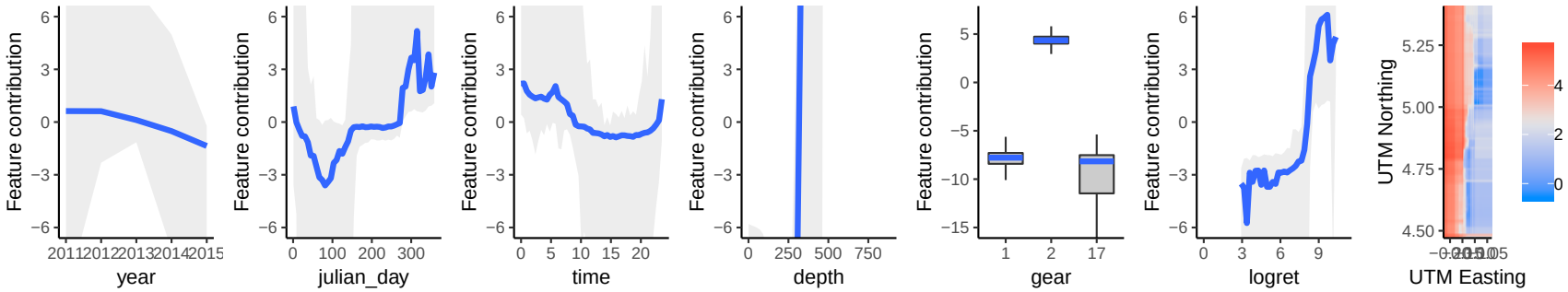


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	34.7	1.29000046063504e-318	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	0.0479	0.344	0.139	0.889	
bimonth2	-0.445	0.11	-4.03	5.55e-05	***
bimonth3	0.231	0.124	1.86	0.0627	.
bimonth4	0.651	0.134	4.85	1.25e-06	***
bimonth5	0.714	0.131	5.47	4.59e-08	***
bimonth6	0.293	0.126	2.33	0.02	*
year2012	0.118	0.104	1.14	0.256	
year2013	0.0936	0.105	0.892	0.372	
year2014	-0.0451	0.117	-0.384	0.701	
year2015	-0.325	0.129	-2.53	0.0114	*
depth_interval0-125	-8.2	0.581	-14.1	3.17e-45	***
depth_interval126-250	-2.79	0.135	-20.6	1.34e-93	***
logret	0.294	0.0388	7.57	3.95e-14	***

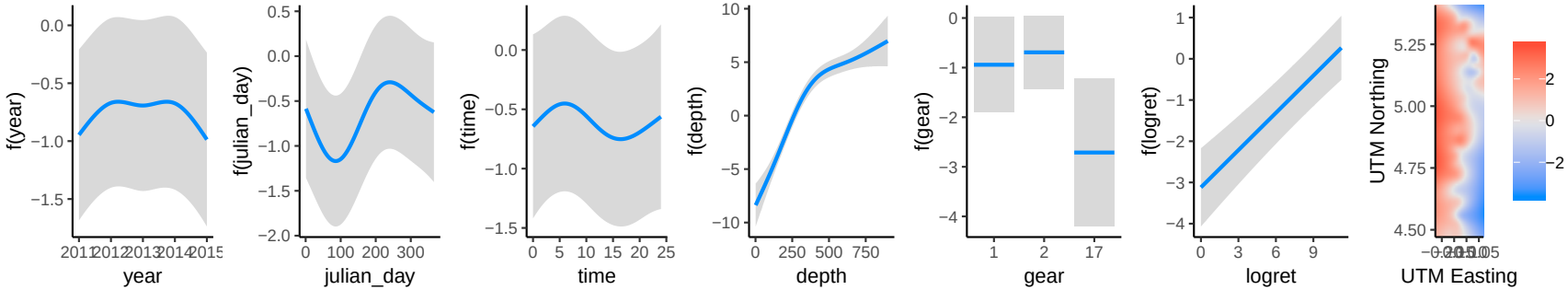
$RF_{nonlinear}$

```
Mean of squared residuals: 15794.11
          % Var explained: 55.71
```



GAM$_{\text{nonlinear}}$

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



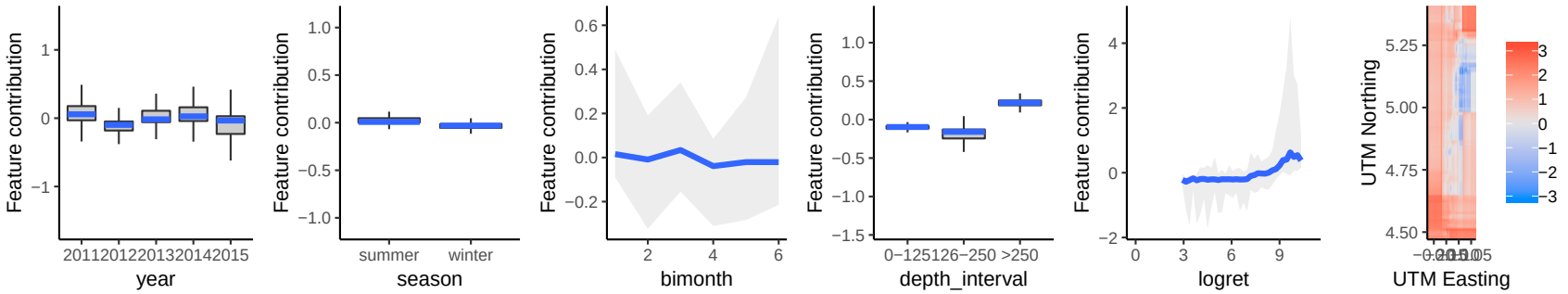
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	45.7	48.1	11.6	3.84e-87	***
s(year)	3.85	3.99	3.98	0.00323	**
s(depth)	3.97	4	371	2.52000000822904e-316	***
s(julian_day)	3.97	4	18.2	5.15e-15	***
s(time)	3.79	3.97	2.32	0.0554	.

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

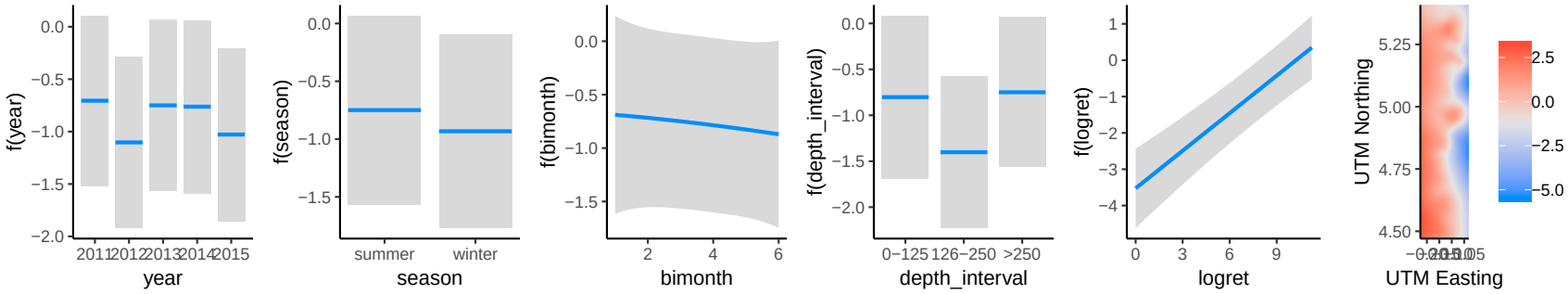
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 36.92363
          % Var explained: 2.19
```



$\mathbf{GAM_{strat1}}$

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

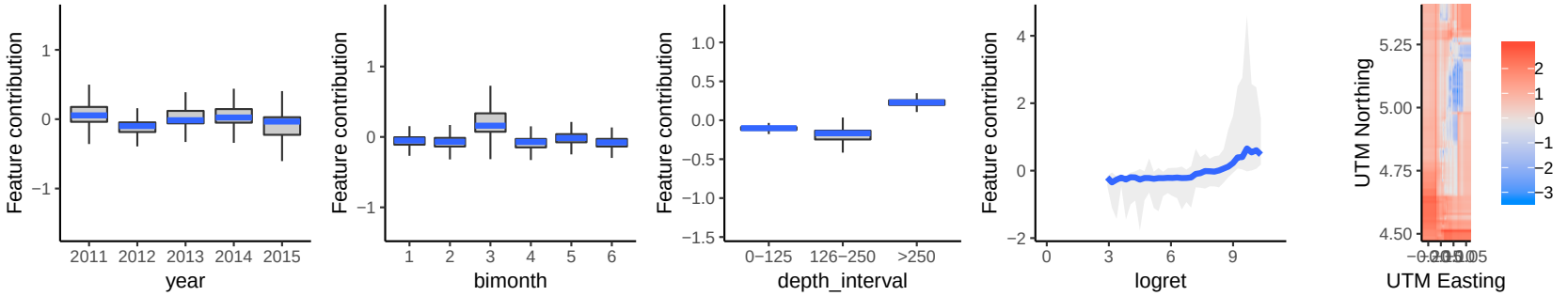


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	43.5	47.7	7.92	1.99e-52	***

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	***

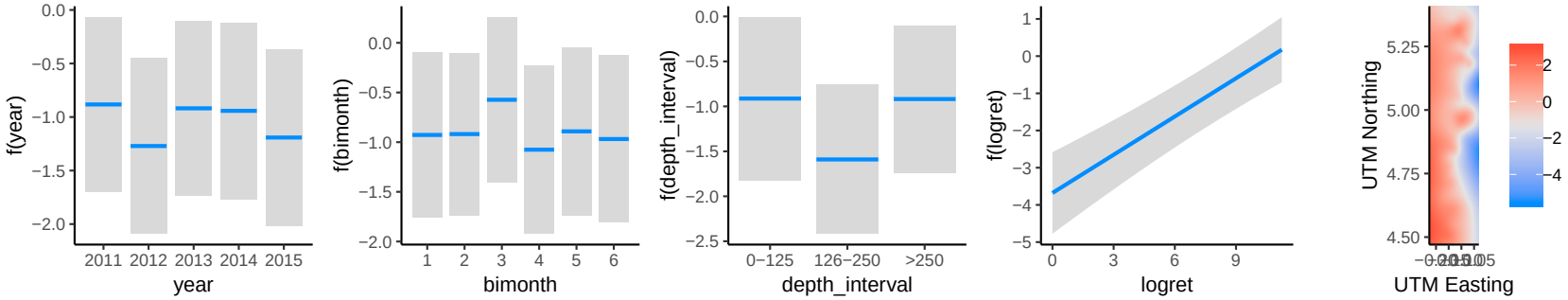
RF_{strat2}

```
Mean of squared residuals: 36.73508
          % Var explained: 2.69
```



GAM_{strat2}

R-sq.(adj) = 0.0441 Deviance explained = 20.5% -REML = 27687 Scale est. = 57.611 n = 35440

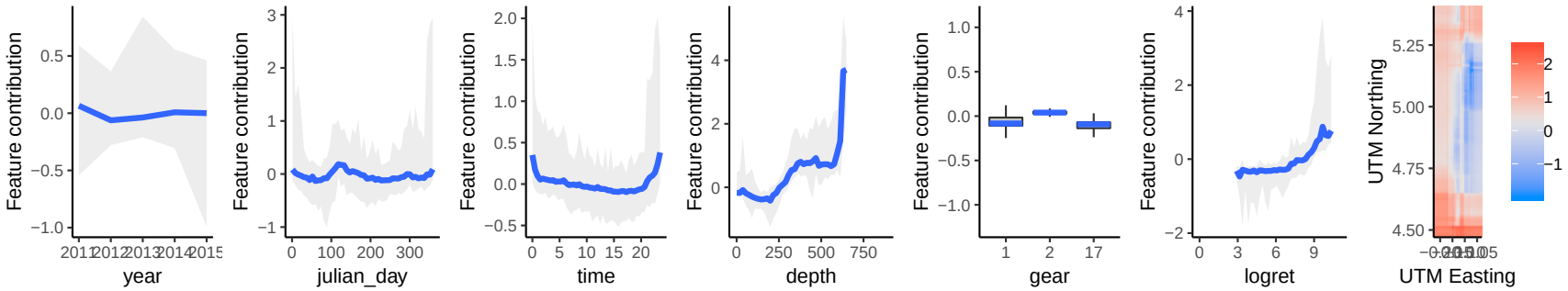


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	43.7	47.7	7.94	1.19e-52	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-3.02	0.408	-7.39	1.47e-13	***
bimonth2	0.0088	0.144	0.0612	0.951	
bimonth3	0.354	0.151	2.34	0.0192	*
bimonth4	-0.148	0.173	-0.858	0.391	
bimonth5	0.0362	0.17	0.213	0.831	
bimonth6	-0.041	0.175	-0.234	0.815	
year2012	-0.388	0.138	-2.82	0.00481	**
year2013	-0.0362	0.128	-0.284	0.777	
year2014	-0.0597	0.137	-0.436	0.663	
year2015	-0.309	0.144	-2.14	0.0324	*
depth_interval0-125	0.00534	0.221	0.0241	0.981	
depth_interval126-250	-0.671	0.126	-5.34	9.15e-08	***
logret	0.343	0.0477	7.18	7.1e-13	***

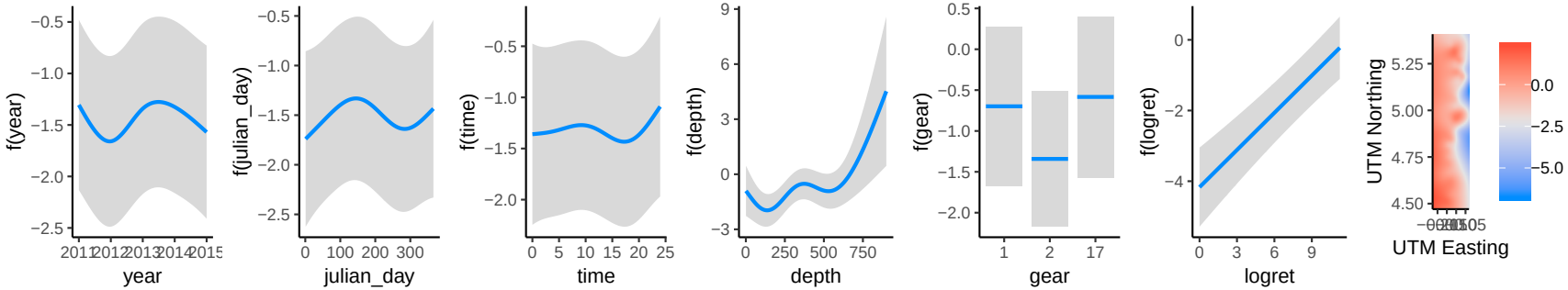
$RF_{nonlinear}$

```
Mean of squared residuals: 36.05839
          % Var explained: 4.48
```



GAM$_{\text{nonlinear}}$

R-sq.(adj) = 0.0407 Deviance explained = 21% -REML = 27672 Scale est. = 55.784 n = 35440



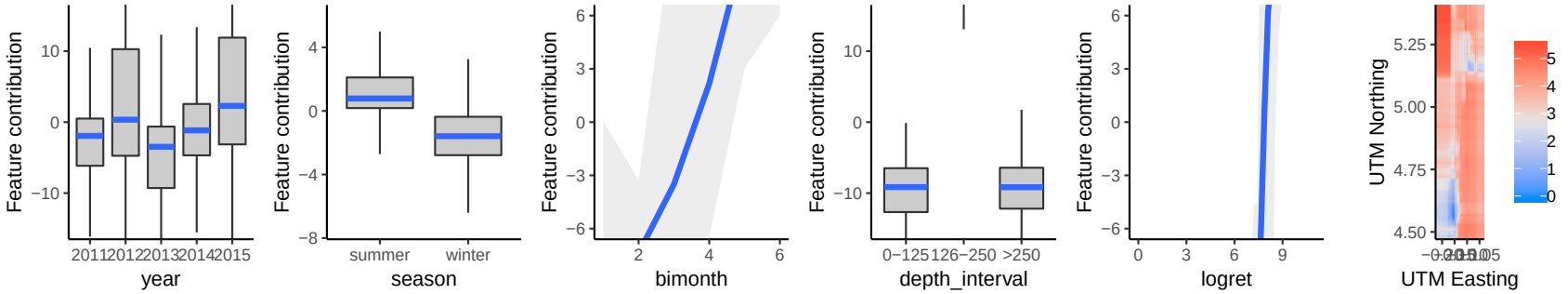
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	44.4	48	7.53	2.21e-49	***
s(year)	3.84	3.98	2.7	0.022	*
s(depth)	3.96	4	12.4	7.68e-10	***
s(julian_day)	3.6	3.91	1.61	0.134	
s(time)	3.52	3.87	0.981	0.513	

term	estimate	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

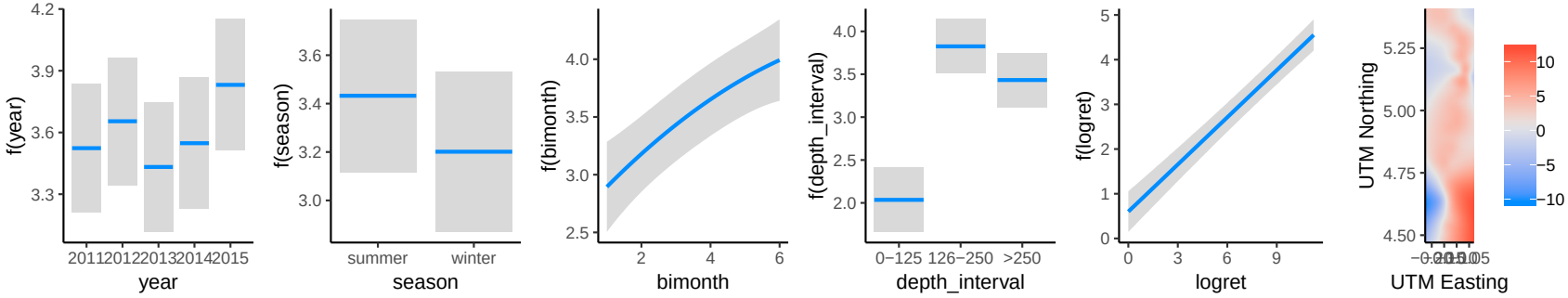
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 18788.2
          % Var explained: 20.61
```



$\mathbf{GAM_{strat1}}$

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

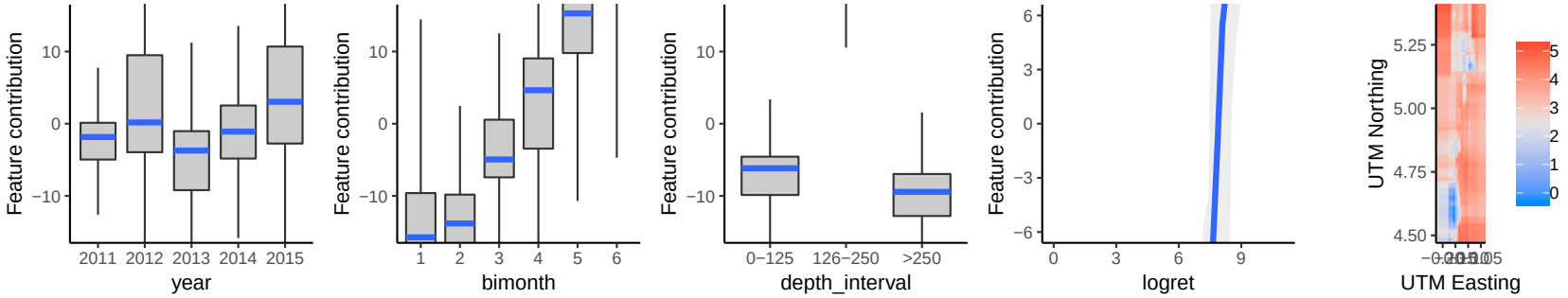


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.7	49	21.1	2.24e-182	***

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_interval126-250	0.393	0.0567	6.93	4.36e-12	***
logret	0.352	0.0216	16.3	2.56e-59	***

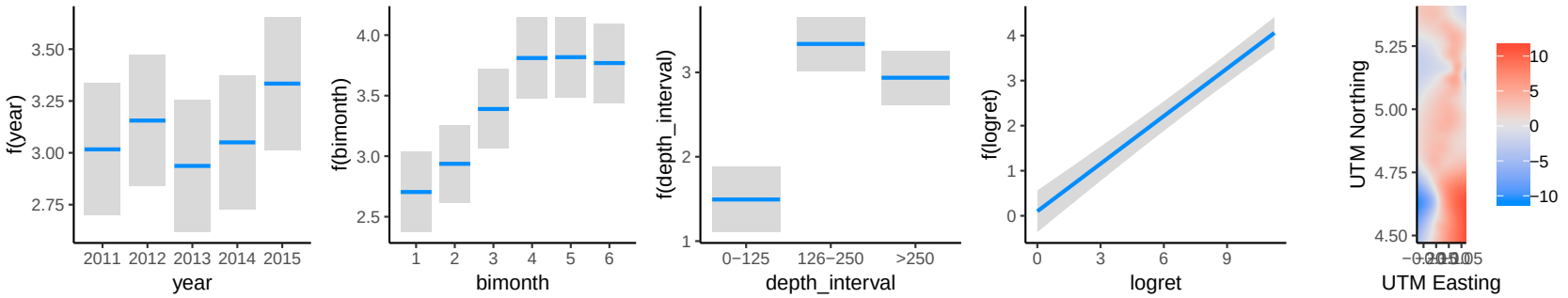
RF_{strat2}

```
Mean of squared residuals: 18775.85
          % Var explained: 20.67
```



GAM_{strat2}

R-sq.(adj) = 0.0775 Deviance explained = 22.9% -REML = 1.2471e+05 Scale est. = 51.454 n = 35440

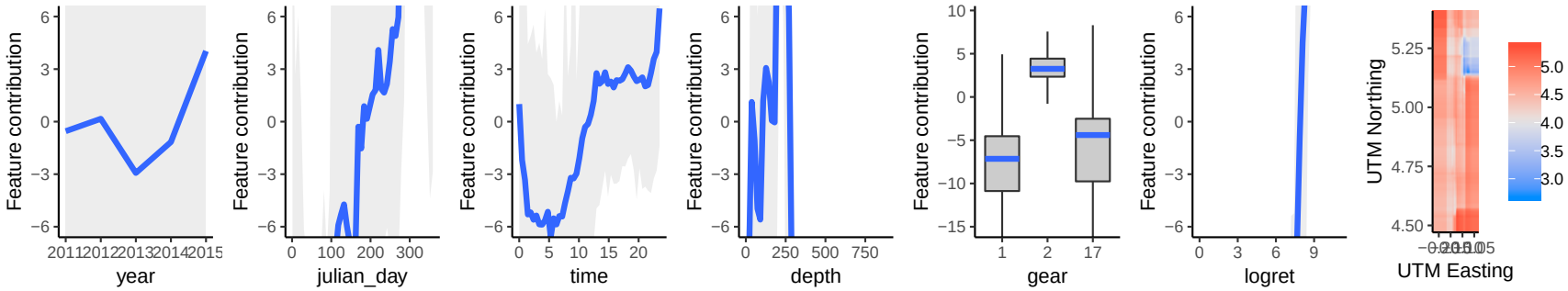


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.7	49	20.2	9.88e-174	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	0.104	0.192	0.543	0.587	
bimonth2	0.233	0.0769	3.03	0.00247	**
bimonth3	0.686	0.0805	8.51	1.75e-17	***
bimonth4	1.11	0.0865	12.8	2.27e-37	***
bimonth5	1.11	0.0851	13.1	4.72e-39	***
bimonth6	1.07	0.0834	12.8	2.87e-37	***
year2012	0.139	0.0657	2.12	0.0343	*
year2013	-0.0798	0.0652	-1.22	0.221	
year2014	0.0338	0.0698	0.484	0.628	
year2015	0.317	0.0694	4.57	4.93e-06	***
depth_interval0-125	-1.44	0.119	-12.1	1.33e-33	***
depth_interval126-250	0.401	0.0579	6.92	4.46e-12	***
logret	0.352	0.0221	16	3.44e-57	***

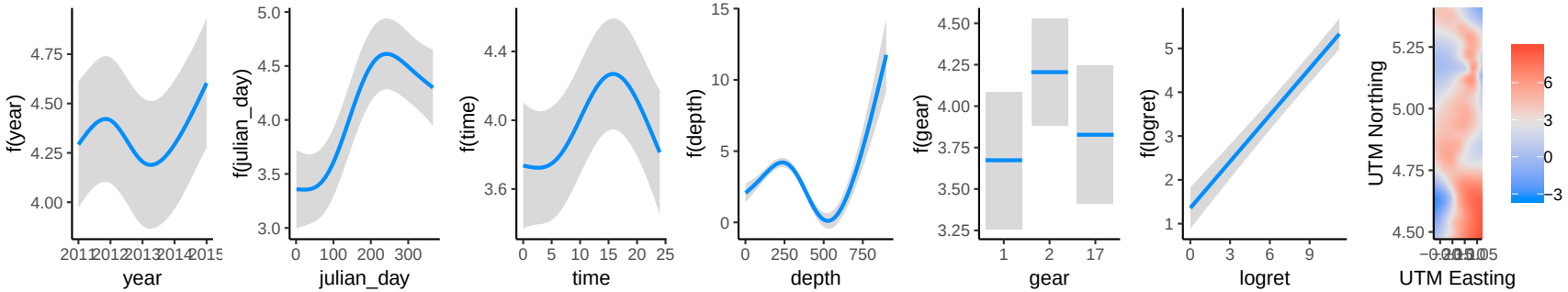
$RF_{nonlinear}$

```
Mean of squared residuals: 17934.13
          % Var explained: 24.22
```



GAM$_{\text{nonlinear}}$

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



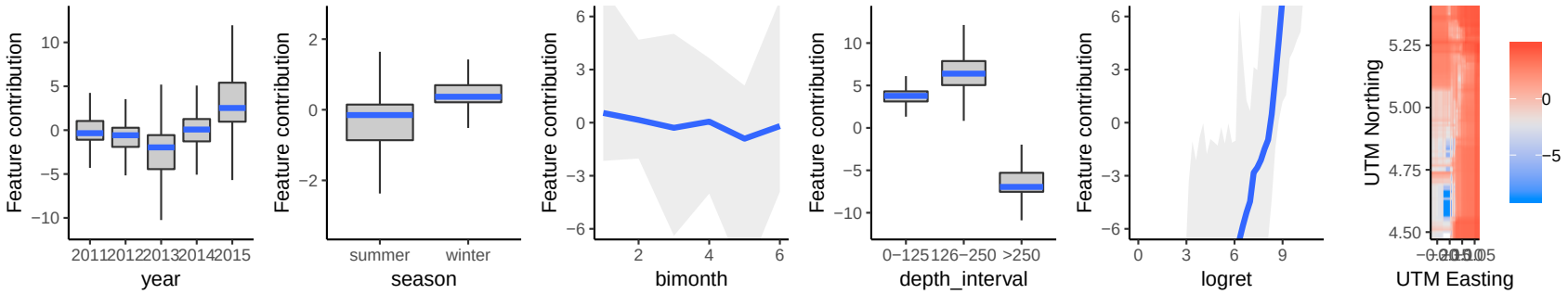
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.5	49	15.9	2.96e-130	***
s(year)	3.89	3.99	10.5	1.19e-08	***
s(depth)	3.99	4	138	3.94e-117	***
s(julian_day)	3.95	4	92.4	2.8e-78	***
s(time)	3.88	3.99	20.4	8.62e-17	***

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

Species 9: Pacific Halibut

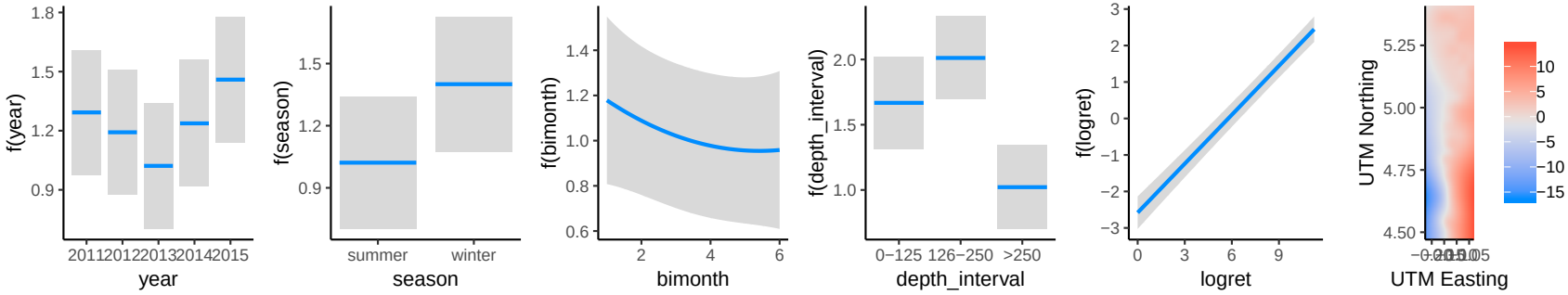
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 3041.369
          % Var explained: 18.98
```



$\mathbf{GAM_{strat1}}$

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

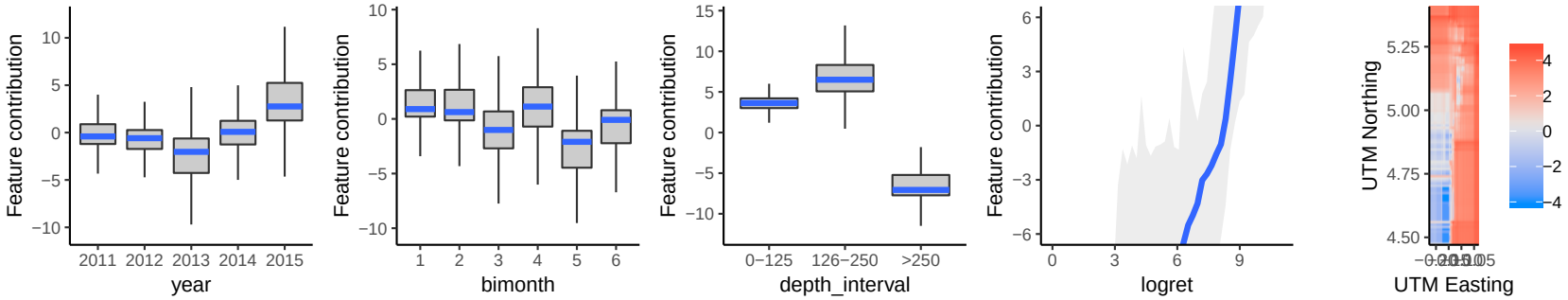


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.7	48.9	43.8	0	***

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	***
year2014	-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	***

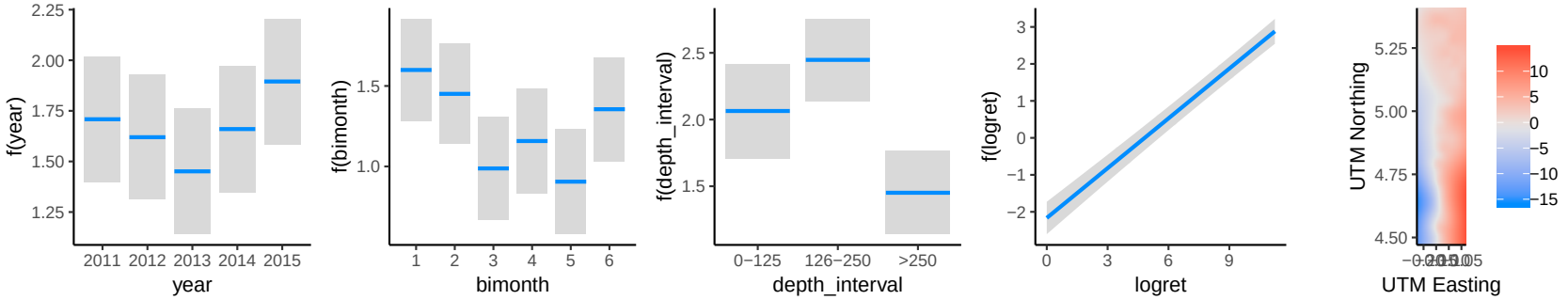
RF_{strat2}

```
Mean of squared residuals: 3032.182
          % Var explained: 19.22
```



GAM_{strat2}

R-sq.(adj) = 0.104 Deviance explained = 27.5% -REML = 74245 Scale est. = 43.765 n = 35440

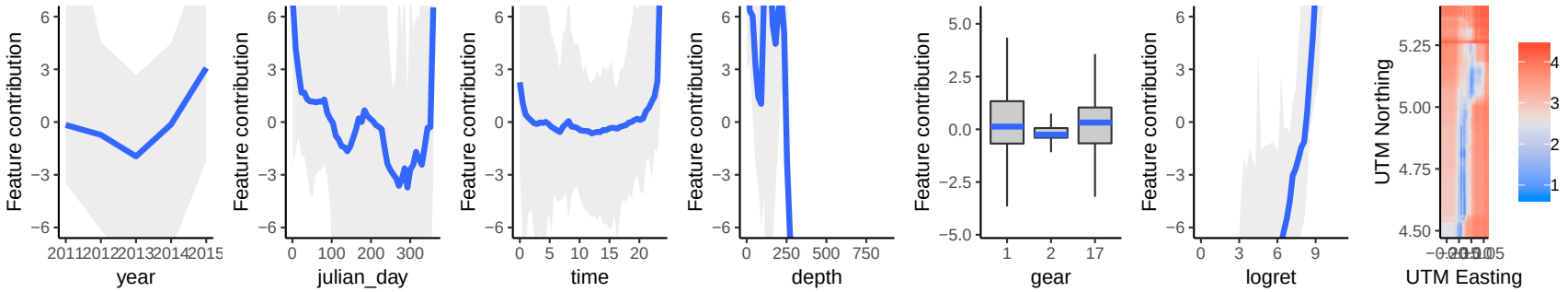


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.7	48.9	44.6	0	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.36	0.174	-7.81	5.91e-15	***
bimonth2	-0.148	0.0593	-2.5	0.0123	*
bimonth3	-0.612	0.0649	-9.42	4.58e-21	***
bimonth4	-0.441	0.0719	-6.14	8.36e-10	***
bimonth5	-0.694	0.0745	-9.32	1.23e-20	***
bimonth6	-0.244	0.0747	-3.26	0.0011	**
year2012	-0.0887	0.0559	-1.59	0.112	
year2013	-0.257	0.0558	-4.62	3.9e-06	***
year2014	-0.0485	0.0588	-0.825	0.41	
year2015	0.187	0.057	3.27	0.00106	**
depth_interval0-125	0.613	0.0985	6.23	4.82e-10	***
depth_interval126-250	0.996	0.0518	19.2	4.15e-82	***
logret	0.449	0.02	22.4	2.03e-110	***

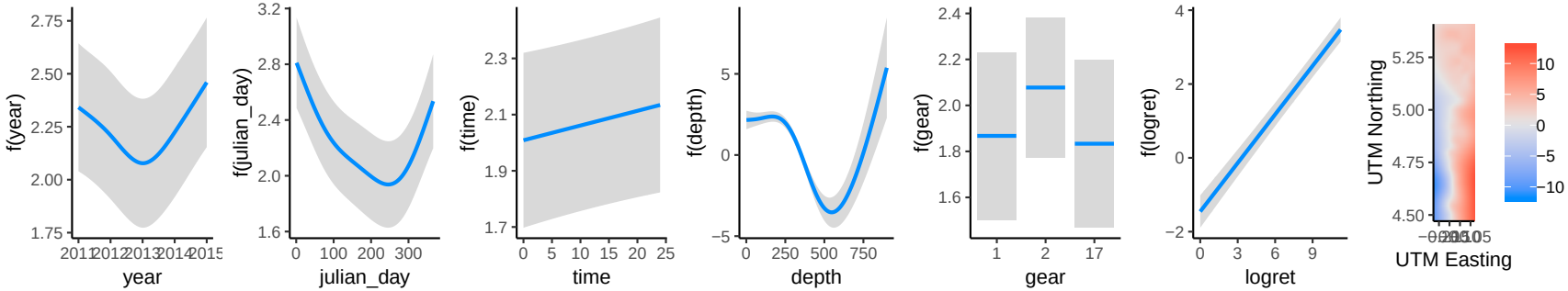
$RF_{nonlinear}$

```
Mean of squared residuals: 2970.634
          % Var explained: 20.86
```



GAM$_{\text{nonlinear}}$

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



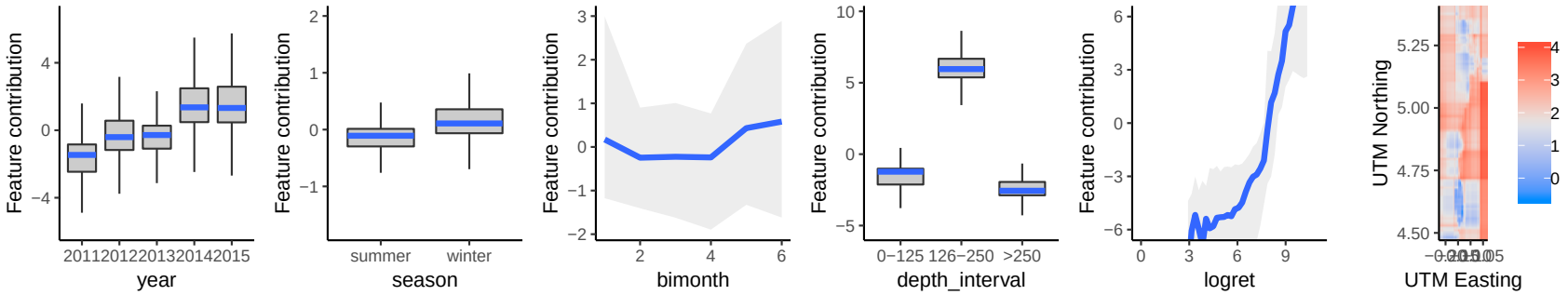
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.2	48.8	41.1	0	***
s(year)	3.59	3.9	15.1	5.61e-12	***
s(depth)	3.98	4	140	2.4e-119	***
s(julian_day)	3.83	3.98	26.3	6.75e-22	***
s(time)	1.01	1.02	3.54	0.0592	.

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

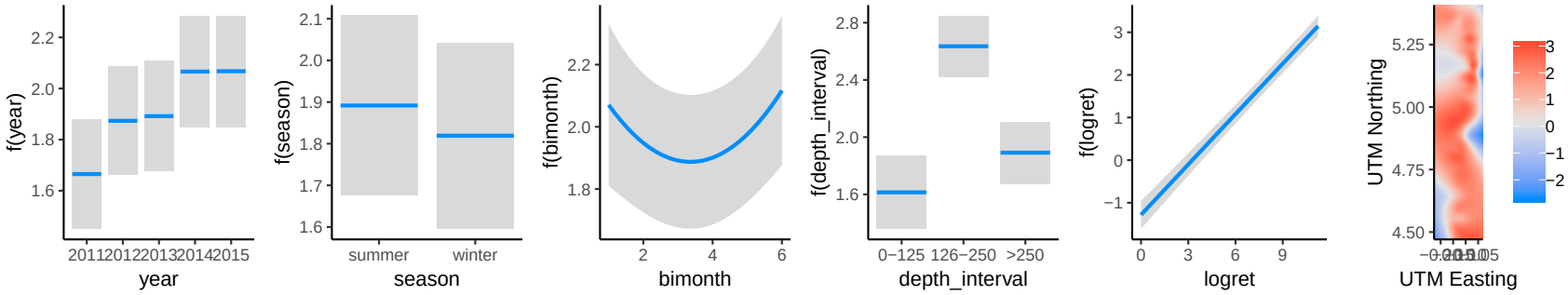
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 385.3107
          % Var explained: 22.37
```



$\mathbf{GAM_{strat1}}$

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

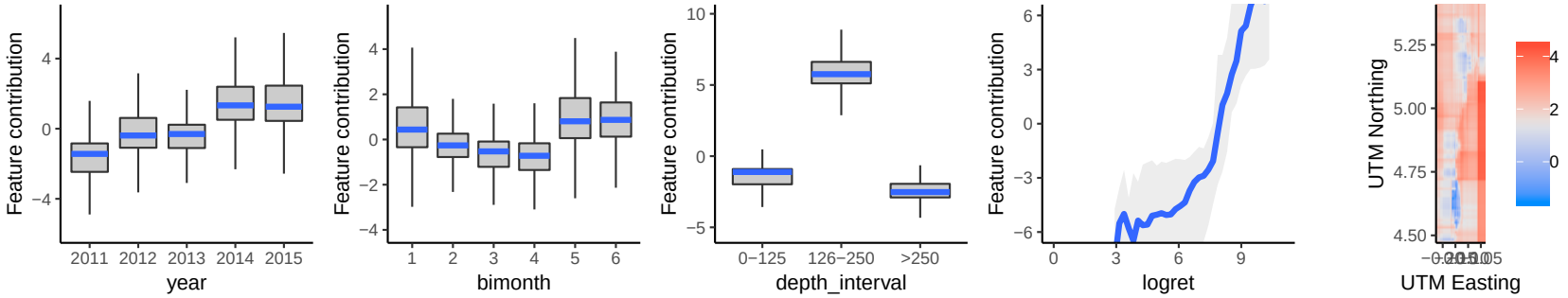


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.3	49	28.5	1.56e-256	***

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^2)	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	***

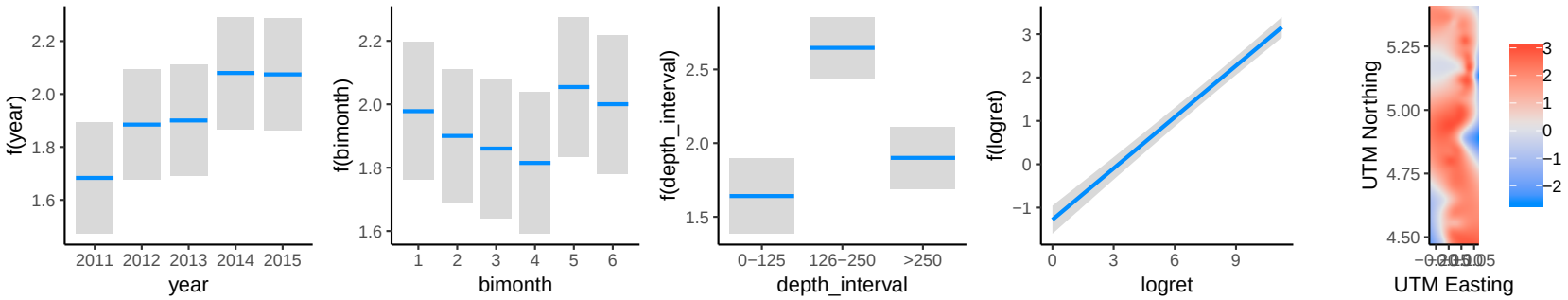
RF_{strat2}

```
Mean of squared residuals: 384.3252
          % Var explained: 22.57
```



GAM_{strat2}

R-sq.(adj) = 0.126 Deviance explained = 20.1% -REML = 92517 Scale est. = 20.012 n = 35440

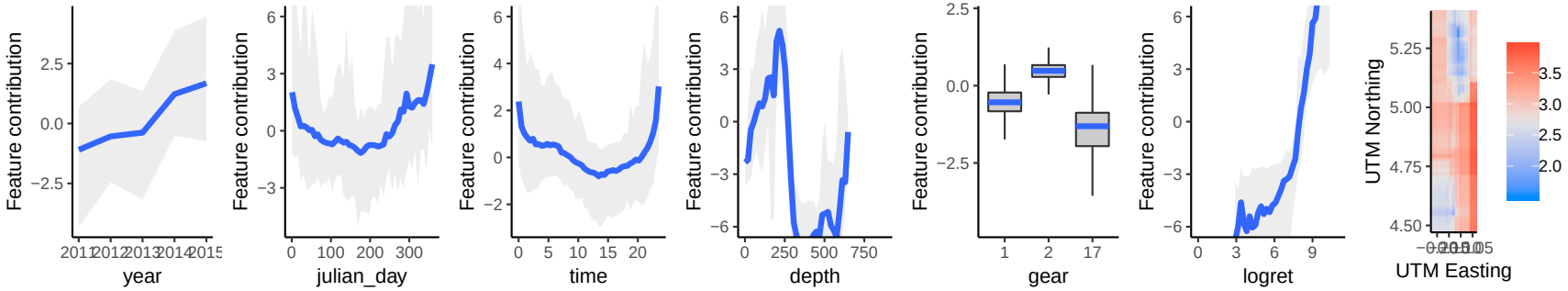


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.3	49	29	1.22e-261	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-1.44	0.139	-10.4	3.35e-25	***
bimonth2	-0.0782	0.0467	-1.67	0.0944	.
bimonth3	-0.118	0.0513	-2.3	0.0213	*
bimonth4	-0.163	0.0573	-2.85	0.00432	**
bimonth5	0.0761	0.055	1.38	0.167	
bimonth6	0.0221	0.0535	0.414	0.679	
year2012	0.201	0.0457	4.4	1.07e-05	***
year2013	0.217	0.0446	4.87	1.12e-06	***
year2014	0.396	0.0466	8.5	1.98e-17	***
year2015	0.391	0.0472	8.29	1.17e-16	***
depth_interval0-125	-0.259	0.0798	-3.25	0.00116	**
depth_interval126-250	0.746	0.0384	19.4	1.11e-83	***
logret	0.395	0.0161	24.6	4.46e-132	***

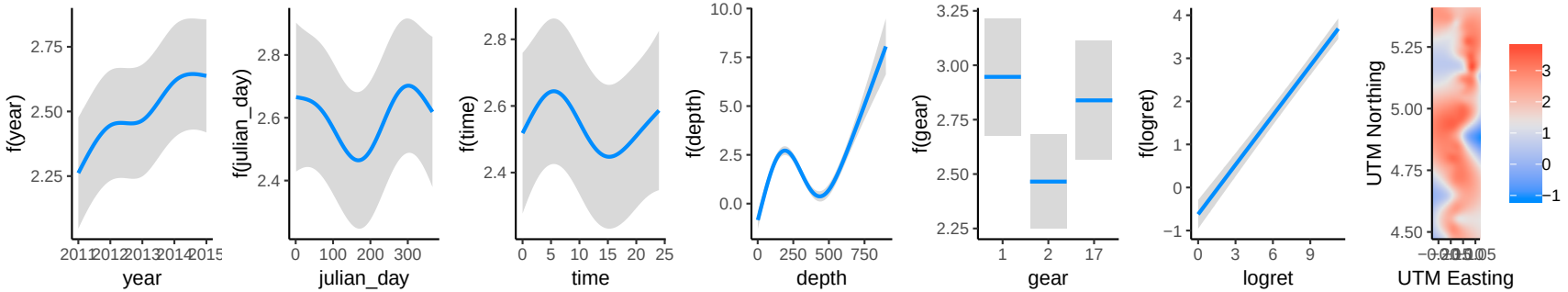
$RF_{nonlinear}$

```
Mean of squared residuals: 371.2104
          % Var explained: 25.21
```



GAM$_{\text{nonlinear}}$

R-sq.(adj) = 0.0892 Deviance explained = 23.4% -REML = 91802 Scale est. = 19.723 n = 35440



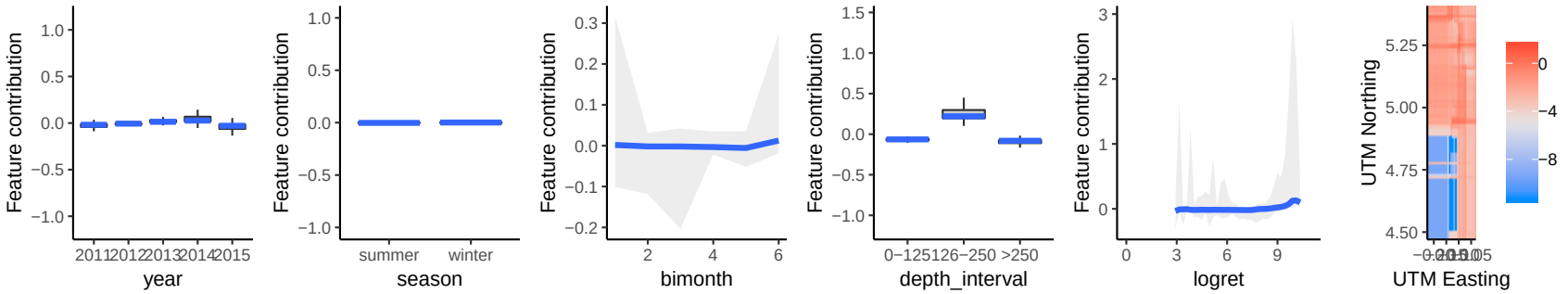
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	26.7	4.88e-238	***
s(year)	3.76	3.97	20.7	6.69e-17	***
s(depth)	4	4	277	8.25e-236	***
s(julian_day)	3.83	3.98	6.43	2.58e-05	***
s(time)	3.73	3.96	5.01	0.000366	***

term	estimate	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

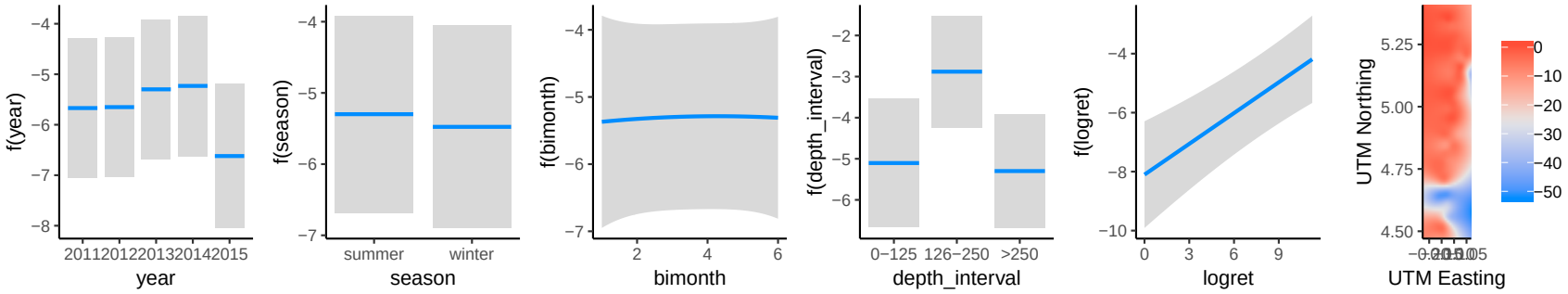
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 7.098818
          % Var explained: 17.73
```



$\mathbf{GAM_{strat1}}$

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

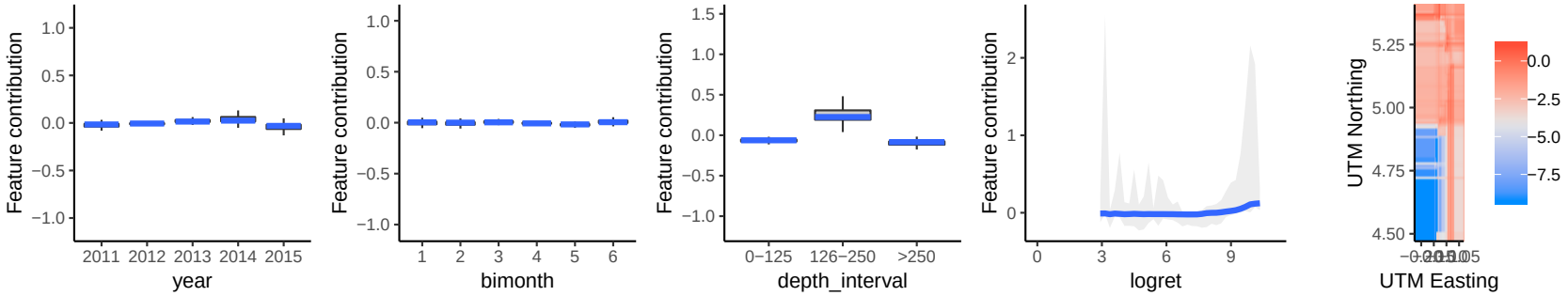


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	45.9	47.2	9.94	4.45e-69	***

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_interval126-250	2.42	0.211	11.4	2.85e-30	***
logret	0.348	0.0767	4.54	5.74e-06	***

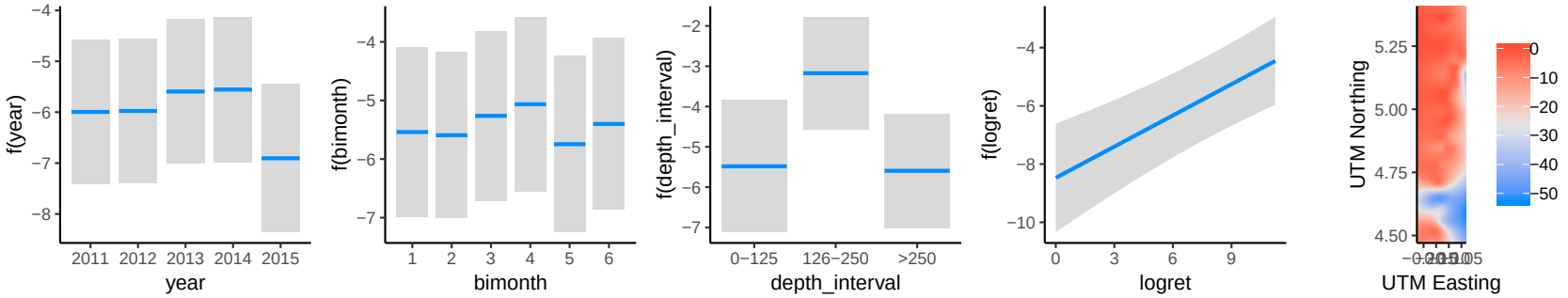
RF_{strat2}

```
      Mean of squared residuals: 7.062963
                % Var explained: 18.14
```



GAM_{strat2}

R-sq.(adj) = -0.00526 Deviance explained = 52.9% -REML = 8395.8 Scale est. = 66.163 n = 35440

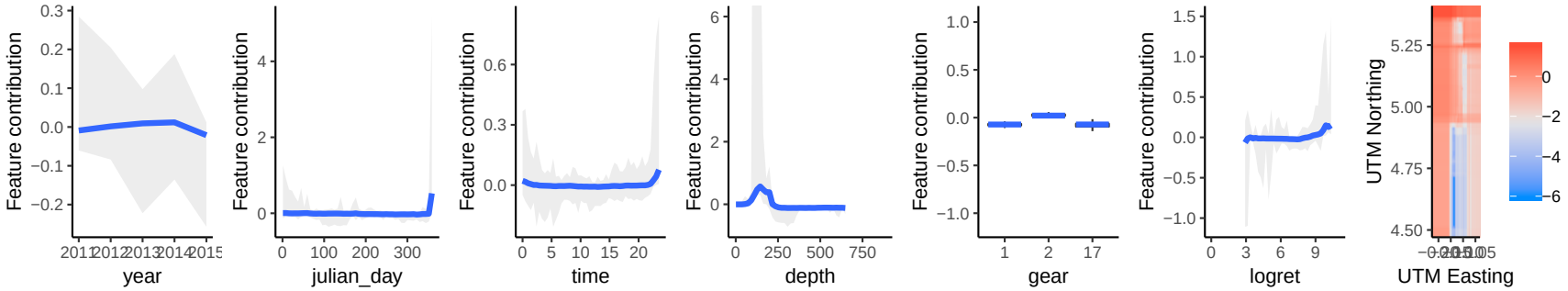


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	45.8	47.2	9.45	1.09e-64	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-11.2	1.32	-8.5	1.94e-17	***
bimonth2	-0.0543	0.249	-0.218	0.827	
bimonth3	0.276	0.271	1.02	0.307	
bimonth4	0.474	0.347	1.37	0.172	
bimonth5	-0.207	0.368	-0.563	0.573	
bimonth6	0.138	0.317	0.436	0.662	
year2012	0.0189	0.259	0.0728	0.942	
year2013	0.401	0.24	1.67	0.0951	.
year2014	0.439	0.263	1.67	0.0946	.
year2015	-0.911	0.307	-2.96	0.00305	**
depth_interval0-125	0.11	0.456	0.241	0.81	
depth_interval126-250	2.42	0.216	11.2	5.58e-29	***
logret	0.357	0.0789	4.53	5.98e-06	***

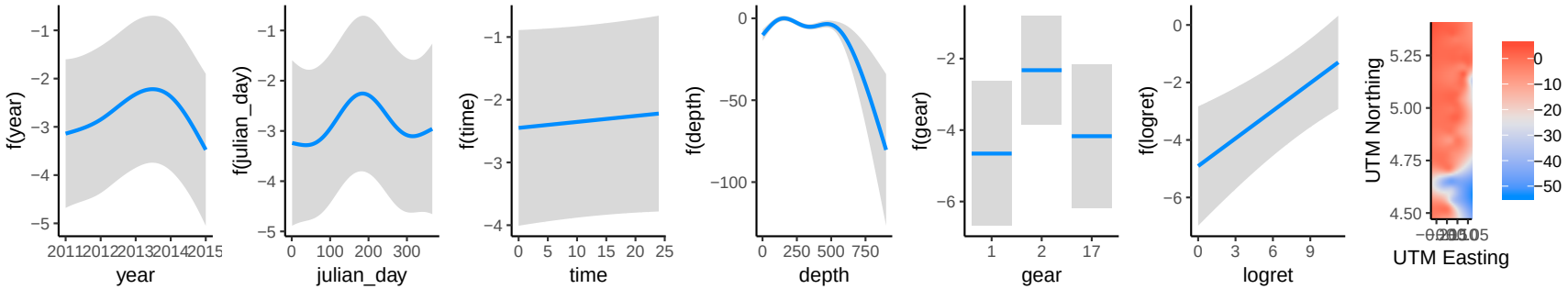
$RF_{nonlinear}$

```
Mean of squared residuals: 6.679766
          % Var explained: 22.59
```



GAM$_{\text{nonlinear}}$

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



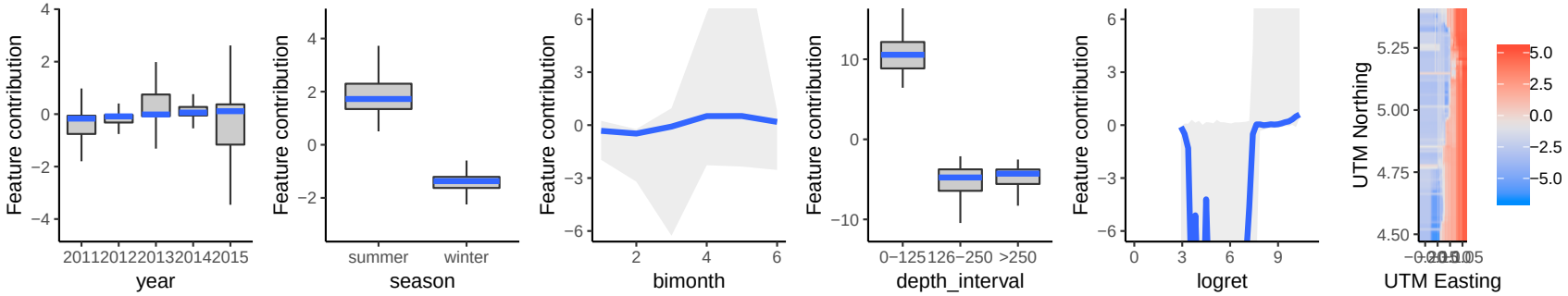
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	45.6	47.1	6.42	2.83e-38	***
s(year)	3.68	3.94	5.16	0.000429	***
s(depth)	3.97	4	44.7	1.72e-37	***
s(julian_day)	3.77	3.97	2.44	0.0378	*
s(time)	1.01	1.01	0.461	0.499	

term	estimate	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

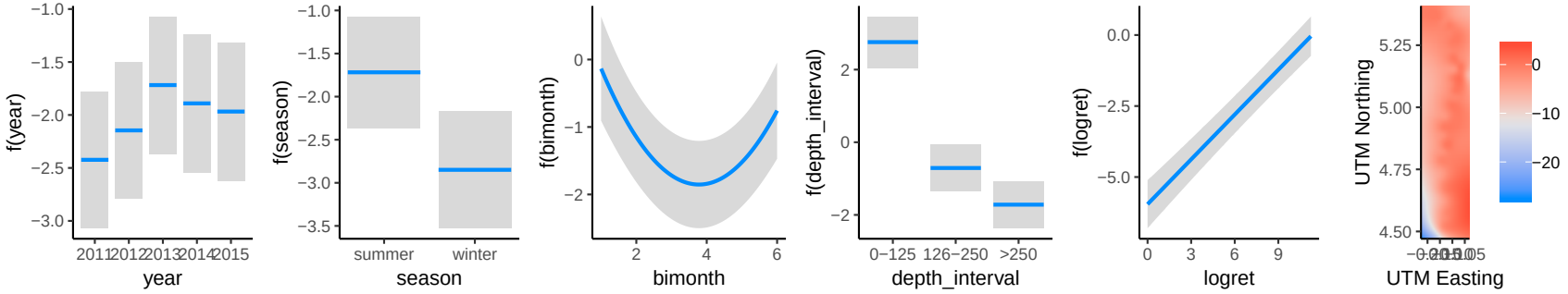
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 2193.15
          % Var explained: 28.43
```



$\mathbf{GAM_{strat1}}$

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

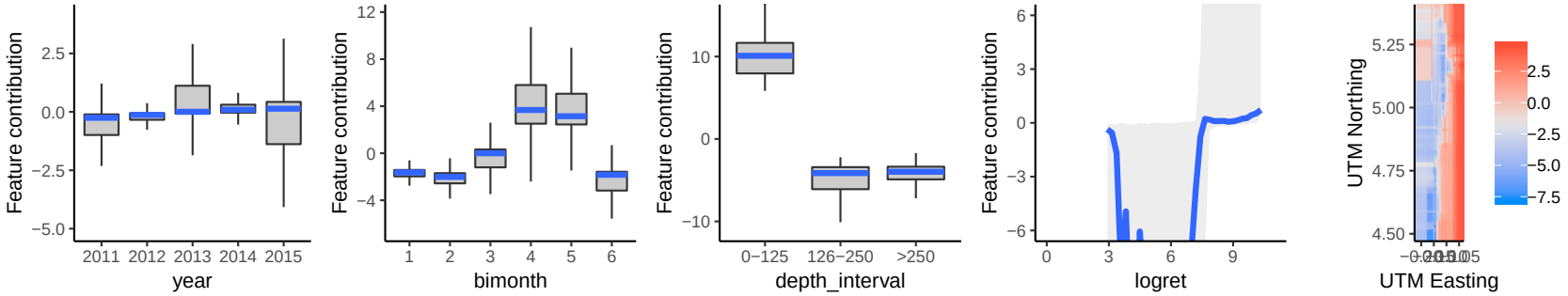


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	13.1	9.18e-104	***

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	***

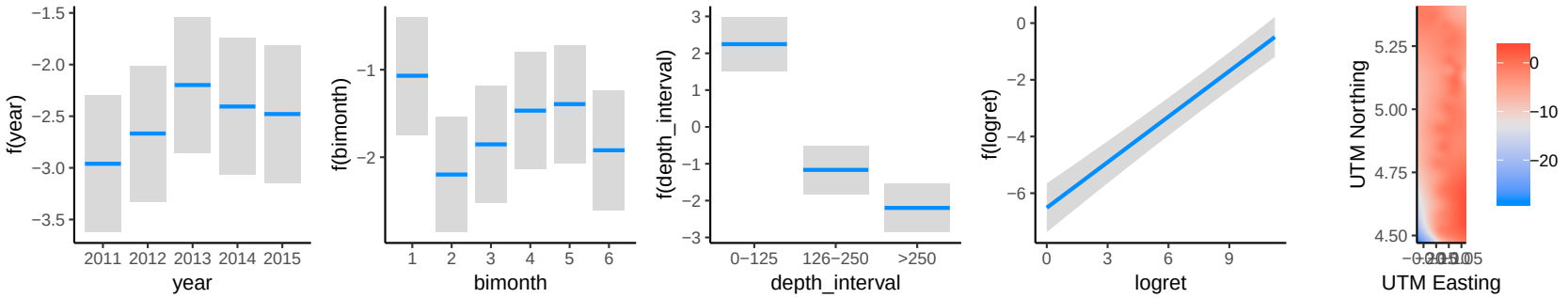
RF_{strat2}

```
Mean of squared residuals: 2197.043
          % Var explained: 28.31
```



GAM_{strat2}

R-sq.(adj) = 0.133 Deviance explained = 47.5% -REML = 51460 Scale est. = 49.624 n = 35440

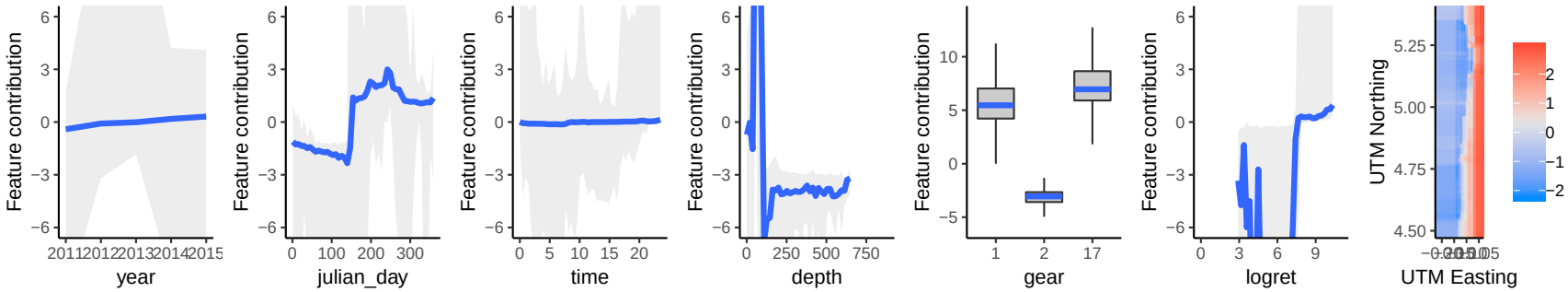


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	12	5.56e-93	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-6.13	0.327	-18.8	4.4e-78	***
bimonth2	-1.13	0.146	-7.73	1.09e-14	***
bimonth3	-0.784	0.143	-5.47	4.57e-08	***
bimonth4	-0.399	0.148	-2.69	0.00708	**
bimonth5	-0.324	0.15	-2.16	0.031	*
bimonth6	-0.852	0.168	-5.07	4.03e-07	***
year2012	0.293	0.117	2.5	0.0123	*
year2013	0.762	0.112	6.78	1.26e-11	***
year2014	0.554	0.12	4.64	3.56e-06	***
year2015	0.481	0.121	3.99	6.51e-05	***
depth_interval0-125	4.45	0.204	21.8	2.47e-104	***
depth_interval126-250	1.03	0.125	8.26	1.45e-16	***
logret	0.536	0.0369	14.5	1.36e-47	***

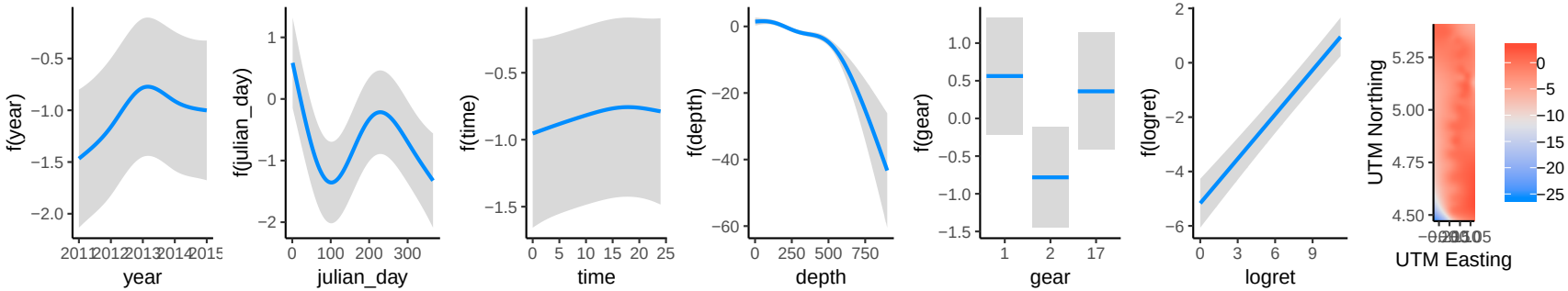
$RF_{nonlinear}$

```
Mean of squared residuals: 2006.376
          % Var explained: 34.53
```



GAM$_{\text{nonlinear}}$

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



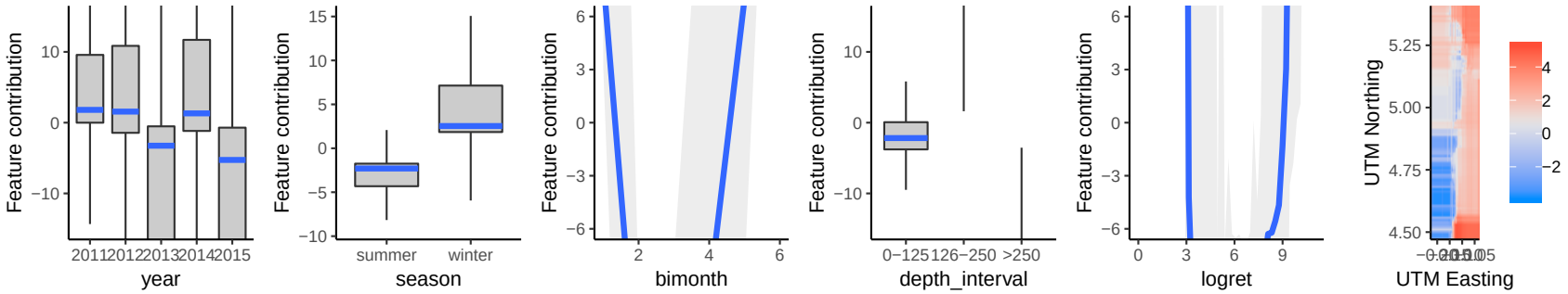
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.9	48.9	10.2	1.47e-75	***
s(year)	3.73	3.96	10.1	4.15e-08	***
s(depth)	3.95	4	59.9	2.91e-50	***
s(julian_day)	3.98	4	29.4	1.56e-24	***
s(time)	2.05	2.53	0.809	0.434	

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

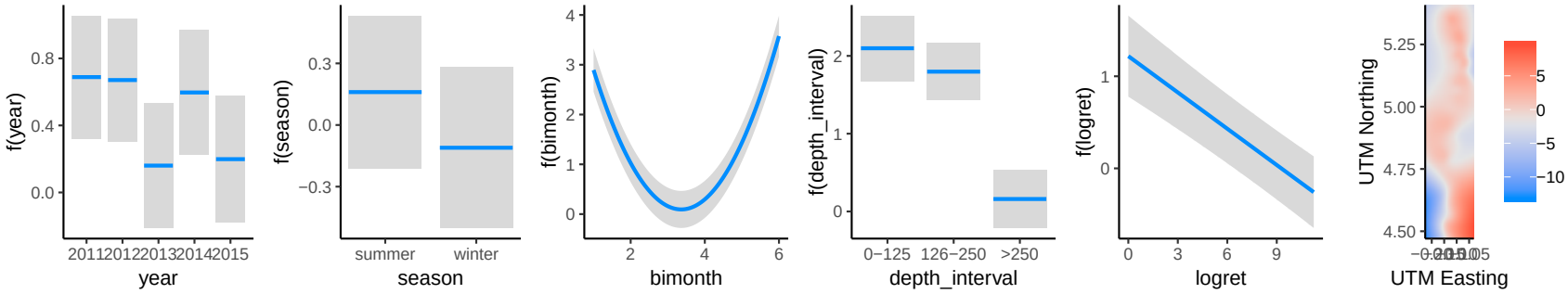
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 444124.3
          % Var explained: 25.06
```



$\mathbf{GAM_{strat1}}$

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

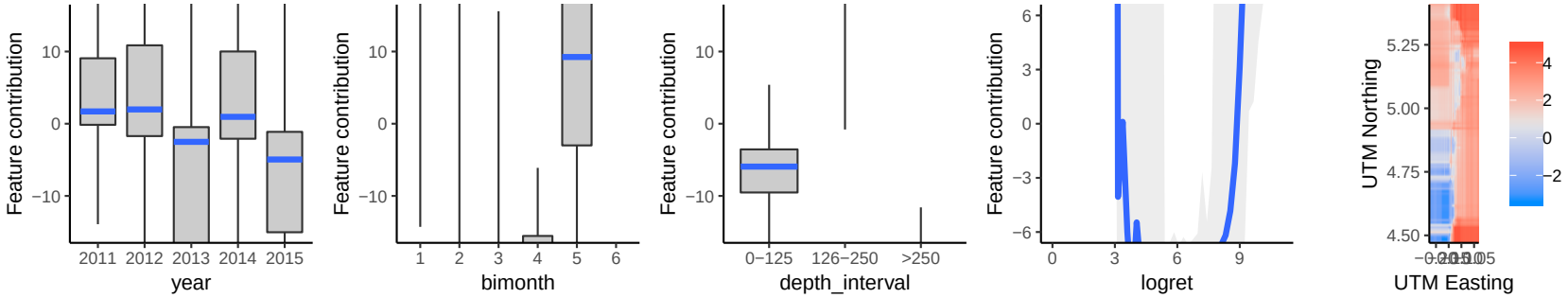


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.2	49	40.8	0	***

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_interval126-250	1.64	0.0654	25	6.35e-137	***
logret	-0.131	0.0158	-8.31	9.88e-17	***

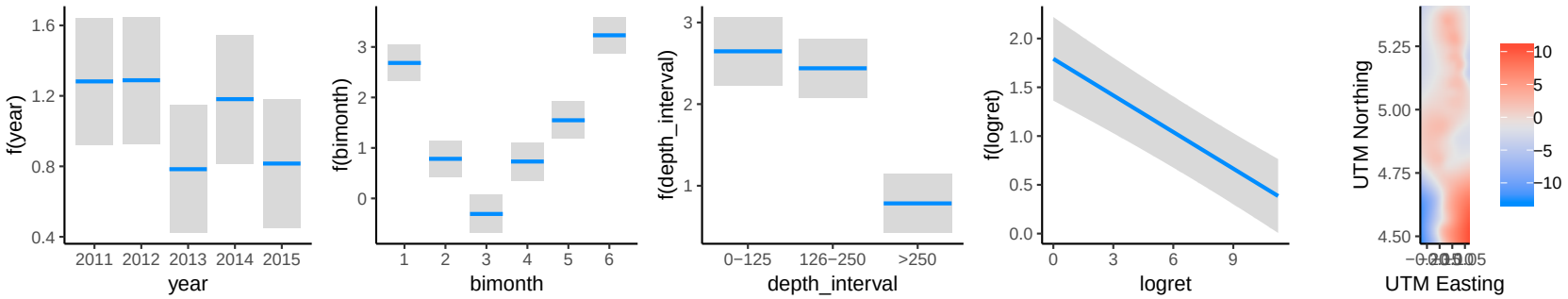
RF_{strat2}

```
Mean of squared residuals: 437552.2
          % Var explained: 26.17
```



GAM_{strat2}

R-sq.(adj) = 0.0405 Deviance explained = 41.4% -REML = 1.0462e+05 Scale est. = 61.069 n = 35440

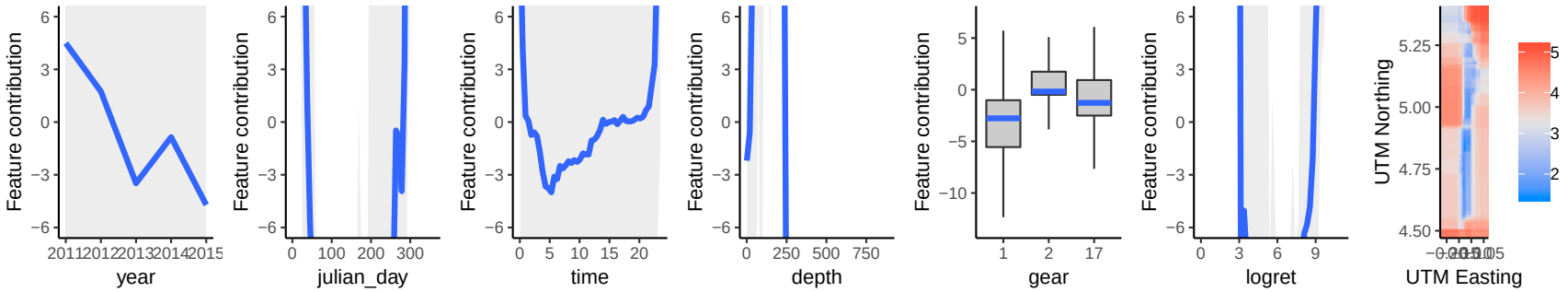


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.2	49	41.6	0	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	4.51	0.142	31.7	1.98e-217	***
bimonth2	-1.9	0.0758	-25.1	2.64e-137	***
bimonth3	-2.99	0.0857	-34.9	1.1e-262	***
bimonth4	-1.95	0.087	-22.4	8.08e-111	***
bimonth5	-1.14	0.0834	-13.6	4.79e-42	***
bimonth6	0.548	0.0741	7.39	1.48e-13	***
year2012	0.0066	0.0677	0.0976	0.922	
year2013	-0.498	0.0674	-7.39	1.52e-13	***
year2014	-0.101	0.0709	-1.42	0.156	
year2015	-0.466	0.0736	-6.33	2.41e-10	***
depth_interval0-125	1.86	0.122	15.3	7.32e-53	***
depth_interval126-250	1.66	0.0638	25.9	5.89e-147	***
logret	-0.125	0.0155	-8.05	8.23e-16	***

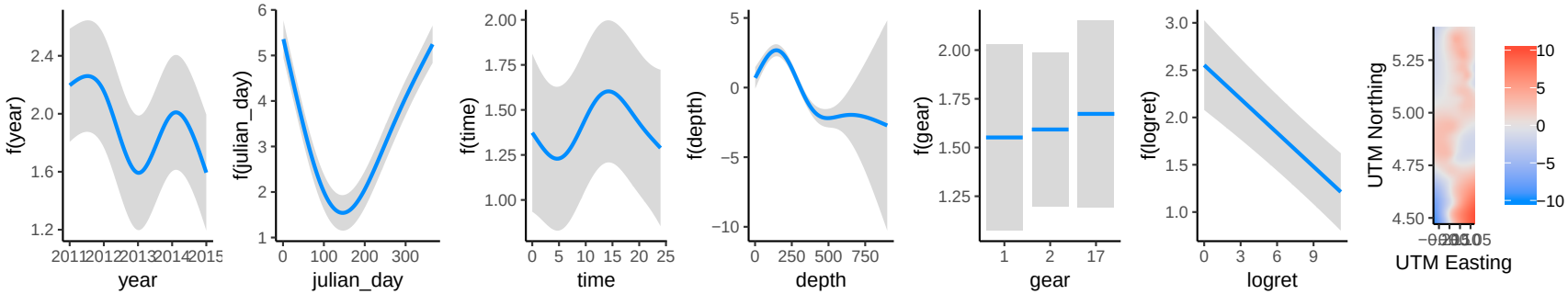
$RF_{nonlinear}$

```
Mean of squared residuals: 468651
          % Var explained: 20.92
```



GAM$_{\text{nonlinear}}$

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



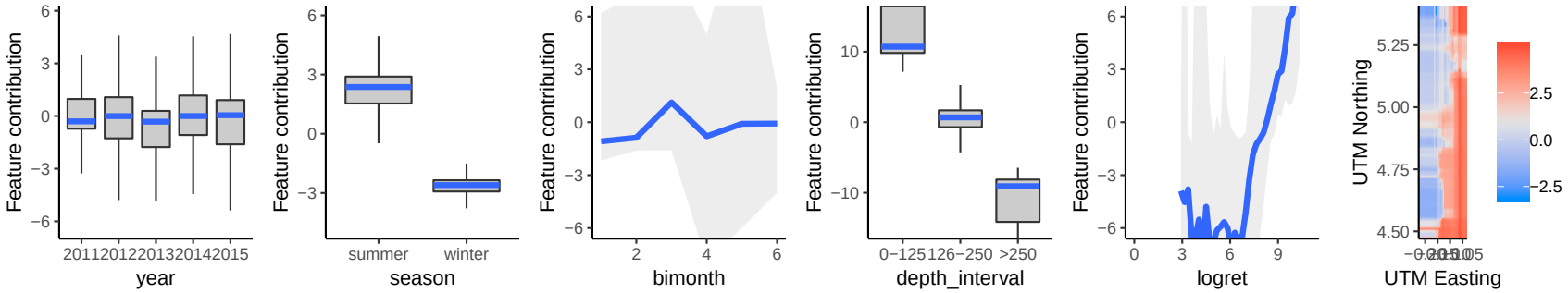
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	47.8	48.9	30.2	4.04e-273	***
s(year)	3.98	4	30.3	3.05e-25	***
s(depth)	3.96	4	218	2.39e-185	***
s(julian_day)	3.99	4	514	0	***
s(time)	3.84	3.98	6.37	3.26e-05	***

term	estimate	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

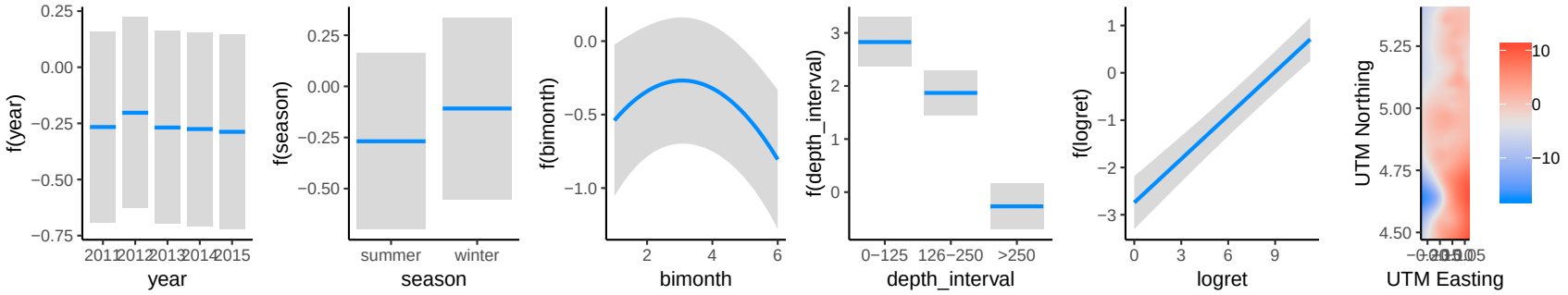
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 6941.341
          % Var explained: 7.86
```



$\mathbf{GAM_{strat1}}$

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

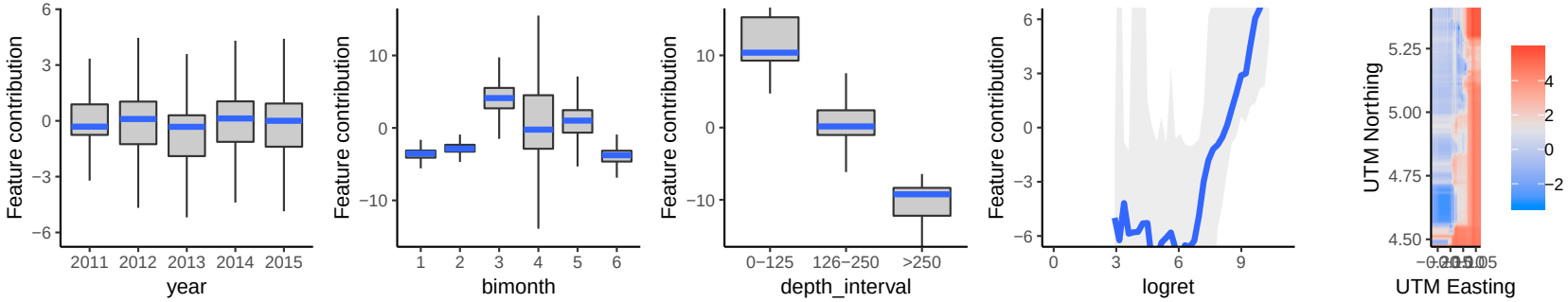


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.4	49	23.6	3.53e-205	***

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^2)	-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	***

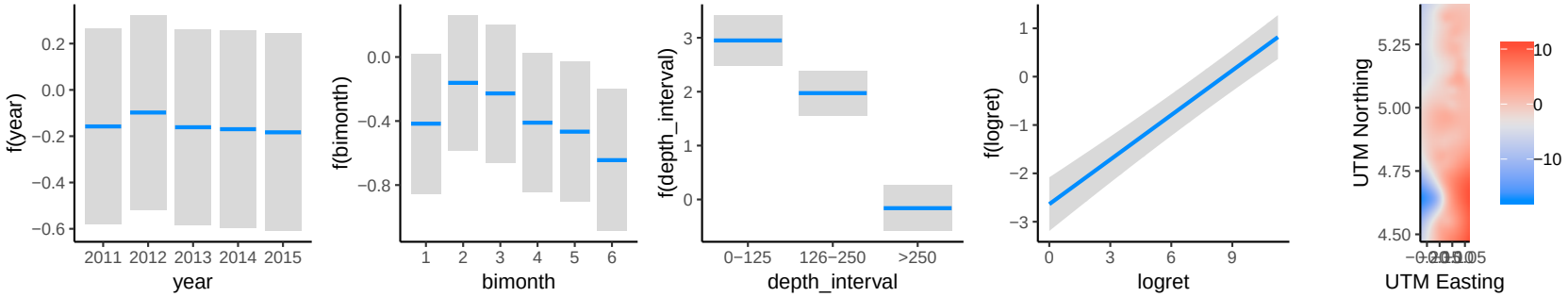
RF_{strat2}

```
Mean of squared residuals: 6845.978
          % Var explained: 9.13
```



GAM_{strat2}

R-sq.(adj) = 0.025 Deviance explained = 34.5% -REML = 88364 Scale est. = 45.918 n = 35440

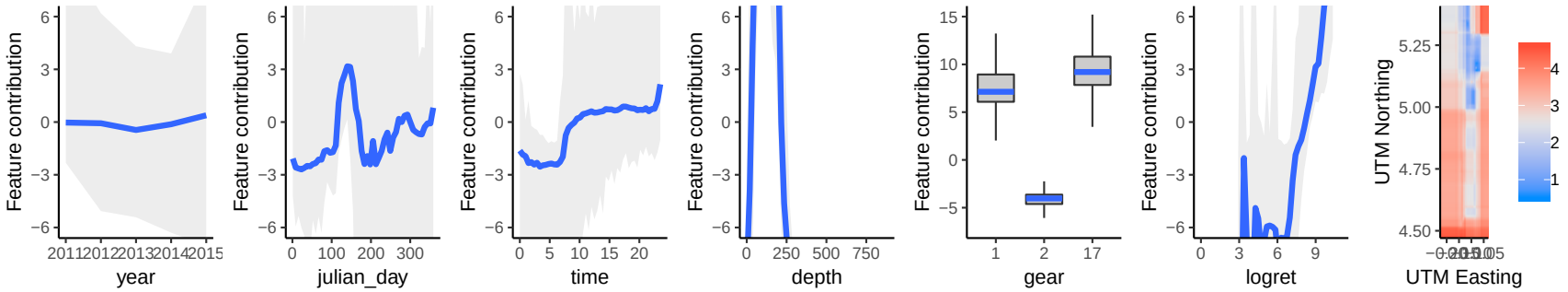


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.4	49	23.9	3.45e-208	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	-2.24	0.213	-10.5	9.13e-26	***
bimonth2	0.255	0.0935	2.73	0.00638	**
bimonth3	0.189	0.0953	1.98	0.0475	*
bimonth4	0.0062	0.101	0.0613	0.951	
bimonth5	-0.05	0.103	-0.485	0.628	
bimonth6	-0.228	0.111	-2.05	0.0403	*
year2012	0.06	0.0756	0.794	0.427	
year2013	-0.0036	0.0744	-0.0485	0.961	
year2014	-0.012	0.0785	-0.152	0.879	
year2015	-0.0255	0.079	-0.323	0.747	
depth_interval0-125	3.11	0.131	23.8	4.06e-124	***
depth_interval126-250	2.13	0.0775	27.6	2.21e-165	***
logret	0.307	0.0238	12.9	6.58e-38	***

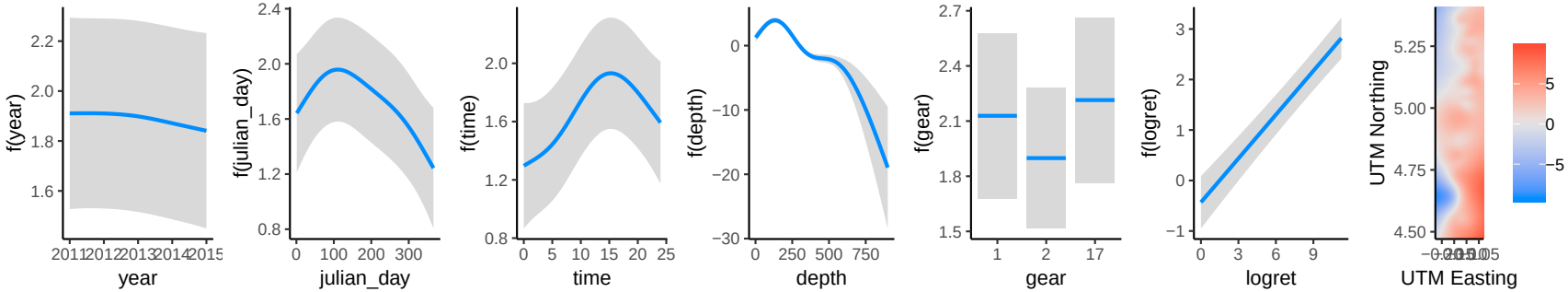
$RF_{nonlinear}$

```
Mean of squared residuals: 6596.819
          % Var explained: 12.44
```



GAM$_{\text{nonlinear}}$

R-sq.(adj) = 0.0547 Deviance explained = 39.8% -REML = 87121 Scale est. = 38.584 n = 35440



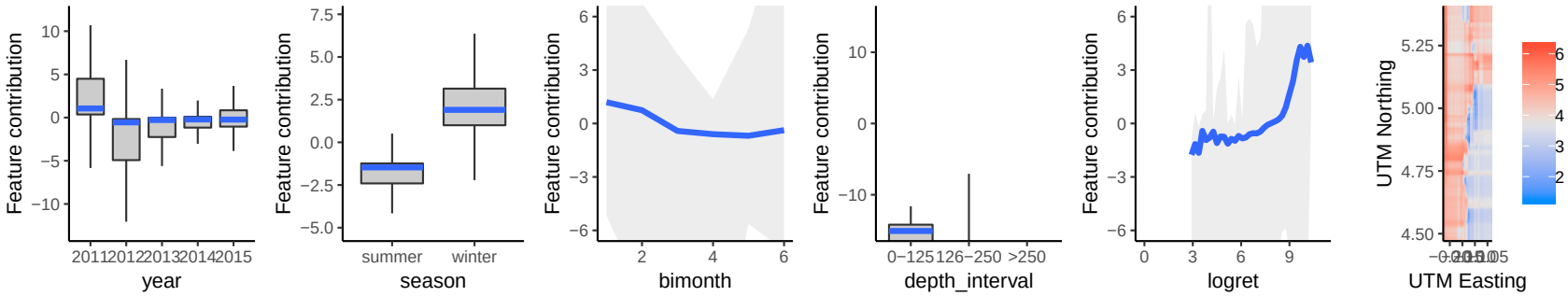
term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48.1	49	22.1	7.34e-190	***
s(year)	1.7	2.09	0.693	0.495	
s(depth)	3.98	4	375	5.5399580868179e-320	***
s(julian_day)	3.65	3.93	12.4	5.92e-10	***
s(time)	3.73	3.96	16.3	3.7e-13	***

term	estimate	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

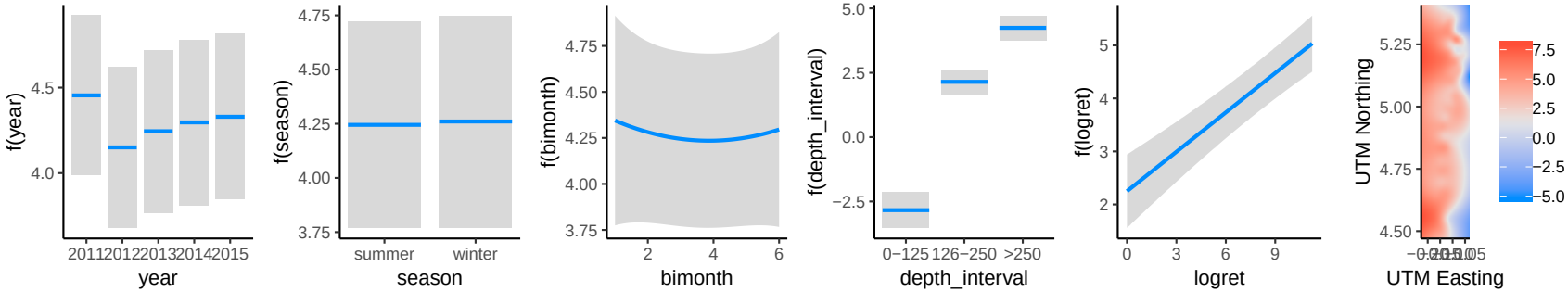
$\mathbf{RF_{strat1}}$

```
Mean of squared residuals: 7438.715
          % Var explained: 34.52
```



$\mathbf{GAM_{strat1}}$

R-sq.(adj) = 0.176 Deviance explained = 53.9% -REML = 85417 Scale est. = 85.96 n = 35440

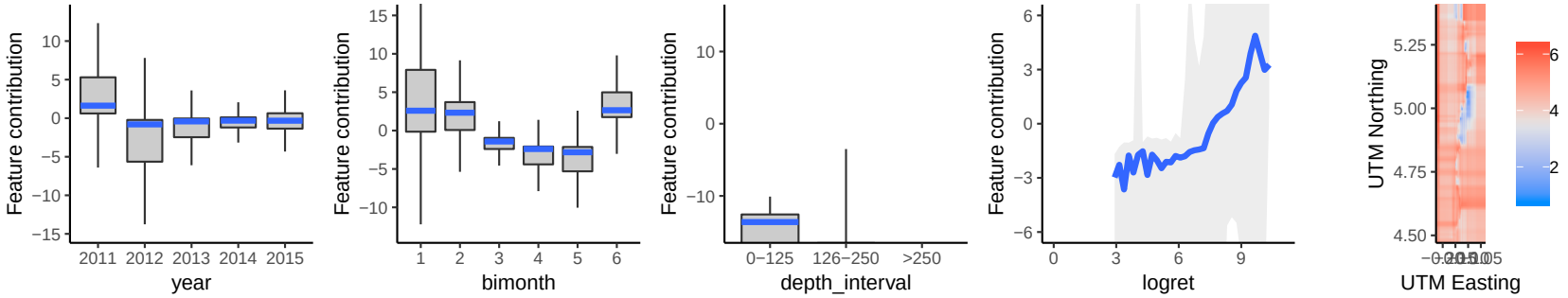


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	8.84	2.42e-62	***

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^2)	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	***

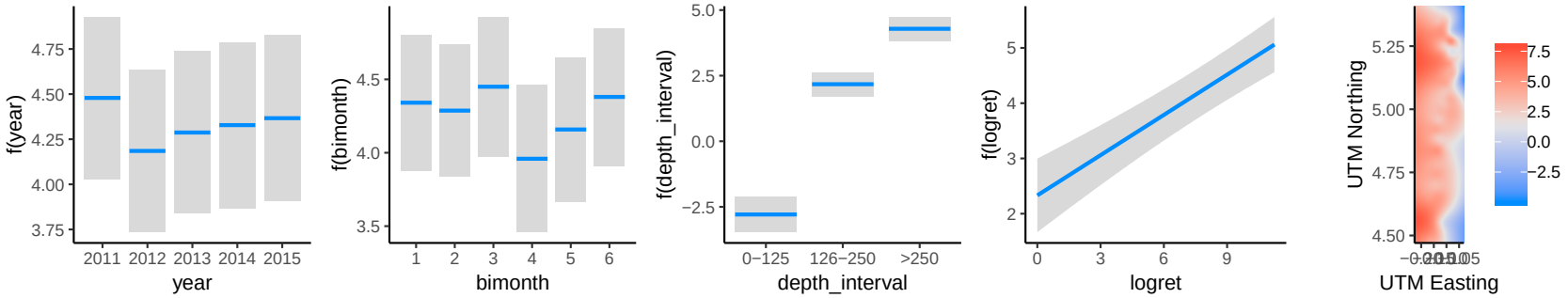
RF_{strat2}

```
Mean of squared residuals: 7511.038
          % Var explained: 33.88
```



GAM_{strat2}

R-sq.(adj) = 0.178 Deviance explained = 54.1% -REML = 85368 Scale est. = 80.769 n = 35440

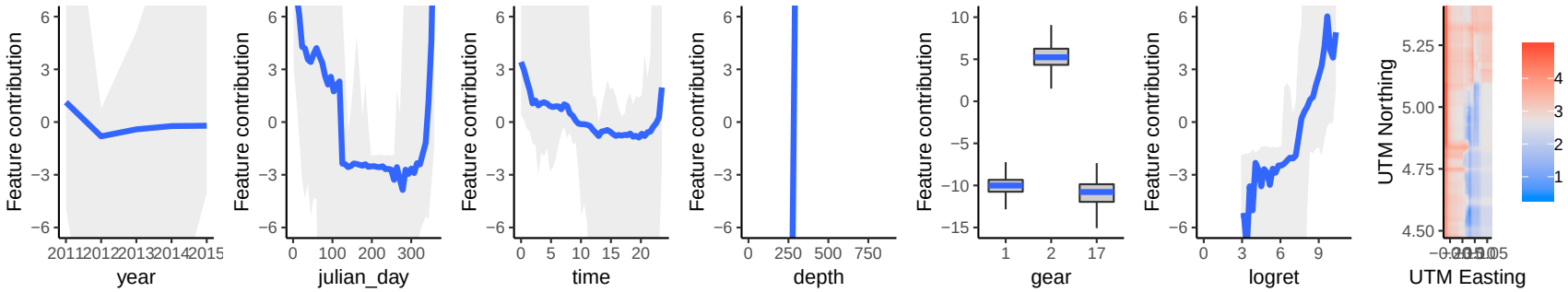


term	edf	ref.df	statistic	p.value	signif
s(avg_lat,avg_long)	48	48.9	9.45	4.57e-68	***

term	estimate	std.error	statistic	p.value	signif
(Intercept)	1.67	0.281	5.95	2.69e-09	***
bimonth2	-0.0539	0.0968	-0.557	0.578	
bimonth3	0.109	0.113	0.961	0.337	
bimonth4	-0.382	0.141	-2.71	0.00671	**
bimonth5	-0.183	0.133	-1.37	0.17	
bimonth6	0.0394	0.118	0.335	0.738	
year2012	-0.294	0.102	-2.89	0.0039	**
year2013	-0.192	0.0985	-1.95	0.0515	.
year2014	-0.151	0.107	-1.41	0.159	
year2015	-0.112	0.11	-1.02	0.309	
depth_interval0-125	-7.08	0.266	-26.6	4.55e-154	***
depth_interval126-250	-2.11	0.0929	-22.8	8.82e-114	***
logret	0.243	0.0321	7.55	4.32e-14	***

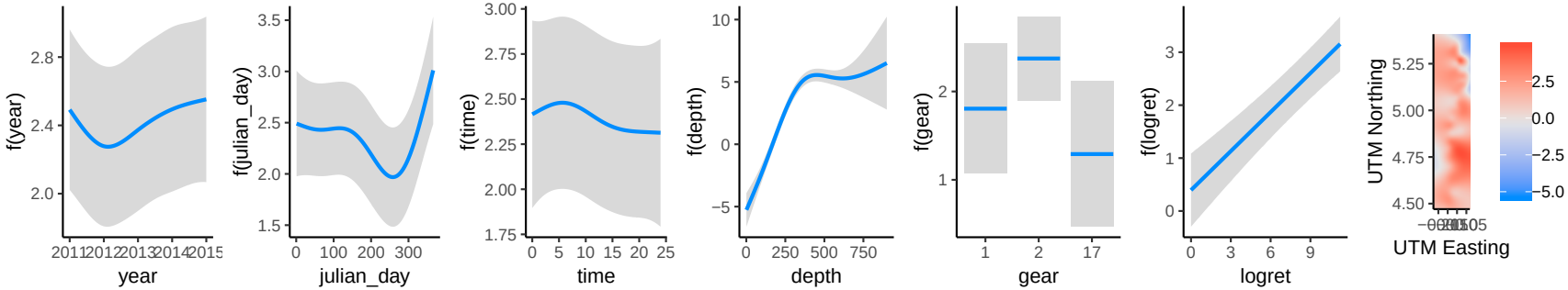
$RF_{nonlinear}$

```
Mean of squared residuals: 6860.649
          % Var explained: 39.61
```



GAM$_{\text{nonlinear}}$

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



term	edf	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	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	***