

GAMs and NonLinearity

Sar, North, Henry, Quinn

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```
library(ISLR)
library(dplyr)
```

```
## Warning: replacing previous import 'lifecycle::last_warnings' by
## 'rlang::last_warnings' when loading 'pillar'
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(readr)
```

```
## Warning: replacing previous import 'lifecycle::last_warnings' by
## 'rlang::last_warnings' when loading 'hms'
```

```
library(broom)
```

```
## Warning: package 'broom' was built under R version 4.1.2
```

```
library(ggplot2)
library(splines)
library(tidymodels)
```

```
## Registered S3 method overwritten by 'tune':
##   method                from
##   required_pkgs.model_spec parsnip
```

```
## -- Attaching packages ----- tidymodels 0.1.4 --
```

```
## v dials          0.0.10      v tibble      3.1.6
## v infer          1.0.0      v tidyr       1.1.4
## v modeldata      0.1.1      v tune      0.1.6
## v parsnip        0.1.7      v workflows 0.2.4
## v purrr          0.3.4      v workflowsets 0.1.0
## v recipes        0.1.17     v yardstick  0.0.9
## v rsample        0.1.1

## -- Conflicts ----- tidymodels_conflicts() --
## x purrr::discard() masks scales::discard()
## x dplyr::filter()  masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## x yardstick::spec() masks readr::spec()
## x recipes::step()  masks stats::step()
## * Use suppressPackageStartupMessages() to eliminate package startup messages
```

```
tidymodels_prefer()
```

```
COVID_State <- read.csv("COVID - State - Daily.csv", na.strings = ".")
```

```
Employment_State <- read.csv("Employment - State - Daily.csv", na.strings = ".")
```

```
Mobility_State <- read.csv("Google Mobility - State - Daily.csv", na.strings = ".")
```

```
Spending_State <- read.csv("Affinity - State - Daily.csv", na.strings = ".")
```

```
COVID_State$Date<-as.Date(with(COVID_State,paste(year,month,day,sep="-")), "%Y-%m-%d")
```

```
Employment_State$Date<-as.Date(with(Employment_State,paste(year,month,day,sep="-")), "%Y-%m-%d")
```

```
Mobility_State$Date<-as.Date(with(Mobility_State,paste(year,month,day,sep="-")), "%Y-%m-%d")
```

```
Spending_State$Date<-as.Date(with(Spending_State,paste(year,month,day,sep="-")), "%Y-%m-%d")
```

```
full_data <- merge(merge(merge(COVID_State, Employment_State, by=c("Date", "statefips")), Mobility_State,
```

```
## Warning in merge.data.frame(merge(merge(COVID_State, Employment_State, by =
## c("Date", : column names 'year.x', 'month.x', 'day.x', 'year.y', 'month.y',
## 'day.y' are duplicated in the result
```

```
head(full_data)
```

```
##           Date statefips year.x month.x day.x new_case_count new_death_count
## 1 2020-02-24          1  2020        2    24              NA              NA
## 2 2020-02-24         10  2020        2    24              NA              NA
## 3 2020-02-24         11  2020        2    24              NA              NA
## 4 2020-02-24         12  2020        2    24              NA              NA
## 5 2020-02-24         13  2020        2    24              NA              NA
## 6 2020-02-24         15  2020        2    24              NA              NA
##   case_count death_count vaccine_count fullvaccine_count booster_first_count
## 1          NA          NA            NA              NA              NA
## 2          NA          NA            NA              NA              NA
```

## 3	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA
##	new_vaccine_count	new_fullvaccine_count	new_booster_first_count		
## 1	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	
##	new_test_count	test_count	hospitalized_count	new_case_rate	case_rate
## 1	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA
## 6	NA	NA	0	NA	NA
##	new_death_rate	death_rate	new_test_rate	test_rate	new_vaccine_rate
## 1	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA
##	vaccine_rate	new_fullvaccine_rate	fullvaccine_rate	new_booster_first_rate	
## 1	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	
##	booster_first_rate	hospitalized_rate	year.y	month.y	day.y
## 1	NA	NA	2020	2	24
## 2	NA	NA	2020	2	24
## 3	NA	NA	2020	2	24
## 4	NA	NA	2020	2	24
## 5	NA	NA	2020	2	24
## 6	NA	0	2020	2	24
##	emp_incq2	emp_incq3	emp_incq4	emp_incmiddle	emp_incbelowmed
## 1	0.02320	0.01680	NA	0.01960	0.013600
## 2	0.00570	0.01680	0.0242	0.01170	-0.011400
## 3	NA	NA	NA	NA	NA
## 4	-0.00458	0.01070	0.0164	0.00324	-0.003550
## 5	0.00520	0.00873	0.0140	0.00710	-0.000838
## 6	-0.04920	-0.00520	NA	-0.02980	-0.058300
##	emp_ss40	emp_ss60	emp_ss65	emp_ss70	year.x
## 1	0.001540	-0.00399	0.05300	-0.01620	2020
## 2	0.015400	0.01340	0.01030	-0.05550	2020
## 3	NA	NA	NA	NA	2020
## 4	-0.002320	0.00134	0.00576	0.01620	2020
## 5	-0.000237	0.00168	0.00889	0.00964	2020
## 6	0.054800	NA	NA	-0.01530	2020
##	gps_retail_and_recreation	gps_grocery_and_pharmacy	gps_parks		

```

## 1          0.00286          -0.00714      0.0557
## 2          0.03710          0.01290      0.2340
## 3         -0.01140         -0.03290      0.1400
## 4          0.02710          0.00714      0.0943
## 5         -0.00571         -0.02290      0.0186
## 6          0.01140         -0.00571      0.0814
##   gps_transit_stations gps_workplaces gps_residential gps_away_from_home year.y
## 1          0.06000          0.01290          0.00857      -0.00798      2020
## 2          0.07000          0.02860         -0.00571          0.00850      2020
## 3          0.00571         -0.01430          0.00714      -0.00492      2020
## 4          0.03430          0.01000          0.00143      -0.00138      2020
## 5          0.01710         -0.01140          0.01000      -0.00781      2020
## 6          0.02570          0.00714          0.00143      -0.00049      2020
##   month.y day.y freq spend_all spend_aap spend_acf spend_aer spend_apg
## 1         2   24   d  -0.0198  -0.1320  -0.0220  -0.1000  -0.0810
## 2         2   24   d  -0.0461   0.1130  -0.0279  -0.6280   0.4140
## 3         2   24   d   0.0192  -0.1280  -0.0113   0.0740  -0.0855
## 4         2   24   d  -0.0452  -0.0847  -0.0493  -0.1020  -0.0675
## 5         2   24   d  -0.0163  -0.0321  -0.0334   0.0287  -0.0308
## 6         2   24   d  -0.0504  -0.1210  -0.0447  -0.1650  -0.0851
##   spend_durables spend_nondurables spend_grf spend_gen spend_hic spend_hcs
## 1         -0.0317         -0.04750  -0.0223  -0.01050  -0.06180  -0.07310
## 2          0.0208          0.13400  -0.0284   0.63600   0.13400  -0.01060
## 3          0.0311         -0.00364   0.0294   0.00856   0.59500   0.02630
## 4         -0.0492         -0.04720  -0.0468  -0.03810  -0.08320   0.00175
## 5         -0.0164         -0.02450  -0.0110  -0.03000  -0.00361  -0.02010
## 6         -0.0118         -0.04380  -0.0173  -0.04770   0.16600  -0.08730
##   spend_inpersonmisc spend_remoteservices spend_sgh spend_tws
## 1          0.0062          0.02110  -0.0453  -0.1020
## 2         -0.1380         -0.15500  -0.1540  -0.0929
## 3          0.2100         -0.03610  -0.1230  -0.1360
## 4         -0.0815         -0.04600  -0.0426  -0.1030
## 5         -0.0658         -0.00774   0.0940  -0.1060
## 6         -0.0645         -0.04000  -0.2270  -0.0909
##   spend_retail_w_grocery spend_retail_no_grocery spend_all_incmiddle
## 1         -0.03910          -0.0459          -0.02970
## 2          0.10200          0.1560          -0.06480
## 3         -0.00169          -0.0124          -0.06430
## 4         -0.04390          -0.0421          -0.03880
## 5         -0.01640          -0.0176          -0.01870
## 6         -0.03610          -0.0498           0.00268
##   spend_all_q1 spend_all_q2 spend_all_q3 spend_all_q4 provisional
## 1         -0.0158         -0.0717   0.036100   0.009840           0
## 2          0.2240         -0.0565  -0.068700  -0.016000           0
## 3         -0.0265         -0.5850  -0.047300   0.039400           0
## 4         -0.0677         -0.0420  -0.035100  -0.035700           0
## 5         -0.0386         -0.0234  -0.015600  -0.000937           0
## 6           NA          0.0134   0.000257  -0.076700           0

```

```

full_data1 <- full_data %>%
  select(-year.x, -month.x, -day.x, - year.y, -month.y, -day.y, -year.x )

```

```

minnesota <- full_data1 %>%

```

```
filter(statefips==27)
```

```
minnesota_cut <- minnesota %>%
  filter(Date > "2020-04-13")
```

```
set.seed(123)
```

```
# Don't necessarily need to use gam_spec, can use lm_spec instead
```

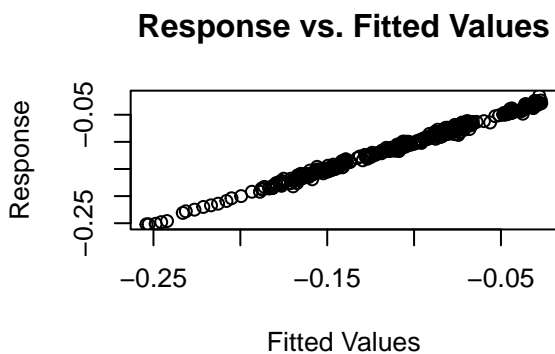
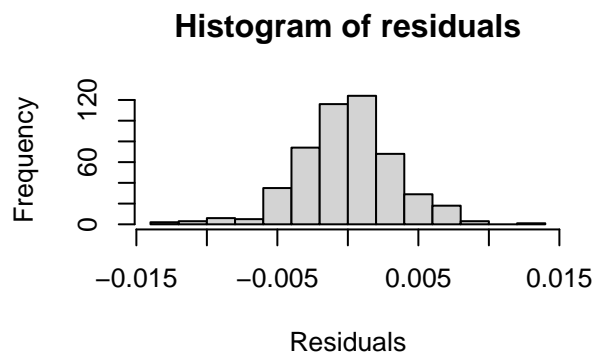
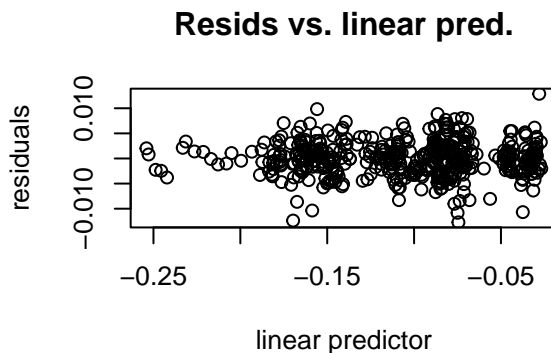
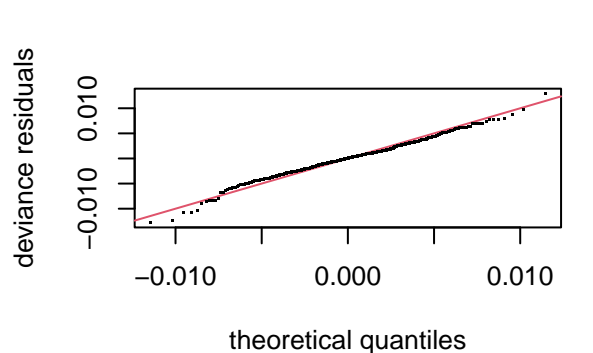
```
gam_spec <-
  gen_additive_mod() %>%
  set_engine(engine = 'mgcv') %>%
  set_mode('regression')
```

```
lm_spec <-
  linear_reg() %>%
  set_engine(engine = 'lm') %>%
  set_mode('regression')
```

```
gam_mod <- fit(gam_spec,
  gps_away_from_home ~ s(case_rate) + s(hospitalized_rate, k=20) + s(emp_incq1, k=20) + s(emp_
  data = minnesota_cut)
```

```
# Diagnostics: Check to see if the number of knots is large enough
```

```
par(mfrow=c(2,2))
gam_mod %>% pluck('fit') %>% mgcv::gam.check()
```



```

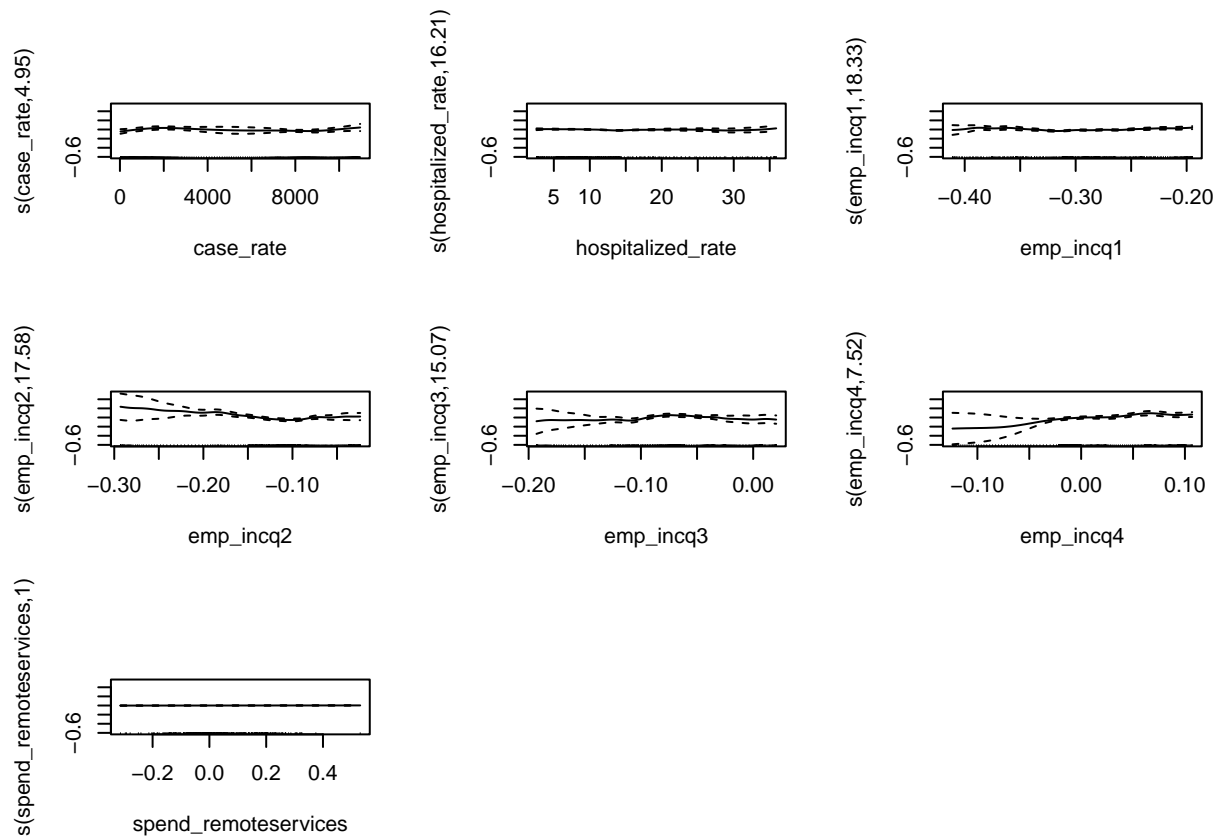
##
## Method: GCV Optimizer: magic
## Smoothing parameter selection converged after 17 iterations.
## The RMS GCV score gradient at convergence was 3.618495e-08 .
## The Hessian was not positive definite.
## Model rank = 104 / 104
##
## Basis dimension (k) checking results. Low p-value (k-index<1) may
## indicate that k is too low, especially if edf is close to k'.
##
##           k'   edf k-index p-value
## s(case_rate)      9.00  4.95   0.51 <2e-16 ***
## s(hospitalized_rate) 19.00 16.21   1.03   0.65
## s(emp_incq1)      19.00 18.33   0.76 <2e-16 ***
## s(emp_incq2)      19.00 17.58   0.80 <2e-16 ***
## s(emp_incq3)      19.00 15.07   0.82 <2e-16 ***
## s(emp_incq4)       9.00  7.52   0.81 <2e-16 ***
## s(spend_remoteservices) 9.00  1.00   0.97   0.21
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Parameter (linear) estimates and then Smooth Terms (H0: no relationship)
gam_mod %>% pluck('fit') %>% summary()

##
## Family: gaussian
## Link function: identity
##
## Formula:
## gps_away_from_home ~ s(case_rate) + s(hospitalized_rate, k = 20) +
##   s(emp_incq1, k = 20) + s(emp_incq2, k = 20) + s(emp_incq3,
##   k = 20) + s(emp_incq4) + s(spend_remoteservices)
##
## Parametric coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.104091  0.000169  -615.9   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##           edf Ref.df    F p-value
## s(case_rate)      4.95  5.911 3.199 0.004059 **
## s(hospitalized_rate) 16.21 18.004 3.976 7.14e-07 ***
## s(emp_incq1)      18.33 18.799 9.028 < 2e-16 ***
## s(emp_incq2)      17.58 18.313 7.447 < 2e-16 ***
## s(emp_incq3)      15.07 16.728 4.682 < 2e-16 ***
## s(emp_incq4)       7.52  8.058 4.203 0.000407 ***
## s(spend_remoteservices) 1.00  1.001 1.368 0.242846
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.995 Deviance explained = 99.5%
## GCV = 1.6631e-05 Scale est. = 1.3826e-05 n = 484

```

```
# Looking at possible non-linear functions
gam_mod %>% pluck('fit') %>% plot(all.terms = TRUE, pages = 1)
```



```
formula = gps_away_from_home ~ case_rate + hospitalized_rate + emp_incq1 + emp_incq2 + emp_incq3 + emp_
gam_rec <- recipe(formula, data=minnesota_cut)
```

```
gam_rec_new <- gam_rec %>%
  step_ns(case_rate, deg_free = 6) %>%
  step_ns(hospitalized_rate, deg_free = 9) %>%
  step_ns(emp_incq1, deg_free = 8) %>%
  step_ns(emp_incq2, deg_free = 9) %>%
  step_ns(emp_incq3, deg_free = 9) %>%
  step_ns(emp_incq4, deg_free = 7) %>%
  step_ns(spend_remoteservices, deg_free = 6)
```

```
data_cv8 <- minnesota_cut %>%
  vfold_cv(v = 8)
```

```
gam_wf <- workflow() %>%
  add_model(lm_spec) %>%
  add_recipe(gam_rec)
```

```
fit_resamples(
  gam_wf,
  resamples = data_cv8, # cv folds
```

```

  metrics = metric_set(mae,rmse,rsq)
) %>% collect_metrics()

```

```
## Warning: package 'rlang' was built under R version 4.1.2
```

```
## # A tibble: 3 x 6
##   .metric .estimator  mean     n std_err .config
##   <chr>   <chr>      <dbl> <int>   <dbl> <chr>
## 1 mae     standard  0.0115     8 0.000386 Preprocessor1_Model1
## 2 rmse     standard  0.0145     8 0.000523 Preprocessor1_Model1
## 3 rsq      standard  0.917      8 0.00669  Preprocessor1_Model1
```

```

gam_new_wf <- workflow() %>%
  add_model(lm_spec) %>%
  add_recipe(gam_rec_new)

fit_resamples(
  gam_new_wf,
  resamples = data_cv8, # cv folds
  metrics = metric_set(mae,rmse,rsq)
) %>% collect_metrics()

```

```
## # A tibble: 3 x 6
##   .metric .estimator  mean     n std_err .config
##   <chr>   <chr>      <dbl> <int>   <dbl> <chr>
## 1 mae     standard  0.00424     8 0.000173 Preprocessor1_Model1
## 2 rmse     standard  0.00555     8 0.000198 Preprocessor1_Model1
## 3 rsq      standard  0.988      8 0.000842 Preprocessor1_Model1
```