Tidy Models Parameter Tuning

Setup

```
suppressPackageStartupMessages(library(tidyworse))
suppressPackageStartupMessages(library(tidymodels))
suppressPackageStartupMessages(library(ggformula))
suppressPackageStartupMessages(library(GGally))
suppressPackageStartupMessages(library(future))
library(ISLR2)

## Multitasking setup (old)
#suppressPackageStartupMessages(library(doMC))
#cores <- detectCores(logical=TRUE)
#cat('Available cores: ',cores,'\n')
#registerDoMC(cores=cores-1)

# New multitasking setup
cores <- availableCores()
cat('Available cores: ',cores,'\n')</pre>
```

Available cores: 8

```
plan(strategy=multisession, workers=cores-1)

# Load data and setup training/test/fold partitions
set.seed(2025)
hitters <- na.omit(Hitters)

hitters_split <- initial_split(hitters, strata = "Salary")</pre>
```

```
hitters_train <- training(hitters_split)
hitters_test <- testing(hitters_split)
hitters_fold <- vfold_cv(hitters_train, v = 10)</pre>
```

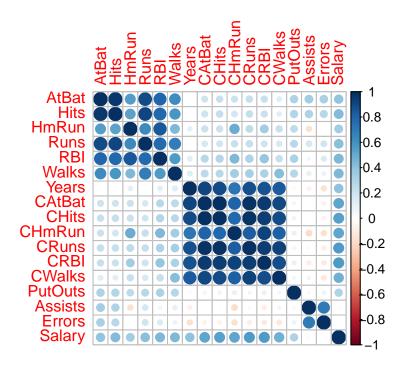
Source

Exploratory analysis

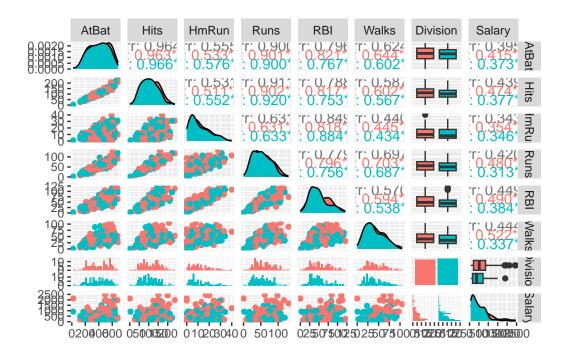
```
glimpse(hitters)
```

```
Rows: 263
Columns: 20
$ AtBat
            <int> 315, 479, 496, 321, 594, 185, 298, 323, 401, 574, 202, 418, ~
$ Hits
            <int> 81, 130, 141, 87, 169, 37, 73, 81, 92, 159, 53, 113, 60, 43,~
            <int> 7, 18, 20, 10, 4, 1, 0, 6, 17, 21, 4, 13, 0, 7, 20, 2, 8, 16~
$ HmRun
$ Runs
            <int> 24, 66, 65, 39, 74, 23, 24, 26, 49, 107, 31, 48, 30, 29, 89,~
$ RBI
            <int> 38, 72, 78, 42, 51, 8, 24, 32, 66, 75, 26, 61, 11, 27, 75, 8~
$ Walks
            <int> 39, 76, 37, 30, 35, 21, 7, 8, 65, 59, 27, 47, 22, 30, 73, 15~
            <int> 14, 3, 11, 2, 11, 2, 3, 2, 13, 10, 9, 4, 6, 13, 15, 5, 8, 1,~
$ Years
$ CAtBat
            <int> 3449, 1624, 5628, 396, 4408, 214, 509, 341, 5206, 4631, 1876~
$ CHits
            <int> 835, 457, 1575, 101, 1133, 42, 108, 86, 1332, 1300, 467, 392~
            <int> 69, 63, 225, 12, 19, 1, 0, 6, 253, 90, 15, 41, 4, 36, 177, 5~
$ CHmRun
            <int> 321, 224, 828, 48, 501, 30, 41, 32, 784, 702, 192, 205, 309,~
$ CRuns
            <int> 414, 266, 838, 46, 336, 9, 37, 34, 890, 504, 186, 204, 103, ~
$ CRBI
$ CWalks
            <int> 375, 263, 354, 33, 194, 24, 12, 8, 866, 488, 161, 203, 207, ~
$ League
            <fct> N, A, N, N, A, N, A, N, A, A, N, N, A, N, N, A, N, N, A, N, ~
$ Division <fct> W, W, E, E, W, E, W, W, E, E, W, E, E, E, W, W, W, E, W, W, ~
$ PutOuts
            <int> 632, 880, 200, 805, 282, 76, 121, 143, 0, 238, 304, 211, 121~
$ Assists
            <int> 43, 82, 11, 40, 421, 127, 283, 290, 0, 445, 45, 11, 151, 45,~
            <int> 10, 14, 3, 4, 25, 7, 9, 19, 0, 22, 11, 7, 6, 8, 10, 16, 2, 5~
$ Errors
            <dbl> 475.000, 480.000, 500.000, 91.500, 750.000, 70.000, 100.000,~
$ Salary
$ NewLeague <fct> N, A, N, N, A, A, A, N, A, A, N, N, A, N, N, A, N, N, N, N, N, ~
```

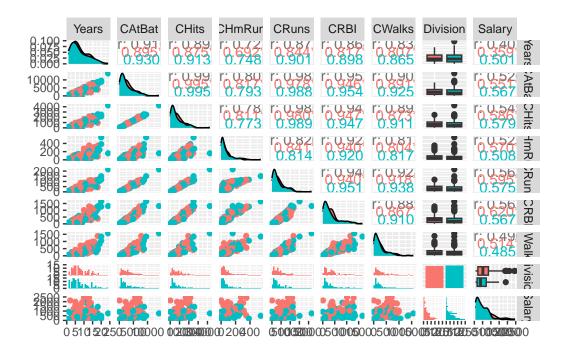
```
library(corrplot)
hitters %>%
  dplyr::select(-c(League, Division, New League)) %>%
  cor() %>% corrplot()
```



hitters %>%
 select(AtBat:Walks,Division,Salary) %>%
 ggpairs(mapping=aes(col=Division))



```
hitters %>%
select(Years:CWalks,Division,Salary) %>%
ggpairs(mapping=aes(col=Division))
```



```
library(car)
vif(lm(Salary~.,data=hitters))
```

```
AtBat
                 Hits
                           HmRun
                                                             Walks
                                                                        Years
                                       Runs
                                                    RBI
22.944366
           30.281255
                        7.758668
                                  15.246418
                                             11.921715
                                                          4.148712
                                                                     9.313280
                          CHmRun
    CAtBat
                CHits
                                      CRuns
                                                   CRBI
                                                            CWalks
                                                                       League
251.561160 502.954289
                       46.488462 162.520810 131.965858 19.744105
                                                                     4.134115
 Division
              PutOuts
                         Assists
                                     Errors NewLeague
  1.075398
             1.236317
                        2.709341
                                   2.214543
                                               4.099063
```

Ridge regression

```
ridge_spec <- linear_reg(mixture = 0, penalty = 0) %>%
  set_mode("regression") %>%
  set_engine("glmnet")
```

```
ridge_fit <- fit(ridge_spec, Salary ~ ., data = hitters)
tidy(ridge_fit)</pre>
```

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

expand, pack, unpack

Loaded glmnet 4.1-8

A tibble: 20 x 3

	term	estimate	penalty
	<chr></chr>	<dbl></dbl>	<dbl></dbl>
1	(Intercept)	81.1	0
2	AtBat	-0.682	0
3	Hits	2.77	0
4	HmRun	-1.37	0
5	Runs	1.01	0
6	RBI	0.713	0
7	Walks	3.38	0
8	Years	-9.07	0
9	CAtBat	-0.00120	0
10	CHits	0.136	0
11	CHmRun	0.698	0
12	CRuns	0.296	0
13	CRBI	0.257	0
14	CWalks	-0.279	0
15	LeagueN	53.2	0
16	DivisionW	-123.	0
17	PutOuts	0.264	0
18	Assists	0.170	0
19	Errors	-3.69	0
20	${\tt NewLeagueN}$	-18.1	0

```
tidy(ridge_fit, penalty = 11498)
```

A tibble: 20 x 3

term estimate penalty

	<chr></chr>	<dbl></dbl>	<dbl></dbl>
1	(Intercept)	407.	11498
2	AtBat	0.0370	11498
3	Hits	0.138	11498
4	HmRun	0.525	11498
5	Runs	0.231	11498
6	RBI	0.240	11498
7	Walks	0.290	11498
8	Years	1.11	11498
9	CAtBat	0.00314	11498
10	CHits	0.0117	11498
11	CHmRun	0.0876	11498
12	CRuns	0.0234	11498
13	CRBI	0.0242	11498
14	CWalks	0.0250	11498
15	LeagueN	0.0866	11498
16	DivisionW	-6.23	11498
17	PutOuts	0.0165	11498
18	Assists	0.00262	11498
19	Errors	-0.0206	11498
20	NewLeagueN	0.303	11498

tidy(ridge_fit, penalty = 705)

A tibble: 20 x 3

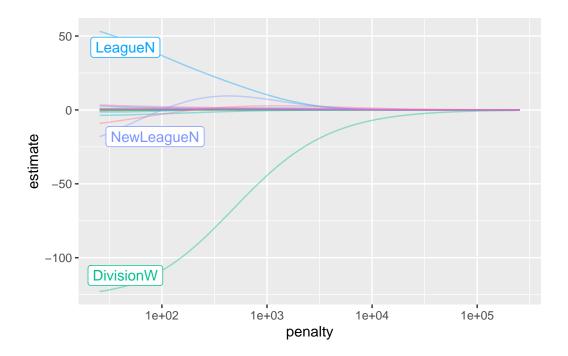
	term	estimate	penalty
	<chr></chr>	<dbl></dbl>	<dbl></dbl>
1	(Intercept)	54.4	705
2	AtBat	0.112	705
3	Hits	0.656	705
4	HmRun	1.18	705
5	Runs	0.937	705
6	RBI	0.847	705
7	Walks	1.32	705
8	Years	2.58	705
9	CAtBat	0.0108	705
10	CHits	0.0468	705
11	CHmRun	0.338	705
12	CRuns	0.0937	705
13	CRBI	0.0979	705
14	CWalks	0.0718	705
15	LeagueN	13.7	705

```
16 DivisionW -54.7 705
17 PutOuts 0.119 705
18 Assists 0.0161 705
19 Errors -0.704 705
20 NewLeagueN 8.61 705
```

tidy(ridge_fit, penalty = 50)

A tibble: 20 x 3 estimate penalty term <chr> <dbl> <dbl> 50 1 (Intercept) 48.2 2 AtBat 50 -0.3543 Hits 1.95 50 4 HmRun -1.2950 5 Runs 1.16 50 6 RBI 0.809 50 7 Walks 2.71 50 8 Years -6.20 50 9 CAtBat 0.00609 50 10 CHits 0.107 50 11 CHmRun 0.629 50 12 CRuns 0.217 50 13 CRBI 0.215 50 14 CWalks -0.149 50 15 LeagueN 45.9 50 16 DivisionW 50 -118. 0.250 17 PutOuts 50 18 Assists 0.121 50 19 Errors -3.2850 20 NewLeagueN -9.42 50

ridge_fit %>%
autoplot()



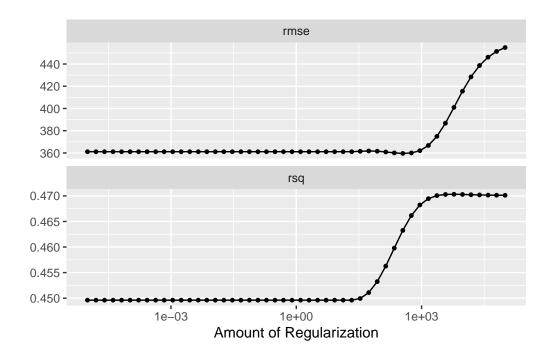
predict(ridge_fit, new_data = hitters)

predict(ridge_fit, new_data = hitters, penalty = 500)

```
# A tibble: 263 x 1 .pred
```

```
dbl>
 1 525.
 2 620.
 3 895.
 4 425.
 5 589.
 6 179.
 7 147.
 8 187.
 9 841.
10 840.
# i 253 more rows
ridge_recipe <-</pre>
  recipe(formula = Salary ~ ., data = hitters_train) %>%
  step_novel(all_nominal_predictors()) %>% # add 'new' level to factors and chr -> fctr
  step_dummy(all_nominal_predictors()) %>% # factors to dummy variables
  step_zv(all_predictors()) %>%
                                            # remove zero variance predictors
  step_normalize(all_predictors())
                                           # normalize all predictors
  ridge_spec <-
  linear_reg(penalty = tune(), mixture = 0) %>%
  set_mode("regression") %>%
  set_engine("glmnet")
  ridge_workflow <- workflow() %>%
  add_recipe(ridge_recipe) %>%
  add_model(ridge_spec)
  penalty_grid <- grid_regular(penalty(range = c(-5, 5)), levels = 50)</pre>
penalty_grid
# A tibble: 50 \times 1
     penalty
       <dbl>
 1 0.00001
 2 0.0000160
 3 0.0000256
 4 0.0000409
 5 0.0000655
 6 0.000105
 7 0.000168
```

```
8 0.000268
        9 0.000429
  10 0.000687
  # i 40 more rows
  tune_res <- tune_grid(</pre>
                ridge_workflow,
                resamples = hitters_fold,
                grid = penalty_grid,
                control(parallel_over = "everything")
 Warning: The `...` are not used in this function but one or more objects were
passed: ''
tune_res
  # Tuning results
  # 10-fold cross-validation
  # A tibble: 10 x 4
                          splits
                                                                                                                                                                           id
                                                                                                                                                                                                                                       .metrics
                                                                                                                                                                                                                                                                                                                                                                                                         .notes
                          t>
                                                                                                                                                                           <chr> <chr>>
                                                                                                                                                                                                                                                                                                                                                                                                        st>
         1 <split [176/20] > Fold01 <tibble [100 x 5] > <tibble [0 x 3] >
        2 <split [176/20] > Fold02 <tibble [100 x 5] > <tibble [0 x 3] >
        3 <split [176/20] > Fold03 <tibble [100 x 5] > <tibble [0 x 3] >
        4 <split [176/20] > Fold04 <tibble [100 x 5] > <tibble [0 x 3] >
        5 \left[\frac{176}{20}\right] > Fold05 < tibble [100 x 5] > \left(\frac{1}{20}\right) > \left(
        6 <split [176/20] > Fold06 <tibble [100 x 5] > <tibble [0 x 3] >
        7 <split [177/19] > Fold07 <tibble [100 x 5] > <tibble [0 x 3] >
        8 <split [177/19]> Fold08 <tibble [100 \times 5]> <tibble [0 \times 3]>
        9 \left[\frac{177}{19}\right] > Fold09 \left[\frac{100 \times 5}{5}\right] > \left[\frac{100 \times 5}{5}\right] 
  10 <split [177/19] > Fold10 <tibble [100 x 5] > <tibble [0 x 3] >
  autoplot(tune_res)
```



collect_metrics(tune_res) %>% filter(.metric=="rmse")

```
# A tibble: 50 \times 7
    penalty .metric .estimator mean
                                           n std_err .config
       <dbl> <chr>
                     <chr>
                                               <dbl> <chr>
                                 <dbl> <int>
1 0.00001
             rmse
                     standard
                                 361.
                                          10
                                                28.3 Preprocessor1_Model01
2 0.0000160 rmse
                     standard
                                 361.
                                          10
                                                28.3 Preprocessor1_Model02
3 0.0000256 rmse
                                 361.
                    standard
                                          10
                                                28.3 Preprocessor1_Model03
4 0.0000409 rmse
                    standard
                                 361.
                                                28.3 Preprocessor1_Model04
                                          10
                                 361.
5 0.0000655 rmse
                    standard
                                          10
                                                28.3 Preprocessor1 Model05
                                 361.
                                                28.3 Preprocessor1_Model06
6 0.000105 rmse
                    standard
                                          10
7 0.000168 rmse
                     standard
                                 361.
                                                28.3 Preprocessor1_Model07
                                          10
                                 361.
8 0.000268
            rmse
                     standard
                                         10
                                                28.3 Preprocessor1_Model08
                                          10
9 0.000429
                     standard
                                  361.
                                                28.3 Preprocessor1_Model09
             rmse
10 0.000687
                     standard
                                  361.
                                          10
                                                28.3 Preprocessor1_Model10
             rmse
# i 40 more rows
```

```
best_penalty <- select_best(tune_res, metric = "rmse")
best_penalty</pre>
```

A tibble: 1 x 2

```
penalty .config
    <dbl> <chr>
     356. Preprocessor1_Model38
1
ridge_final <- finalize_workflow(ridge_workflow, best_penalty)</pre>
ridge_final_fit <- fit(ridge_final, data = hitters_train)</pre>
ridge_aug <- augment(ridge_final_fit, new_data = hitters_test)</pre>
ridge_result <- rbind(</pre>
  rmse(ridge_aug,truth = Salary, estimate = .pred),
  rsq(ridge_aug, truth = Salary, estimate = .pred))
ridge_result
# A tibble: 2 x 3
  .metric .estimator .estimate
  <chr>
          <chr>
                          <dbl>
1 rmse
          standard
2 rsq
          standard
                          0.515
Final ridge model
ridge_final_fit <- fit(ridge_final, data=hitters)</pre>
tidy(ridge_final_fit,conf_int=TRUE)
# A tibble: 20 x 3
                estimate penalty
   term
   <chr>>
                   <dbl>
                            <dbl>
 1 (Intercept)
                  536.
                             356.
 2 AtBat
                   12.3
                             356.
 3 Hits
                   37.6
                             356.
 4 HmRun
                   6.00
                            356.
 5 Runs
                   26.9
                            356.
 6 RBI
                   22.8
                            356.
 7 Walks
                   34.5
                            356.
                   7.44
 8 Years
                            356.
 9 CAtBat
                   25.9
                            356.
10 CHits
                   36.4
                            356.
11 CHmRun
                   32.7
                            356.
```

12 CRuns

14 CWalks

15 PutOuts

13 CRBI

37.1

38.2

14.9

45.0

356.

356.

356.

356.

```
16 Assists 3.96 356.

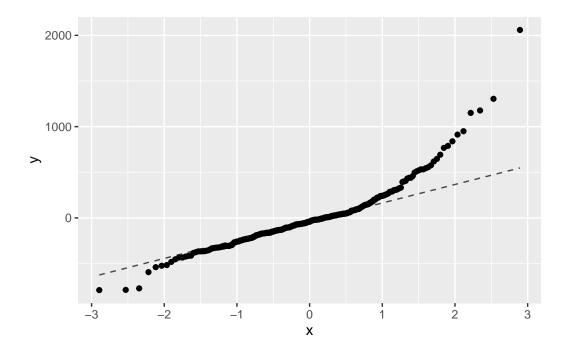
17 Errors -8.43 356.

18 League_N 10.6 356.

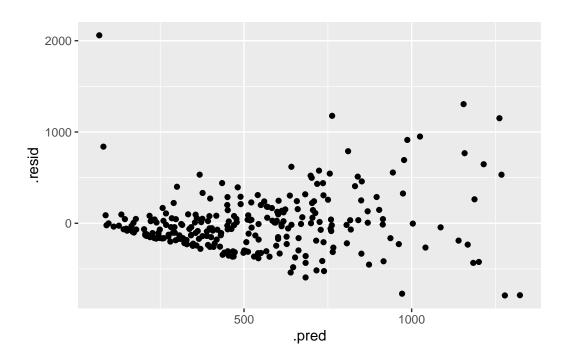
19 Division_W -38.2 356.

20 NewLeague_N 4.65 356.
```

```
ridge_aug <- augment(ridge_final_fit, new_data=hitters) %>% mutate(.resid=Salary-.pred)
ridge_aug %>%
   gf_qq(~.resid) %>% gf_qqline()
```



```
ridge_aug %>%
  gf_point(.resid~.pred)
```



The Lasso

```
lasso_recipe <-
    recipe(formula = Salary ~ ., data = hitters_train) %>%
    step_novel(all_nominal_predictors()) %>%
    step_dummy(all_nominal_predictors()) %>%
    step_zv(all_predictors()) %>%
    step_normalize(all_predictors())

lasso_spec <-
linear_reg(penalty = tune(), mixture = 1) %>%
    set_mode("regression") %>%
    set_engine("glmnet")

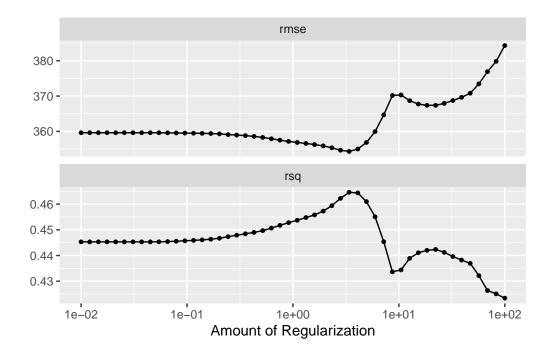
lasso_workflow <- workflow() %>%
    add_recipe(lasso_recipe) %>%
    add_model(lasso_spec)

penalty_grid <- grid_regular(penalty(range = c(-2, 2)), levels = 50)</pre>
```

```
tune_res <- tune_grid(
  lasso_workflow,
  resamples = hitters_fold,
  grid = penalty_grid,
  control(parallel_over = "everything")
)</pre>
```

Warning: The `...` are not used in this function but one or more objects were passed: $^{\prime\prime}$

autoplot(tune_res)



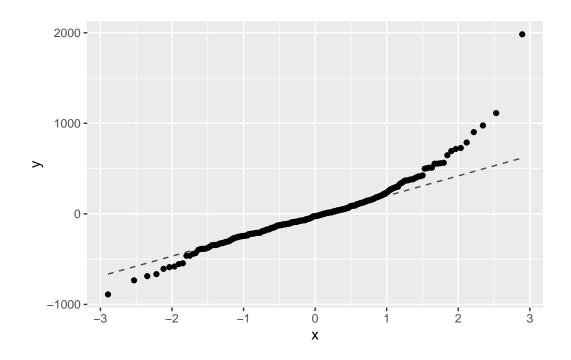
```
best_penalty <- select_best(tune_res, metric = "rsq")
lasso_final <- finalize_workflow(lasso_workflow, best_penalty)
lasso_final_fit <- fit(lasso_final, data = hitters_train)
lasso_aug <- augment(lasso_final_fit, new_data = hitters_test)
lasso_result <- rbind(
   rmse(lasso_aug,truth = Salary, estimate = .pred),
   rsq(lasso_aug, truth = Salary, estimate = .pred))
lasso_result</pre>
```

Final lasso model

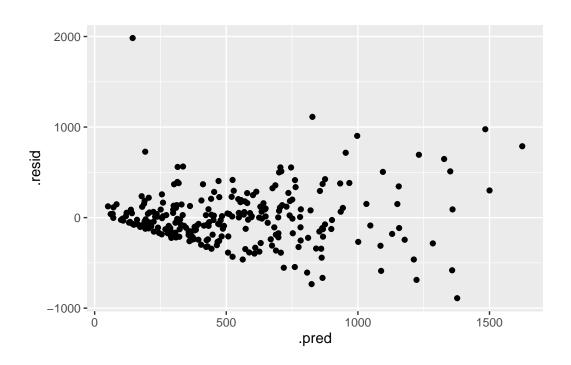
```
lasso_final_fit <- fit(lasso_final, data=hitters)
tidy(lasso_final_fit,conf_int=TRUE)</pre>
```

```
# A tibble: 20 x 3
               estimate penalty
  term
   <chr>
                  <dbl>
                          <dbl>
1 (Intercept)
                  536.
                           3.39
2 AtBat
                 -198.
                           3.39
3 Hits
                  235.
                           3.39
4 HmRun
                    0
                           3.39
5 Runs
                    0
                           3.39
6 RBI
                    0
                           3.39
                           3.39
7 Walks
                  94.9
8 Years
                  -40.7
                           3.39
9 CAtBat
                    0
                           3.39
10 CHits
                    0
                           3.39
11 CHmRun
                   35.2
                           3.39
12 CRuns
                  197.
                           3.39
13 CRBI
                           3.39
                  130.
14 CWalks
                 -118.
                           3.39
15 PutOuts
                   74.7
                           3.39
16 Assists
                   19.6
                           3.39
17 Errors
                  -11.5
                           3.39
18 League_N
                   15.4
                           3.39
19 Division_W
                  -59.7
                           3.39
20 NewLeague_N
                    0
                           3.39
```

```
lasso_aug <- augment(lasso_final_fit, new_data=hitters) %>% mutate(.resid=Salary-.pred)
lasso_aug %>%
    gf_qq(~.resid) %>% gf_qqline()
```



lasso_aug %>%
 gf_point(.resid~.pred)



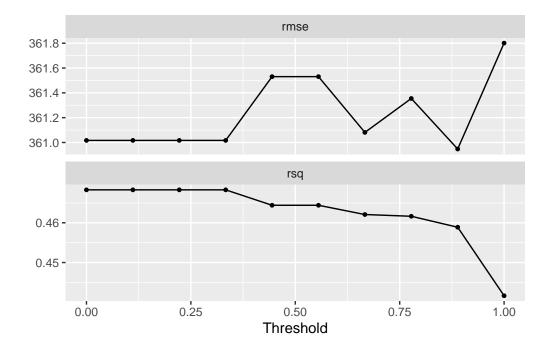
Principal components regression

```
lm spec <-</pre>
  linear_reg() %>%
  set_mode("regression") %>%
  set_engine("lm")
pca_recipe <-</pre>
  recipe(formula = Salary ~ ., data = hitters_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_zv(all_predictors()) %>%
  step_normalize(all_predictors()) %>%
  step_pca(all_predictors(), threshold = tune())
pca_workflow <-</pre>
  workflow() %>%
  add_recipe(pca_recipe) %>%
  add_model(lm_spec)
  threshold_grid <- grid_regular(threshold(), levels = 10)</pre>
threshold_grid
# A tibble: 10 x 1
   threshold
       <dbl>
 1
 2
       0.111
 3
       0.222
 4
      0.333
 5
      0.444
 6
      0.556
 7
      0.667
 8
      0.778
 9
       0.889
10
tune_res <- tune_grid(</pre>
  pca_workflow,
  resamples = hitters_fold,
```

```
grid = threshold_grid,
control(parallel_over = "everything")
)
```

Warning: The `...` are not used in this function but one or more objects were passed: ''

autoplot(tune_res)



```
best_threshold <- select_best(tune_res, metric = "rmse")
pca_final <- finalize_workflow(pca_workflow, best_threshold)
pca_final_fit <- fit(pca_final, data = hitters_train)
pca_aug <- augment(pca_final_fit, new_data = hitters_test)
pca_result <- rbind(
   rmse(pca_aug,truth = Salary, estimate = .pred),
   rsq(pca_aug, truth = Salary, estimate = .pred))
pca_result</pre>
```

```
# A tibble: 2 x 3
.metric .estimator .estimate
<chr> <chr> <chr>
```

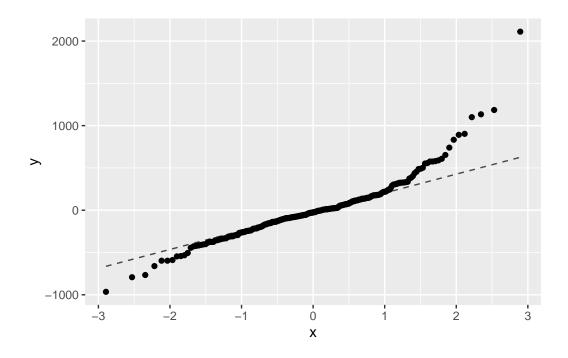
```
1 rmse standard 271.
2 rsq standard 0.521
```

Final pca model

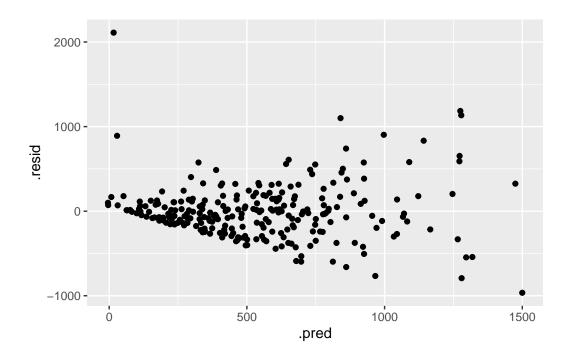
```
pca_final_fit <- fit(pca_final, data=hitters)
tidy(pca_final_fit,conf_int=TRUE)</pre>
```

A tibble: 8 x 5 estimate std.error statistic p.value term <chr> <dbl> <dbl> <dbl> <dbl> 20.6 1 (Intercept) 536. 26.0 9.20e-74 2 PC1 107. 7.65 13.9 3.24e-333 PC2 21.6 10.1 2.14 3.35e- 2 -1.68 9.39e- 2 4 PC3 -24.314.5 5 PC4 -37.1 16.5 -2.24 2.59e- 2 6 PC5 58.5 20.6 2.84 4.94e- 3 7 PC6 62.3 22.7 2.75 6.48e- 3 -0.994 3.21e- 1 8 PC7 -24.724.8

```
pca_aug <- augment(pca_final_fit, new_data=hitters) %>% mutate(.resid=Salary-.pred)
pca_aug %>%
    gf_qq(~.resid) %>% gf_qqline()
```



```
pca_aug %>%
  gf_point(.resid~.pred)
```



Partial least squares

Partial least squares in Tidy Models requires the package 'mixOmics' which is only available with Bioconductor which is a huge installation.

```
if (!require("BiocManager", quietly = TRUE))
   install.packages("BiocManager")
```

BiocManager::install("mixOmics")

```
pls_recipe <-
  recipe(formula = Salary ~ ., data = hitters_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_zv(all_predictors()) %>%
  step_normalize(all_predictors()) %>%
  step_pls(all_predictors(), num_comp = tune(), outcome = "Salary")
```

```
lm_spec <- linear_reg() %>%
    set_mode("regression") %>%
    set_engine("lm")

pls_workflow <- workflow() %>%
    add_recipe(pls_recipe) %>%
    add_model(lm_spec)

num_comp_grid <- grid_regular(num_comp(c(1, 20)), levels = 10)

tune_res <- tune_grid(
    pls_workflow,
    resamples = hitters_fold,
    grid = num_comp_grid,
    control(parallel_over = "everything")
)</pre>
```

Warning: The `...` are not used in this function but one or more objects were passed: $^{\prime\prime}$

```
best_threshold <- select_best(tune_res, metric = "rmse")

pls_final <- finalize_workflow(pls_workflow, best_threshold)

pls_final_fit <- fit(pls_final, data = hitters_train)

pls_aug <- augment(pls_final_fit, new_data = hitters_test)

pls_result <- rbind(
    rmse(pls_aug,truth = Salary, estimate = .pred),
    rsq(pls_aug, truth = Salary, estimate = .pred))

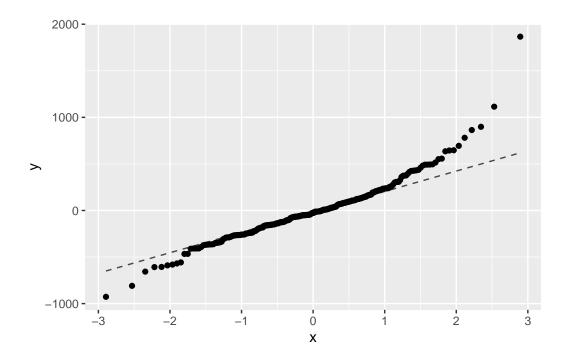
pls_result</pre>
```

Final pls model

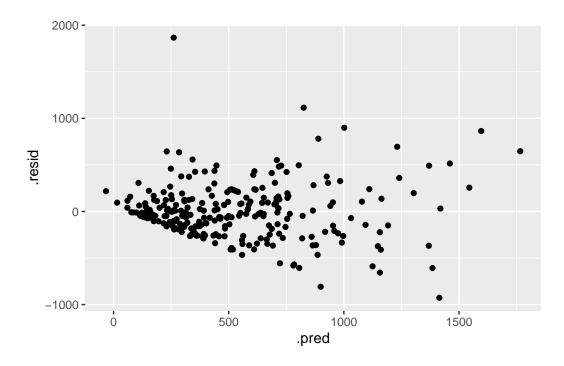
pls_final_fit <- fit(pls_final, data=hitters) tidy(pls_final_fit,conf_int=TRUE)</pre>

A tibble: 12 x 5 estimate std.error statistic p.value term <chr> <dbl> <dbl> <dbl> <dbl> 536. 19.3 27.8 1.24e-78 1 (Intercept) 15.3 6.78e-38 2 PLS01 7.24 111. 4.28 2.69e- 5 3 PLS02 65.6 15.3 4 PLS03 34.7 12.9 2.68 7.76e- 3 2.32 2.10e- 2 5 PLS04 49.5 21.3 6 PLS05 99.6 31.6 3.15 1.81e- 3 7 PLS06 52.5 2.49 1.36e- 2 21.1 8 PLS07 59.5 30.9 1.92 5.57e- 2 9 PLS08 84.7 37.8 2.24 2.61e- 2 10 PLS09 37.6 31.2 1.21 2.29e- 1 57.2 11 PLS10 48.8 1.17 2.42e- 1 12 PLS11 48.6 40.5 1.20 2.31e- 1

```
pls_aug <- augment(pls_final_fit, new_data=hitters) %>% mutate(.resid=Salary-.pred)
pls_aug %>%
    gf_qq(~.resid) %>% gf_qqline()
```



```
pls_aug %>%
  gf_point(.resid~.pred)
```



Blended Ridge and Lasso

```
mixed_recipe <-
  recipe(formula = Salary ~ ., data = hitters_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_zv(all_predictors()) %>%
  step_normalize(all_predictors())
prep(mixed_recipe) #%>% bake(hitters_train)
```

-- Recipe ------

```
-- Inputs
```

Number of variables by role

outcome: 1
predictor: 19

-- Training information

Training data contained 196 data points and no incomplete rows.

- -- Operations
- * Novel factor level assignment for: League, Division, NewLeague | Trained
- * Dummy variables from: League, Division, NewLeague | Trained
- * Zero variance filter removed: League_new and Division_new, ... | Trained
- * Centering and scaling for: AtBat, Hits, HmRun, Runs, RBI, ... | Trained

```
mixed_spec <-
linear_reg(penalty = tune("penalty"), mixture = tune("mixture")) %>%
set_mode("regression") %>%
set_engine("glmnet")

extract_parameter_set_dials(mixed_spec)
```

Collection of 2 parameters for tuning

identifier type object
 penalty penalty nparam[+]
 mixture mixture nparam[+]

```
mixed_workflow <- workflow() %>%
  add_recipe(mixed_recipe) %>%
  add_model(mixed_spec)

mixed_grid <- grid_regular(penalty(range = c(-5, 5)), mixture(),levels = 50)
mixed_grid %>%
  pull(penalty) %>% range()
```

[1] 1e-05 1e+05

```
grid_regular(penalty(), mixture(),levels = 50) %>%
pull(penalty) %>% range()
```

[1] 1e-10 1e+00

```
expand.grid("penalty"=seq(1e-5,1e5,2000), "mixture"=seq(0,1,1/50)) %>%
pull(penalty) %>% range()
```

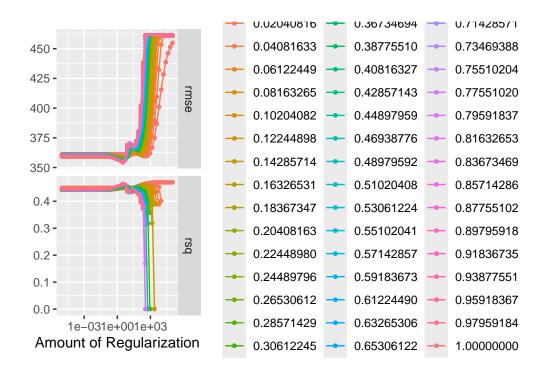
[1] 1.0e-05 9.8e+04

```
system.time(
suppressMessages(
  tune_res <- tune_grid(
    mixed_workflow,
    resamples = hitters_fold,
    grid = mixed_grid,
    control(parallel_over = "everything")
)
)
)</pre>
```

Warning: The `...` are not used in this function but one or more objects were passed: $^{\prime\prime}$

user system elapsed 1.769 0.042 22.779

autoplot(tune_res)



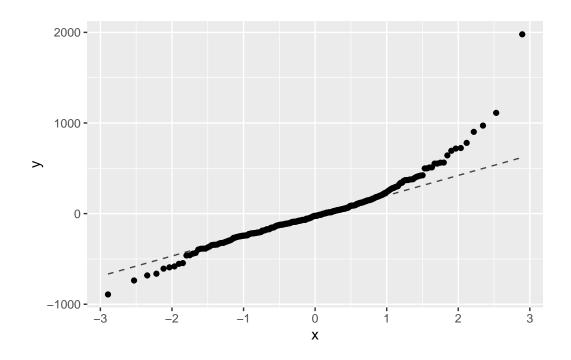
```
best_penalty <- select_best(tune_res, metric = "rmse")
mixed_final <- finalize_workflow(mixed_workflow, best_penalty)
mixed_final_fit <- fit(mixed_final, data = hitters_train)
mixed_aug <- augment(mixed_final_fit, new_data = hitters_test)
mixed_result <- rbind(
   rmse(mixed_aug,truth = Salary, estimate = .pred),
   rsq(mixed_aug, truth = Salary, estimate = .pred))
mixed_result</pre>
```

Final mixed model

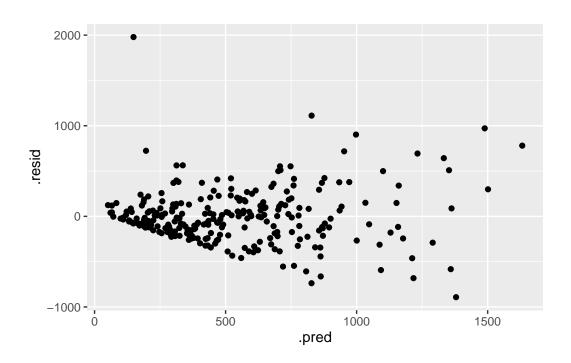
```
mixed_final_fit <- fit(mixed_final, data=hitters)
tidy(mixed_final_fit,conf_int=TRUE)</pre>
```

```
# A tibble: 20 x 3
               estimate penalty
   term
                  <dbl>
                           <dbl>
   <chr>
 1 (Intercept)
                  536.
                            3.24
2 AtBat
                 -205.
                            3.24
3 Hits
                  240.
                            3.24
4 HmRun
                    0
                            3.24
5 Runs
                     0
                            3.24
6 RBI
                    0
                            3.24
7 Walks
                   96.6
                            3.24
                            3.24
8 Years
                  -41.8
9 CAtBat
                    0
                            3.24
10 CHits
                            3.24
                    0
11 CHmRun
                   36.6
                            3.24
12 CRuns
                  201.
                            3.24
13 CRBI
                  129.
                            3.24
14 CWalks
                 -123.
                            3.24
15 PutOuts
                   75.1
                            3.24
16 Assists
                   20.8
                            3.24
17 Errors
                            3.24
                  -11.9
18 League_N
                   15.6
                            3.24
19 Division_W
                  -59.7
                            3.24
20 NewLeague_N
                            3.24
                    0
```

```
mixed_aug <- augment(mixed_final_fit, new_data=hitters) %>% mutate(.resid=Salary-.pred)
mixed_aug %>%
    gf_qq(~.resid) %>% gf_qqline()
```



mixed_aug %>%
 gf_point(.resid~.pred)



Results comparison

6 lm

7 lm

8 pls

rmse

rsq

rsq

298.

0.421

0.447

Compare models fit on the training data and assessed on the test data. The base linear model fit wasn't done above, so do it here.

```
lm_final_fit <- fit(lm_spec, Salary ~ ., data = hitters_train)</pre>
lm_aug <- augment(lm_final_fit, new_data = hitters_test)</pre>
lm_result <- rbind(</pre>
  rmse(lm_aug,truth = Salary, estimate = .pred),
  rsq(lm_aug, truth = Salary, estimate = .pred))
lm_result
# A tibble: 2 x 3
  .metric .estimator .estimate
         <chr>
                         <dbl>
  <chr>
                       298.
1 rmse
          standard
2 rsq
          standard
                         0.421
results <-
  rbind(
    lm_result %>% mutate(method='lm'),
    ridge_result %>% mutate(method='ridge'),
    lasso_result %>% mutate(method='lasso'),
    pca_result %>% mutate(method='pca'),
    pls_result %>% mutate(method='pls'),
    mixed_result %>% mutate(method='mixed')) %>%
  dplyr::select(method,.metric,.estimate) %>%
  arrange(.metric,.estimate)
results
# A tibble: 12 x 3
   method .metric .estimate
   <chr> <chr>
                    <dbl>
 1 pca
          rmse
                    271.
                    277.
 2 ridge rmse
 3 lasso rmse
                    277.
 4 mixed rmse
                    277.
 5 pls rmse
                    291.
```

```
9 mixed rsq 0.491
10 lasso rsq 0.493
11 ridge rsq 0.515
12 pca rsq 0.521
```

```
results %>%
  filter(.metric=='rmse') %>%
  gf_point(.estimate~method) %>%
  gf_labs(y='rmse',title="Cross-validated rmse by method")
```

Cross-validated rmse by method

