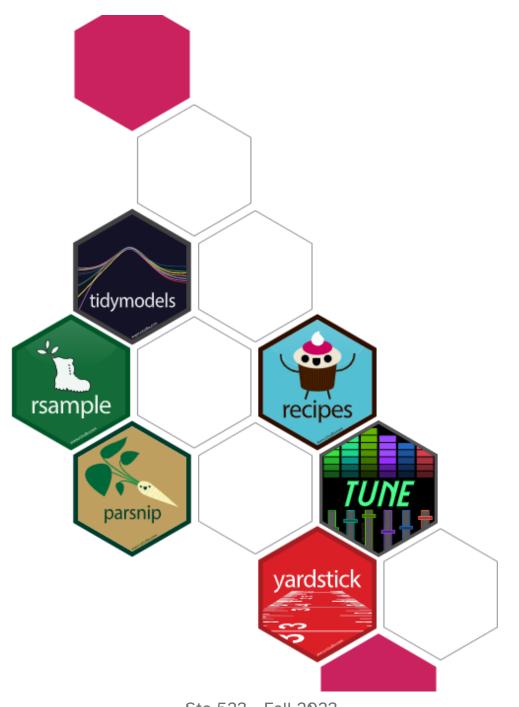
Tidymodels

Lecture 22

Dr. Colin Rundel



Sta 523 - Fall 2023

Tidymodels

```
1 library(tidymodels)
- Attaching packages -
                                                                     - tidymodels 1.1.1 -
✓ broom
              1.0.5
                         ✓ rsample
                                       1.2.0

✓ dials

              1.2.0
                                       1.1.2

✓ tune

✓ infer
              1.0.5
                      ✓ workflows
                                       1.1.3

✓ modeldata

              1.2.0
                        ✓ workflowsets 1.0.1
              1.1.1
✓ parsnip
                        ✓ yardstick
                                      1.2.0
✓ recipes
              1.0.8
- Conflicts -
                                                               - tidymodels conflicts() --
* scales::discard()
                     masks purrr::discard()
                     masks stats::filter()
* dplyr::filter()
* recipes::fixed()
                     masks stringr::fixed()
* dplyr::lag()
                     masks stats::lag()
* rsample::populate() masks Rcpp::populate()
* yardstick::spec()
                     masks readr::spec()
* recipes::step()
                     masks stats::step()
• Search for functions across packages at https://www.tidymodels.org/find/
```

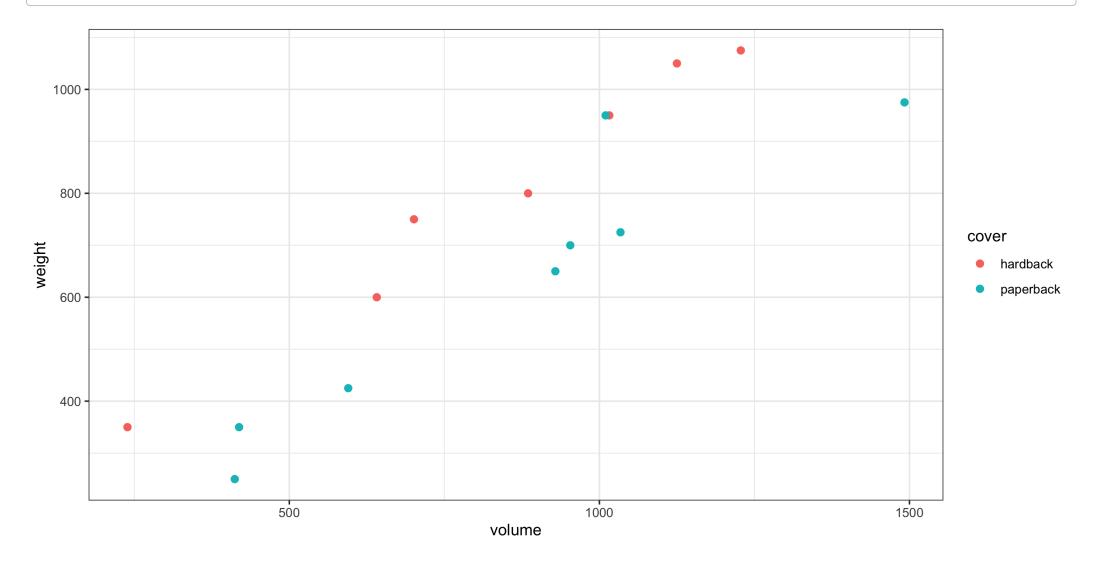
Book data

```
(books = DAAG::allbacks |>
     as tibble() |>
     select(-area) |>
 3
     mutate(
 5
       cover = forcats::fct recode(
 6
         cover,
          "hardback" = "hb",
         "paperback" = "pb"
 9
10
11
```

```
# A tibble: 15 \times 3
  volume weight cover
   <dbl> <dbl> <fct>
     885
            800 hardback
    1016 950 hardback
    1125 1050 hardback
            350 hardback
     239
     701 750 hardback
     641
            600 hardback
    1228
           1075 hardback
     412
            250 paperback
            700 paperback
     953
            650 nanorhadk
1 ^
     020
```

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```
ggplot(books, aes(x=volume, y=weight, color = cover)) +
geom_point(size=2)
```



Building a tidymodel

```
1 linear_reg()
Linear Regression Model Specification (regression)
Computational engine: lm
```

Building a tidymodel

```
1 linear_reg() |>
2 set_engine("lm")
```

Linear Regression Model Specification (regression)

Computational engine: lm

Building a tidymodel

```
1 linear reg() |>
      set engine("lm") |>
      fit(weight ~ volume * cover, data = books)
parsnip model object
Call:
stats::lm(formula = weight ~ volume * cover, data =
data)
Coefficients:
          (Intercept)
                                      volume
            161.58654
                                     0.76159
       coverpaperback volume:coverpaperback
           -120.21407
                                    -0.07573
```



Tidy model objects

```
1 lm_b = lm(weight ~ volume * cover, data = books)
```

```
1 lm_tm = linear_reg() |>
2 set_engine("lm") |>
3 fit(weight ~ volume * cover, data = books)
```

```
1 summary(lm_b)
```

```
1 summary(lm_tm)
```


	Length	Class	Mode
lvl	0	-none-	NULL
spec	7	linear_reg	list
fit	13	lm	list
preproc	1	-none-	list
elapsed	1	-none-	list
censor_probs	0	-none-	list

Coefficients:

```
Estimate Std. Error t value
(Intercept)
                      161.58654
                                  86.51918
                                           1.868
volume
                                           7.837
                        0.76159
                                   0.09718
coverpaperback
                     -120.21407 115.65899 -1.039
volume:coverpaperback
                      -0.07573
                                0.12802 - 0.592
                     Pr(>|t|)
(Intercept)
                       0.0887 .
                     7.94e-06 ***
volume
```

1 broom::tidy(lm tm)

```
# A tibble: 4 \times 5
                      estimate std.error statistic p.value
 term
 <chr>
                        <dbl>
                                 <dbl>
                                          <dbl>
                                                   <dbl>
1 (Intercept)
                      162.
                             86.5
                                          1.87 0.0887
                       0.762
                                0.0972 7.84 0.00000794
2 volume
                    -120. 116.
3 coverpaperback
                                         -1.04 0.321
4 volume:coverpaperback -0.0757
                                0.128
                                         -0.592 0.566
```

Tidy model statistics

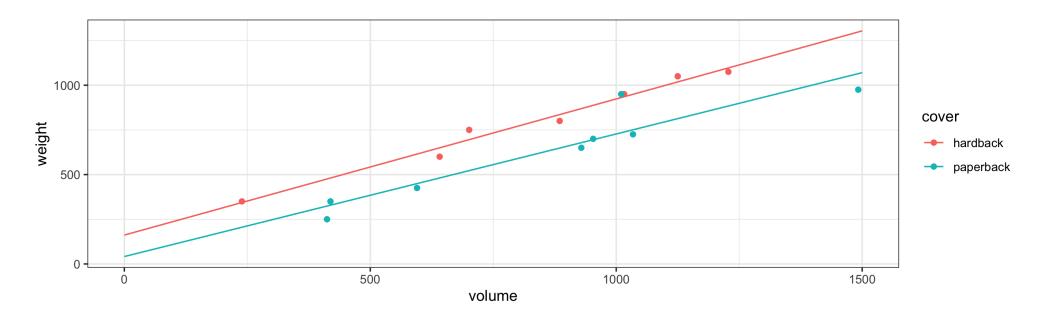
```
1 broom::glance(lm b)
# A tibble: 1 × 12
  r.squared adj.r.squared sigma statistic p.value
                                                     df logLik
                                                                AIC
                                                                      BIC deviance
                                  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
     <dbl>
                   <dbl> <dbl>
                                                                             <dbl>
                                   48.5 1.24e-6
     0.930
                   0.911 80.4
                                                      3 -84.8 180. 183.
                                                                            71118.
1
# i 2 more variables: df.residual <int>, nobs <int>
  1 broom::glance(lm tm)
# A tibble: 1 \times 12
 r.squared adj.r.squared sigma statistic p.value
                                                     df logLik
                                                                AIC
                                                                      BIC deviance
                   <dbl> <dbl>
                                  <dbl>
                                            <dbl> <dbl> <dbl> <dbl> <dbl> <
                                                                             <dbl>
     <dbl>
                   0.911 80.4
                                   48.5
                                         1.24e-6
     0.930
                                                      3 -84.8 180. 183.
                                                                            71118.
# i 2 more variables: df.residual <int>, nobs <int>
```

Tidy model prediction

```
1 broom::augment(lm tm, new data = books)
# A tibble: 15 \times 5
   .pred .resid volume weight cover
   <dbl> <dbl> <dbl> <fct>
   836. -35.6
                   885
                          800 hardback
   935. 14.6
                 1016 950 hardback
                        1050 hardback
 3 1018.
         31.6
                1125
 4 344. 6.39
                   239
                         350 hardback
   695. 54.5
                         750 hardback
                   701
   650. -49.8
               641
                         600 hardback
 7 1097. -21.8
                 1228
                         1075 hardback
   324. -73.9
                   412
                          250 paperback
    695.
           5.00
                   953
                          700 paperback
    670
          20 5
                   0.00
1 ^
                          GEO manarhadis
```

Putting it together

```
1 lm_tm |>
2 augment(
3    new_data = tidyr::expand_grid(
4    volume = seq(0, 1500, by=5),
5    cover = c("hardback", "paperback") |> as.factor()
6    )
7    ) |>
8    rename(weight = .pred) |>
9    ggplot(aes(x = volume, y = weight, color = cover, group = cover)) +
10    geom_line() +
11    geom_point(data = books)
```





Why do we care?

```
1 (bayes_tm = linear_reg() |>
2   set_engine(
3    "stan",
4    prior_intercept = rstanarm::student_t(df = 1)
5    prior = rstanarm::student_t(df = 1),
6    seed = 1234
7   )
8 )
```

Linear Regression Model Specification (regression)
Engine-Specific Arguments:
 prior_intercept = rstanarm::student_t(df = 1)
 prior = rstanarm::student_t(df = 1)
 seed = 1234
Computational engine: stan

Fitting with rstanarm

```
1 (bayes tm = bayes tm |>
      fit(weight ~ volume * cover, data = books)
  3)
parsnip model object
stan_glm
 family:
              gaussian [identity]
 formula:
              weight ~ volume * cover
 observations: 15
 predictors:
                      Median MAD SD
(Intercept)
                      95.6
                             61.8
volume
                       0.8
                             0.1
coverpaperback
                      -0.2
                             3.7
volume:coverpaperback -0.2
                              0.1
Auxiliary parameter(s):
      Median MAD SD
sigma 85.5 17.8
```

What was actually run?

```
linear_reg() |>
set_engine(
    "stan",
    prior_intercept = rstanarm::student_t(df = 1),
    prior = rstanarm::student_t(df = 1),
    seed = 1234
    ) |>
    translate()
Linear Regression Model Specification (regression)
```

```
Engine-Specific Arguments:
    prior_intercept = rstanarm::student_t(df = 1)
    prior = rstanarm::student_t(df = 1)
    seed = 1234

Computational engine: stan

Model fit template:
    rstanarm::stan_glm(formula = missing_arg(), data = missing_arg(),
        weights = missing_arg(), prior_intercept = rstanarm::student_t(df = 1),
        prior = rstanarm::student_t(df = 1), seed = 1234, family = stats::gaussian,
        refresh = 0)
```

Back to broom

```
1 broom::tidy(bayes_tm)
```

Error in warn_on_stanreg(x): The supplied model object seems to be outputted from the rstanarm package. Tidiers for mixed model output now live in the broom.mixed package.

```
1 broom.mixed::tidy(bayes tm)
# A tibble: 4 \times 3
                       estimate std.error
  term
  <chr>
                          <dbl>
                                    <dbl>
                         95.6
                                  61.8
1 (Intercept)
2 volume
                         0.831 0.0743
3 coverpaperback
                         -0.238 3.73
4 volume:coverpaperback
                         -0.198
                                   0.0506
  1 broom.mixed::glance(bayes tm)
# A tibble: 1 \times 4
  algorithm pss nobs sigma
  <chr> <dbl> <int> <dbl>
1 sampling
            4000
                    15 85.5
```

Augment

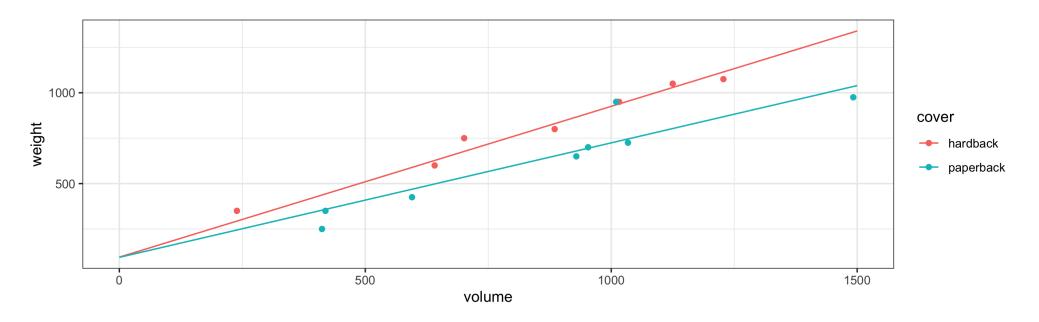
```
augment(bayes tm, new data=books)
# A tibble: 15 \times 5
          .resid volume weight cover
   .pred
   <dbl>
         <dbl> <dbl> <fct>
   830.
                           800 hardback
          -29.8
                    885
                           950 hardback
   939.
           11.5
                   1016
 3 1029.
           21.0
                   1125
                          1050 hardback
    294.
           56.4
                    239
                           350 hardback
                           750 hardback
 5
    677.
           73.0
                    701
    627.
                           600 hardback
          -27.2
                    641
 7 1114. -39.5
                          1075 hardback
                   1228
   354. -104.
 8
                    412
                           250 paperback
    695.
            5.39
                    953
                           700 paperback
    679. -29.5
                    929
                           650 paperback
10
11 1034. -59.3
                   1492
                           975 paperback
```

Predictions

```
bayes_tm |>
augment(
new_data = tidyr::expand_grid(
volume = seq(0, 1500, by=5),
cover = c("hardback", "paperback") |> as.factor()

new_data = tidyr::expand_grid(
volume = seq(0, 1500, by=5),
sequence = c("hardback", "paperback") |> as.factor()

rename(weight = .pred) |>
ggplot(aes(x = volume, y = weight, color = cover, group = cover)) +
geom_line() +
geom_point(data = books)
```





Performance

```
1 lm_tm |>
2 augment(new_data = books) |>
3 yardstick::rmse(weight, .pred)
```

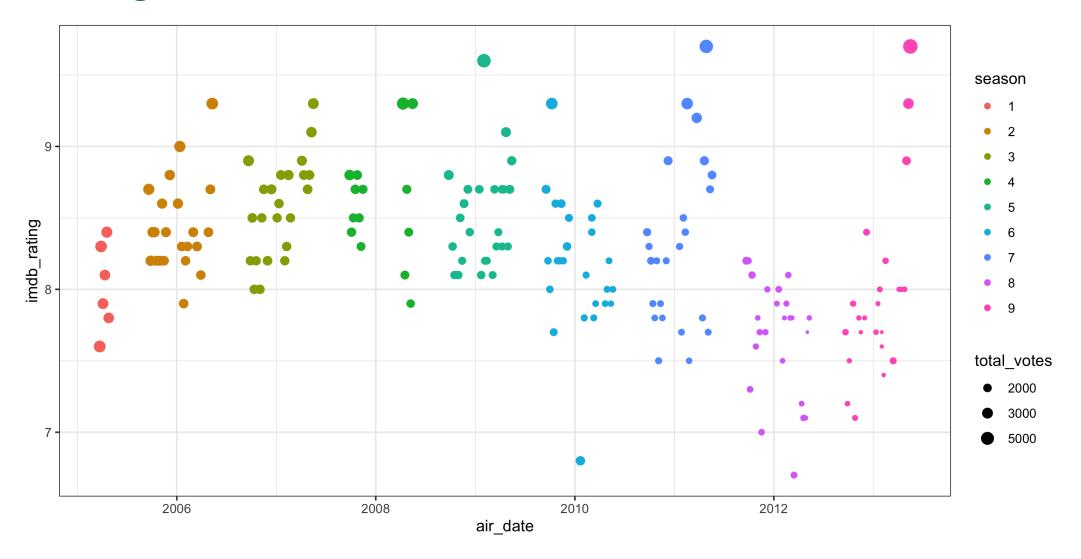
```
1 bayes_tm |>
2 augment(new_data = books) |>
3 yardstick::rmse(weight, .pred)
```

Cross validation and Feature engineering

The Office & IMDB

```
1 (office ratings = read csv("data/office ratings.csv"))
# A tibble: 188 × 6
   season episode title
                                     imdb rating total votes air date
    <dbl>
            <dbl> <chr>
                                           <dbl>
                                                        <dbl> <date>
                1 Pilot
                                             7.6
                                                         3706 2005-03-24
        1
 1
                                                         3566 2005-03-29
                2 Diversity Day
                                             8.3
 2
                3 Health Care
                                             7.9
                                                         2983 2005-04-05
                4 The Alliance
                                             8.1
                                                         2886 2005-04-12
 4
 5
                5 Basketball
                                             8.4
                                                         3179 2005-04-19
                6 Hot Girl
                                             7.8
                                                         2852 2005-04-26
 6
                1 The Dundies
                                             8.7
                                                         3213 2005-09-20
                2 Sexual Harassment
                                                         2736 2005-09-27
 8
                                             8.2
                                                         2742 2005-10-04
                3 Office Olympics
                                             8.4
10
                4 The Fire
                                             8.4
                                                         2713 2005-10-11
# i 178 more rows
```

Rating vs Air Date





Test-train split

```
1 set.seed(123)
  2 (office split = initial split(office ratings, prop = 0.8))
<Training/Testing/Total>
<150/38/188>
  1 (office train = training(office split))
                                                             1 (office test = testing(office split))
# A tibble: 150 \times 6
                                                           # A tibble: 38 \times 6
                                                                                        imdb rating total votes
   season episode title
                           imdb rating total votes
                                                               season episode title
          <dbl> <chr>
                                   <db1>
                                                <dbl>
                                                               <dbl>
                                                                        <dbl> <chr>
                                                                                               <dbl>
                                                                                                            <dbl>
    <dbl>
                                      7.8
                                                                             2 Diversi...
 1
                18 Last Da...
                                                 1429
                                                                    1
                                                                                                 8.3
                                                                                                             3566
                                                                            4 The Fire
 2
               14 Vandali...
                                      7.6
                                                 1402
                                                                                                 8.4
                                                                                                             2713
                                                                             9 E-Mail ...
 3
                 8 Perform...
                                     8.2
                                                 2416
                                                                                                 8.4
                                                                                                             2527
                 5 Here Co...
                                      7.1
                                                 1515
                                                                            12 The Inj...
                                                                                                             3282
 4
                22 Beach G...
                                                 2783
                                                                            22 Casino ...
                                                                                                 9.3
                                                                                                             3644
 5
                                      9.1
 6
                 1 Nepotism
                                     8.4
                                                 1897
                                                                             5 Initiat...
                                                                                                 8.2
                                                                                                             2254
                15 Phyllis...
                                      8.3
                                                  2283
                                                                            16 Busines...
                                                                                                 8.8
                                                                                                             2622
                21 Livin' ...
 8
                                      8.9
                                                 2041
                                                                            17 Cocktai...
                                                                                                 8.5
                                                                                                             2264
 9
                18 Promos
                                                 1445
                                                                             6 Branch ...
                                                                                                 8.5
                                                                                                             2185
        9
                                      8
                                                                             7 Survivo...
10
                12 Pool Pa...
                                                 1612
                                                           10
                                                                                                 8.3
                                                                                                             2110
    140 more rows
                                                           # i 28 more rows
                                                           # i 1 more variable: air date <date>
# i 1 more variable: air date <date>
```

Feature engineering with dplyr

i 140 more rows

```
1 options(width=100)
2 options(tibble.width=100)
```

```
office_train |>
mutate(
season = as_factor(season),
month = lubridate::month(air_date),
wday = lubridate::wday(air_date),
top10_votes = as.integer(total_votes > quantile(total_votes, 0.9))
)
```

```
# A tibble: 150 \times 9
   season episode title
                                      imdb rating total votes air date month wday top10 votes
  <fct>
            <dbl> <chr>
                                            <dbl>
                                                        <dbl> <date>
                                                                         <dbl> <dbl>
                                                                                            <int>
 1 8
              18 Last Day in Florida
                                              7.8
                                                         1429 2012-03-08
                                                                             3
                                                                                   5
              14 Vandalism
                                              7.6
                                                         1402 2013-01-31
 2 9
               8 Performance Review
                                                         2416 2005-11-15
 3 2
                                              8.2
                                                                            11
 4 9
               5 Here Comes Treble
                                             7.1
                                                         1515 2012-10-25
                                                                            1.0
 5 3
                                                         2783 2007-05-10
               22 Beach Games
                                              9.1
 6 7
                                              8.4
                                                         1897 2010-09-23
               1 Nepotism
              15 Phyllis' Wedding
                                              8.3
                                                         2283 2007-02-08
 7 3
 8 9
               21 Livin' the Dream
                                              8.9
                                                         2041 2013-05-02
 9 9
              18 Promos
                                                         1445 2013-04-04
10 8
              12 Pool Party
                                                         1612 2012-01-19
```



Better living through recipes

```
1 r = recipe(imdb_rating ~ ., data = office_train)
```

Recipe roles

1 summary(r)

```
1 r = recipe(
2  imdb_rating ~ ., data = office_train
3 ) |>
4  update_role(title, new_role = "ID")
```

Adding features (month & day of week)

```
1 r = recipe(
2  imdb_rating ~ ., data = office_train
3 ) |>
4  update_role(title, new_role = "ID") |>
5  step_date(air_date, features = c("dow", "month")
```

```
1 summary(r)
# A tibble: 6 \times 4
 variable type
                role
                       source
 <chr> <chr>
                <chr>
                       <chr>
3 title
       <chr [3]> ID
                       original
4 total votes <chr [2]> predictor original
5 air date <chr [1]> predictor original
6 imdb rating <chr [2]> outcome
                       original
```

Adding Holidays

```
1 r = recipe(
     imdb_rating ~ ., data = office train
   ) |>
 3
     update role(title, new role = "ID") |>
 4
     step date(air date, features = c("dow", "month")) |>
     step holiday(
 6
     air date,
 7
       holidays = c("USThanksgivingDay", "USChristmasDay", "USNewYearsDay", "USIndependenceDay"),
 8
      keep original cols = FALSE
 9
10
```

Seasons as factors

```
1 r = recipe(
     imdb rating ~ ., data = office train
   ) |>
 3
     update role(title, new role = "ID") |>
 4
     step date(air date, features = c("dow", "month")) |>
     step_holiday(
 6
     air date,
 7
       holidays = c("USThanksgivingDay", "USChristmasDay", "USNewYearsDay", "USIndependenceDay"),
 8
      keep original cols = FALSE
 9
10
     step num2factor(season, levels = as.character(1:9))
11
```

```
1 summary(r)
```

Dummy coding

```
1 r = recipe(
     imdb rating ~ ., data = office train
     |>
 3
     update role(title, new role = "ID") |>
 4
     step date(air date, features = c("dow", "month")) |>
     step_holiday(
 6
      air date,
       holidays = c("USThanksgivingDay", "USChristmasDay", "USNewYearsDay", "USIndependenceDay"),
 8
      keep original cols = FALSE
 9
1.0
     step num2factor(season, levels = as.character(1:9)) |>
11
12
     step dummy(all nominal predictors())
```

```
1 summary(r)
```

top10_votes

```
1 r = recipe(
     imdb rating ~ ., data = office train
   ) |>
 3
     update role(title, new role = "ID") |>
 4
     step date(air date, features = c("dow", "month")) |>
 5
     step_holiday(
 6
      air date,
 7
       holidays = c("USThanksgivingDay", "USChristmasDay", "USNewYearsDay", "USIndependenceDay"),
 8
      keep original cols = FALSE
 9
10
     step num2factor(season, levels = as.character(1:9)) |>
11
12
     step dummy(all nominal predictors()) |>
     step percentile(total votes) |>
13
     step mutate(top10 = as.integer(total votes >= 0.9)) |>
14
15
     step rm(total votes)
```

```
1 summary(r)
```

Preparing a recipe

```
1 prep(r)
- Recipe -
- Inputs
Number of variables by role
outcome:
predictor: 4
ID:
          1
- Training information
Training data contained 150 data points and no incomplete rows.
- Operations
• Date features from: air date | Trained
• Holiday features from: air date | Trained
• Factor variables from: season | Trained
• Dummy variables from: season, air date dow, air date month | Trained
• Percentile transformation on: total votes | Trained
• Variable mutation for: ~as.integer(total votes >= 0.9) | Trained
• Variables removed: total votes | Trained
```

Baking a recipe

```
1 prep(r) |>
      bake(new data = office train)
# A tibble: 150 × 33
   episode title
                    imdb_rating air_date_USThanksgiv...¹ air_date_USChristmas...² air_date_USNewYearsDay
     <dbl> <fct>
                           <dbl>
                                                   <int>
                                                                           <int>
                                                                                                   <int>
 1
        18 Last Da...
                             7.8
                                                       0
                                                                                                       0
        14 Vandali...
 2
                            7.6
                                                       0
                                                                                                       0
        8 Perform...
                           8.2
 3
                                                                                                       0
        5 Here Co...
                            7.1
        22 Beach G...
                           9.1
 5
        1 Nepotism
                           8.4
 6
        15 Phyllis...
 7
                            8.3
        21 Livin' ...
 8
                             8.9
        18 Promos
 9
                             8
10
        12 Pool Pa...
# i 140 more rows
# i abbreviated names: ¹air date USThanksgivingDay, ²air date USChristmasDay
# i 27 more variables: air date USIndependenceDay <int>, season X2 <dbl>, season X3 <dbl>,
    season X4 <dbl>, season X5 <dbl>, season X6 <dbl>, season X7 <dbl>, season X8 <dbl>,
```

Informative features?

```
1 prep(r) |>
2  bake(new_data = office_train) |>
3  map_int(~ length(unique(.x)))
```

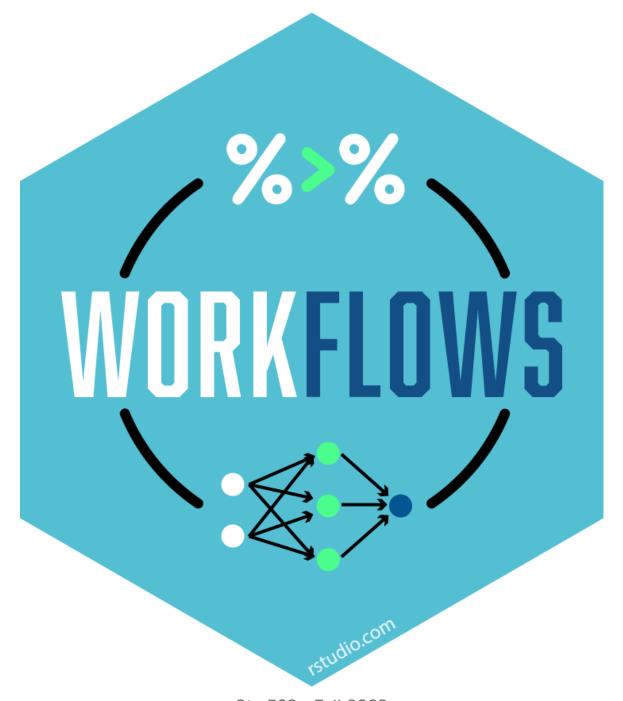
```
episode
                                                 title
                                                                       imdb_rating
                        26
                                                   150
                                                                                26
air_date_USThanksgivingDay
                              air_date_USChristmasDay
                                                           air_date_USNewYearsDay
                         1
air date USIndependenceDay
                                             season X2
                                                                         season X3
                                             season X5
                 season X4
                                                                         season X6
                 season X7
                                             season X8
                                                                         season X9
          air date dow Mon
                                     air date dow Tue
                                                                 air date dow Wed
          air date dow Thu
                                      air date dow Fri
                                                                 air date dow Sat
        air date month Feb
                                   air date month Mar
                                                               air date month Apr
        air date month Mav
                                   air date month Jun
                                                               air date month Jul
```

Removing zero variance predictors

```
1 r = recipe(
       imdb rating ~ ., data = office train
 3
     ) |>
     update role(title, new role = "ID") |>
 4
     step date(air date, features = c("dow", "month")) |>
     step holiday(
 6
     air date,
       holidays = c("USThanksgivingDay", "USChristmasDay", "USNewYearsDay", "USIndependenceDay"),
 8
      keep original cols = FALSE
9
10
     step num2factor(season, levels = as.character(1:9)) |>
11
12
     step dummy(all nominal predictors()) |>
     step percentile(total votes) |>
13
     step mutate(top10 = as.integer(total votes >= 0.9)) |>
14
     step rm(total votes) |>
15
     step zv(all predictors())
16
```

```
1 prep(r) |>
       bake(new data = office train)
  2
# A tibble: 150 × 22
                    imdb_rating season_X2 season_X3 season_X4 season_X5 season_X6 season_X7 season_X8
   episode title
     <dbl> <fct>
                           <dbl>
                                      <dbl>
                                                 <dbl>
                                                            <dbl>
                                                                                  <dbl>
                                                                       <dbl>
                                                                                             <dbl>
                                                                                                        <dbl>
        18 Last D...
                             7.8
 1
                                           0
                                                      0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             1
        14 Vandal...
                             7.6
 2
                                           0
                                                      0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             0
        8 Perfor...
                             8.2
 3
                                           1
                                                      0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             0
 4
         5 Here C...
                             7.1
                                           0
                                                      0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             0
        22 Beach ...
                             9.1
 5
                                                                 0
                                                                            0
                                                                                                  0
                                                                                                             0
         1 Nepoti...
                             8.4
 6
                                                                 0
                                                                            0
                                                                                                  1
                                                                                                             0
        15 Phylli...
 7
                             8.3
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             0
        21 Livin'...
                             8.9
 8
                                           0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             0
        18 Promos
 9
                                           0
                                                                 0
                                                                            0
                                                                                                  0
                                                                                                             0
10
        12 Pool P...
                                           0
                                                                 0
                                                                            0
                                                                                       0
                                                                                                  0
                                                                                                             1
# i 140 more rows
# i 12 more variables: season X9 <dbl>, air date dow Tue <dbl>, air date dow Thu <dbl>,
    air date month Feb <dbl>, air date month Mar <dbl>, air date month Apr <dbl>,
#
```

air date month Mav <dbl>. air date month Sep <dbl>. air date month Oct <dbl>. #



Really putting it all together

```
1 (office_work = workflow() |>
2   add_recipe(r) |>
3   add_model(
4   linear_reg() |>
5   set_engine("lm")
6  )
7 )
```

```
= Workflow =====
Preprocessor: Recipe
Model: linear reg()
- Preprocessor -
8 Recipe Steps
• step date()
• step holiday()
• step num2factor()
• step dummy()
• step percentile()
• step mutate()
• step rm()
• step zv()
- Model -
```

Workflow fit

```
1 (office fit = office work |>
     fit(data = office_train))
= Workflow [trained] =-----
Preprocessor: Recipe
Model: linear_reg()
- Preprocessor ----
8 Recipe Steps
• step_date()
• step holiday()
• step_num2factor()
• step dummy()
• step percentile()
• step mutate()
• step rm()
• step zv()
-- Model -----
```

Performance

```
1 office_fit |>
2   augment(office_train) |>
3   rmse(imdb_rating, .pred)
```

```
1 office_fit |>
2   augment(office_test) |>
3   rmse(imdb_rating, .pred)
```

k-fold cross validation

	training					testing
fold 1	validate	train	train	train	train	
fold 2	train	validate	train	train	train	
fold 3	train	train	validate	train	train	
fold 4	train	train	train	validate	train	
fold 5	train	train	train	train	validate	

Creating folds

```
1 (office_fit_folds = office_work |>
2  fit_resamples(folds)
3 )
```

```
# Resampling results
# 5-fold cross-validation
# A tibble: 5 \times 4
  splits
                        id
                                .metrics
  st>
                        <chr> <list>
1 <split [120/30]> Fold1 <tibble [2 × 4]>
2 \langle \text{split} [120/30] \rangle Fold2 \langle \text{tibble} [2 \times 4] \rangle
3 < \text{split} [120/30] > \text{Fold} 3 < \text{tibble} [2 \times 4] >
4 <split [120/30]> Fold4 <tibble [2 \times 4]>
5 <split [120/30]> Fold5 <tibble [2 \times 4]>
   .notes
  t>
1 <tibble [0 × 3]>
2 <tibble [1 × 3]>
3 < tibble [0 \times 3] >
4 < tibble [0 \times 3] >
5 < tibble [0 \times 3] >
```

Fold performance

```
1 tune::collect metrics(office fit folds)
# A tibble: 2 \times 6
  .metric .estimator mean
                               n std err .config
          <chr>
                     <dbl> <int> <dbl> <chr>
  <chr>
          standard 0.420
                               5 0.0182 Preprocessor1 Model1
1 rmse
                               5 0.0597 Preprocessor1 Model1
2 rsq
          standard
                   0.429
  1 tune::collect metrics(office fit folds, summarize = FALSE) |>
      filter(.metric == "rmse")
# A tibble: 5 \times 5
        .metric .estimator .estimate .config
  id
  <chr> <chr>
                <chr>
                               <dbl> <chr>
1 Fold1 rmse
                standard
                               0.467 Preprocessor1 Model1
2 Fold2 rmse
                standard
                               0.403 Preprocessor1 Model1
3 Fold3 rmse
                standard
                               0.405 Preprocessor1 Model1
                               0.454 Preprocessor1 Model1
4 Fold4 rmse
                standard
```

0.368 Preprocessor1 Model1

standard

5 Fold5 rmse