

Importance measures

```
set.seed(42)

library(rcompanion) # KW effect size calculation
library(rstatix) # Wilcox effect size calculation

##
## Attaching package: 'rstatix'
## The following object is masked from 'package:stats':
##
##   filter
library(igraph)

##
## Attaching package: 'igraph'
## The following objects are masked from 'package:stats':
##
##   decompose, spectrum
## The following object is masked from 'package:base':
##
##   union
library(corrplot)

## corrplot 0.95 loaded
library(QuantPsyc) # for the multivariate normality test

## Loading required package: boot
## Loading required package: dplyr
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:igraph':
##
##   as_data_frame, groups, union
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
## Loading required package: purrr
##
## Attaching package: 'purrr'
```

```

## The following objects are masked from 'package:igraph':
##
##   compose, simplify
## Loading required package: MASS
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##   select
## The following object is masked from 'package:rstatix':
##
##   select
##
## Attaching package: 'QuantPsyc'
## The following object is masked from 'package:base':
##
##   norm
library(dunn.test)
library(nFactors) # for the scree plot

## Loading required package: lattice
##
## Attaching package: 'lattice'
## The following object is masked from 'package:boot':
##
##   melanoma
##
## Attaching package: 'nFactors'
## The following object is masked from 'package:lattice':
##
##   parallel
library(psych) # for PA FA

##
## Attaching package: 'psych'
## The following object is masked from 'package:boot':
##
##   logit
## The following object is masked from 'package:rcompanion':
##
##   phi
library(caret) # highly correlated features removal

## Loading required package: ggplot2
##
## Attaching package: 'ggplot2'

```

```
## The following objects are masked from 'package:psych':
##
##   %+%, alpha
##
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
##   lift
library(tidymodels)

## -- Attaching packages ----- tidymodels 1.2.0 --
## v broom      1.0.5    v tibble      3.2.1
## v dials      1.3.0    v tidyr      1.3.1
## v infer      1.0.7    v tune       1.2.1
## v modeldata  1.4.0    v workflows  1.1.4
## v parsnip    1.2.1    v workflowsets 1.1.0
## v recipes    1.1.0    v yardstick  1.3.2
## v rsample    1.2.1

## -- Conflicts ----- tidymodels_conflicts() --
## x ggplot2::%+%( )      masks psych::%+%( )
## x yardstick::accuracy() masks rcompanion::accuracy()
## x scales::alpha( )     masks ggplot2::alpha( ), psych::alpha( )
## x tibble::as_data_frame( ) masks dplyr::as_data_frame( ), igraph::as_data_frame( )
## x infer::chisq_test( )  masks rstatix::chisq_test( )
## x purrr::compose( )     masks igraph::compose( )
## x tidyr::crossing( )    masks igraph::crossing( )
## x dials::degree( )      masks igraph::degree( )
## x scales::discard( )    masks purrr::discard( )
## x dplyr::filter( )      masks rstatix::filter( ), stats::filter( )
## x dials::get_n( )       masks rstatix::get_n( )
## x dplyr::lag( )         masks stats::lag( )
## x caret::lift( )        masks purrr::lift( )
## x dials::neighbors( )   masks igraph::neighbors( )
## x yardstick::precision( ) masks caret::precision( )
## x infer::prop_test( )   masks rstatix::prop_test( )
## x yardstick::recall( )  masks caret::recall( )
## x MASS::select( )       masks dplyr::select( ), rstatix::select( )
## x yardstick::sensitivity( ) masks caret::sensitivity( )
## x purrr::simplify( )    masks igraph::simplify( )
## x yardstick::specificity( ) masks caret::specificity( )
## x recipes::step( )      masks stats::step( )
## x infer::t_test( )      masks rstatix::t_test( )
## * Dig deeper into tidy modeling with R at https://www.tmw.org
library(vip)

##
## Attaching package: 'vip'
## The following object is masked from 'package:utils':
##
##   vi
```

```

library(tidyverse)

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats 1.0.0      v readr 2.1.5
## v lubridate 1.9.3    v stringr 1.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x lubridate::%--%()      masks igraph::%--%()
## x ggplot2::%+%()        masks psych::%+%()
## x scales::alpha()       masks ggplot2::alpha(), psych::alpha()
## x tibble::as_data_frame() masks dplyr::as_data_frame(), igraph::as_data_frame()
## x readr::col_factor()   masks scales::col_factor()
## x purrr::compose()      masks igraph::compose()
## x tidyr::crossing()     masks igraph::crossing()
## x scales::discard()    masks purrr::discard()
## x dplyr::filter()       masks rstatix::filter(), stats::filter()
## x stringr::fixed()     masks recipes::fixed()
## x dplyr::lag()          masks stats::lag()
## x caret::lift()        masks purrr::lift()
## x MASS::select()       masks dplyr::select(), rstatix::select()
## x purrr::simplify()    masks igraph::simplify()
## x readr::spec()        masks yardstick::spec()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors.

library(paletteer) # color palettes

library(conflicted) # to resolve QuantPsyc x dplyr conflicts
conflict_prefer("select", "dplyr")

## [conflicted] Will prefer dplyr::select over any other package.

conflict_prefer("filter", "dplyr")

## [conflicted] Will prefer dplyr::filter over any other package.

```

Load and tidy data

```

pretty_names <- read_csv("../feat_name_mapping.csv")

## Rows: 85 Columns: 2
## -- Column specification -----
## Delimiter: ","
## chr (2): name_orig, name_pretty
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

prettify_feat_name <- function(x) {
  name <- pull(pretty_names %>%
    filter(name_orig == x), name_pretty)
  if (length(name) == 1) {
    return(name)
  } else {
    return(x)
  }
}

```

```

}

prettify_feat_name_vector <- function(x) {
  map(
    x,
    prettify_feat_name
  ) %>% unlist()
}

data <- read_csv("../measurements/measurements.csv")

## Rows: 753 Columns: 108
## -- Column specification -----
## Delimiter: ","
## chr (20): fpath, KUK_ID, FileName, FileFormat, FolderPath, subcorpus, Source...
## dbl (85): RuleAbstractNouns, RuleAmbiguousRegards, RuleAnaphoricReferences, ...
## lgl (3): ClarityPursuit, SyllogismBased, Bindingness
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

.firstnonmetacolumn <- 17

data_no_nas <- data %>%
  select(!c(
    fpath,
    # KUK_ID,
    # FileName,
    FolderPath,
    # subcorpus,
    DocumentTitle,
    ClarityPursuit,
    Readability,
    SyllogismBased,
    SourceDB
  )) %>%
  # replace -1s in variation coefficients with NAs
  mutate(across(c(
    `RuleDoubleAdpos.max_allowable_distance.v`,
    `RuleTooManyNegations.max_negation_frac.v`,
    `RuleTooManyNegations.max_allowable_negations.v`,
    `RuleTooManyNominalConstructions.max_noun_frac.v`,
    `RuleTooManyNominalConstructions.max_allowable_nouns.v`,
    `RuleCaseRepetition.max_repetition_count.v`,
    `RuleCaseRepetition.max_repetition_frac.v`,
    `RulePredSubjDistance.max_distance.v`,
    `RulePredObjDistance.max_distance.v`,
    `RuleInfVerbDistance.max_distance.v`,
    `RuleMultiPartVerbs.max_distance.v`,
    `RuleLongSentences.max_length.v`,
    `RulePredAtClauseBeginning.max_order.v`,
    `mattr.v`,
    `maentropy.v`
  )))

```

```

), ~ na_if(.x, -1))) %>%
# replace NAs with 0s
replace_na(list(
  RuleGPcoordovs = 0,
  RuleGPdeverbaddr = 0,
  RuleGPpatinstr = 0,
  RuleGPdeverbsubj = 0,
  RuleGPadjective = 0,
  RuleGPPatbenperson = 0,
  RuleGPwordorder = 0,
  RuleDoubleAdpos = 0,
  RuleDoubleAdpos.max_allowable_distance.v = 0,
  RuleAmbiguousRegards = 0,
  RuleReflexivePassWithAnimSubj = 0,
  RuleTooManyNegations = 0,
  RuleTooManyNegations.max_negation_frac.v = 0,
  RuleTooManyNegations.max_allowable_negations.v = 0,
  RuleTooManyNominalConstructions.max_noun_frac.v = 0,
  RuleTooManyNominalConstructions.max_allowable_nouns.v = 0,
  RuleFunctionWordRepetition = 0,
  RuleCaseRepetition.max_repetition_count.v = 0,
  RuleCaseRepetition.max_repetition_frac.v = 0,
  RuleWeakMeaningWords = 0,
  RuleAbstractNouns = 0,
  RuleRelativisticExpressions = 0,
  RuleConfirmationExpressions = 0,
  RuleRedundantExpressions = 0,
  RuleTooLongExpressions = 0,
  RuleAnaphoricReferences = 0,
  RuleLiteraryStyle = 0,
  RulePassive = 0,
  RulePredSubjDistance = 0,
  RulePredSubjDistance.max_distance.v = 0,
  RulePredObjDistance = 0,
  RulePredObjDistance.max_distance.v = 0,
  RuleInfVerbDistance = 0,
  RuleInfVerbDistance.max_distance.v = 0,
  RuleMultiPartVerbs = 0,
  RuleMultiPartVerbs.max_distance.v = 0,
  RuleLongSentences.max_length.v = 0,
  RulePredAtClauseBeginning.max_order.v = 0,
  RuleVerbalNouns = 0,
  RuleDoubleComparison = 0,
  RuleWrongValencyCase = 0,
  RuleWrongVerbonominalCase = 0,
  RuleIncompleteConjunction = 0
)) %>%
# merge GPs
mutate(
  GPs = RuleGPcoordovs +
    RuleGPdeverbaddr +
    RuleGPpatinstr +
    RuleGPdeverbsubj +

```

```

    RuleGPadjective +
    RuleGPPatbenperson +
    RuleGPwordorder
) %>%
select(!c(
  RuleGPcoordovs,
  RuleGPdeverbaddr,
  RuleGPPatinstr,
  RuleGPdeverbsubj,
  RuleGPadjective,
  RuleGPPatbenperson,
  RuleGPwordorder
)) %>%
# norm data expected to correlate with text length
mutate(across(c(
  GPs,
  RuleDoubleAdpos,
  RuleAmbiguousRegards,
  RuleFunctionWordRepetition,
  RuleWeakMeaningWords,
  RuleAbstractNouns,
  RuleRelativisticExpressions,
  RuleConfirmationExpressions,
  RuleRedundantExpressions,
  RuleTooLongExpressions,
  RuleAnaphoricReferences,
  RuleLiteraryStyle,
  RulePassive,
  RuleVerbalNouns,
  RuleDoubleComparison,
  RuleWrongValencyCase,
  RuleWrongVerbominalCase,
  RuleIncompleteConjunction,
  num_hapax,
  RuleReflexivePassWithAnimSubj,
  RuleTooManyNominalConstructions,
  RulePredSubjDistance,
  RuleMultiPartVerbs,
  RulePredAtClauseBeginning
), ~ .x / word_count)) %>%
mutate(across(c(
  RuleTooFewVerbs,
  RuleTooManyNegations,
  RuleCaseRepetition,
  RuleLongSentences,
  RulePredObjDistance,
  RuleInfVerbDistance
), ~ .x / sent_count)) %>%
# replace NAs with medians
mutate(across(c(
  RuleDoubleAdpos.max_allowable_distance,
  RuleTooManyNegations.max_negation_frac,
  RuleTooManyNegations.max_allowable_negations,

```

```

    RulePredSubjDistance.max_distance,
    RulePredObjDistance.max_distance,
    RuleInfVerbDistance.max_distance,
    RuleMultiPartVerbs.max_distance
  ), ~ coalesce(., median(., na.rm = TRUE)))

data_clean <- data_no_nas %>%
  # remove variables identified as text-length dependent
  select(!c(
    RuleTooFewVerbs,
    RuleTooManyNegations,
    RuleTooManyNominalConstructions,
    RuleCaseRepetition,
    RuleLongSentences,
    RulePredAtClauseBeginning,
    syllab_count,
    char_count
  )) %>%
  # remove variables identified as unreliable
  select(!c(
    RuleAmbiguousRegards,
    RuleFunctionWordRepetition,
    RuleDoubleComparison,
    RuleWrongValencyCase,
    RuleWrongVerbNominalCase
  )) %>%
  # remove further variables belonging to the 'acceptability' category
  select(!c(RuleIncompleteConjunction)) %>%
  # remove artificially limited variables
  select(!c(
    RuleCaseRepetition.max_repetition_frac,
    RuleCaseRepetition.max_repetition_frac.v
  )) %>%
  # remove variables with too many NAs
  select(!c(
    RuleDoubleAdpos.max_allowable_distance,
    RuleDoubleAdpos.max_allowable_distance.v
  )) %>%
  mutate(across(c(
    class,
    FileFormat,
    subcorpus,
    DocumentVersion,
    LegalActType,
    Objectivity,
    AuthorType,
    RecipientType,
    RecipientIndividuation,
    Anonymized
  ), ~ as.factor(.x)))

# no NAs should be present now
data_clean[!complete.cases(

```



```
data_clean[.firstnonmetacolumn:ncol(data_clean)]
), .firstnonmetacolumn:ncol(data_clean)] %>% as.data.frame()
```

```
## [1] RuleAbstractNouns
## [2] RuleAnaphoricReferences
## [3] RuleCaseRepetition.max_repetition_count
## [4] RuleCaseRepetition.max_repetition_count.v
## [5] RuleConfirmationExpressions
## [6] RuleDoubleAdpos
## [7] RuleInfVerbDistance
## [8] RuleInfVerbDistance.max_distance
## [9] RuleInfVerbDistance.max_distance.v
## [10] RuleLiteraryStyle
## [11] RuleLongSentences.max_length
## [12] RuleLongSentences.max_length.v
## [13] RuleMultiPartVerbs
## [14] RuleMultiPartVerbs.max_distance
## [15] RuleMultiPartVerbs.max_distance.v
## [16] RulePassive
## [17] RulePredAtClauseBeginning.max_order
## [18] RulePredAtClauseBeginning.max_order.v
## [19] RulePredObjDistance
## [20] RulePredObjDistance.max_distance
## [21] RulePredObjDistance.max_distance.v
## [22] RulePredSubjDistance
## [23] RulePredSubjDistance.max_distance
## [24] RulePredSubjDistance.max_distance.v
## [25] RuleRedundantExpressions
## [26] RuleReflexivePassWithAnimSubj
## [27] RuleRelativisticExpressions
## [28] RuleTooFewVerbs.min_verb_frac
## [29] RuleTooFewVerbs.min_verb_frac.v
## [30] RuleTooLongExpressions
## [31] RuleTooManyNegations.max_allowable_negations
## [32] RuleTooManyNegations.max_allowable_negations.v
## [33] RuleTooManyNegations.max_negation_frac
## [34] RuleTooManyNegations.max_negation_frac.v
## [35] RuleTooManyNominalConstructions.max_allowable_nouns
## [36] RuleTooManyNominalConstructions.max_allowable_nouns.v
## [37] RuleTooManyNominalConstructions.max_noun_frac
## [38] RuleTooManyNominalConstructions.max_noun_frac.v
## [39] RuleVerbalNouns
## [40] RuleWeakMeaningWords
## [41] activity
## [42] ari
## [43] atl
## [44] cli
## [45] entropy
## [46] fkg1
## [47] fre
## [48] gf
## [49] hpoint
## [50] maentropy
## [51] maentropy.v
```

```
## [52] mamr
## [53] mattr
## [54] mattr.v
## [55] num_hapax
## [56] sent_count
## [57] smog
## [58] ttr
## [59] verb_dist
## [60] word_count
## [61] GPs
## <0 rows> (or 0-length row.names)

colnames(data_clean) <- prettify_feat_name_vector(colnames(data_clean))

data_clean_scaled <- data_clean %>%
  mutate(across(class, ~ .x == "good")) %>%
  mutate(across(.firstnonmetacolumn:ncol(data_clean), ~ scale(.x)))

## Warning: There was 1 warning in `mutate()`.
## i In argument: `across(.firstnonmetacolumn:ncol(data_clean), ~scale(.x))`.
## Caused by warning:
## ! Using an external vector in selections was deprecated in tidysselect 1.1.0.
## i Please use `all_of()` or `any_of()` instead.
##   # Was:
##   data %>% select(.firstnonmetacolumn)
##
##   # Now:
##   data %>% select(all_of(.firstnonmetacolumn))
##
## See <https://tidysselect.r-lib.org/reference/faq-external-vector.html>.
```

Important features identification

Regularized regression

split the data

```
.no_folds <- 10
.split_prop <- 4 / 5

data_split <- initial_split(data_clean, strata = class, prop = .split_prop)
training_set <- training(data_split)
testing_set <- testing(data_split)

folds <- vfold_cv(training_set, .no_folds)
```

recipe

```
lin_formula <- reformulate(colnames(data_clean)[17:77], "class")
lin_rec <- recipe(lin_formula, data = training_set) %>%
  # step_corr(all_predictors()) %>%
  step_normalize(all_predictors())

lin_wf_base <- workflow() %>% add_recipe(lin_rec)
```

tuning

```

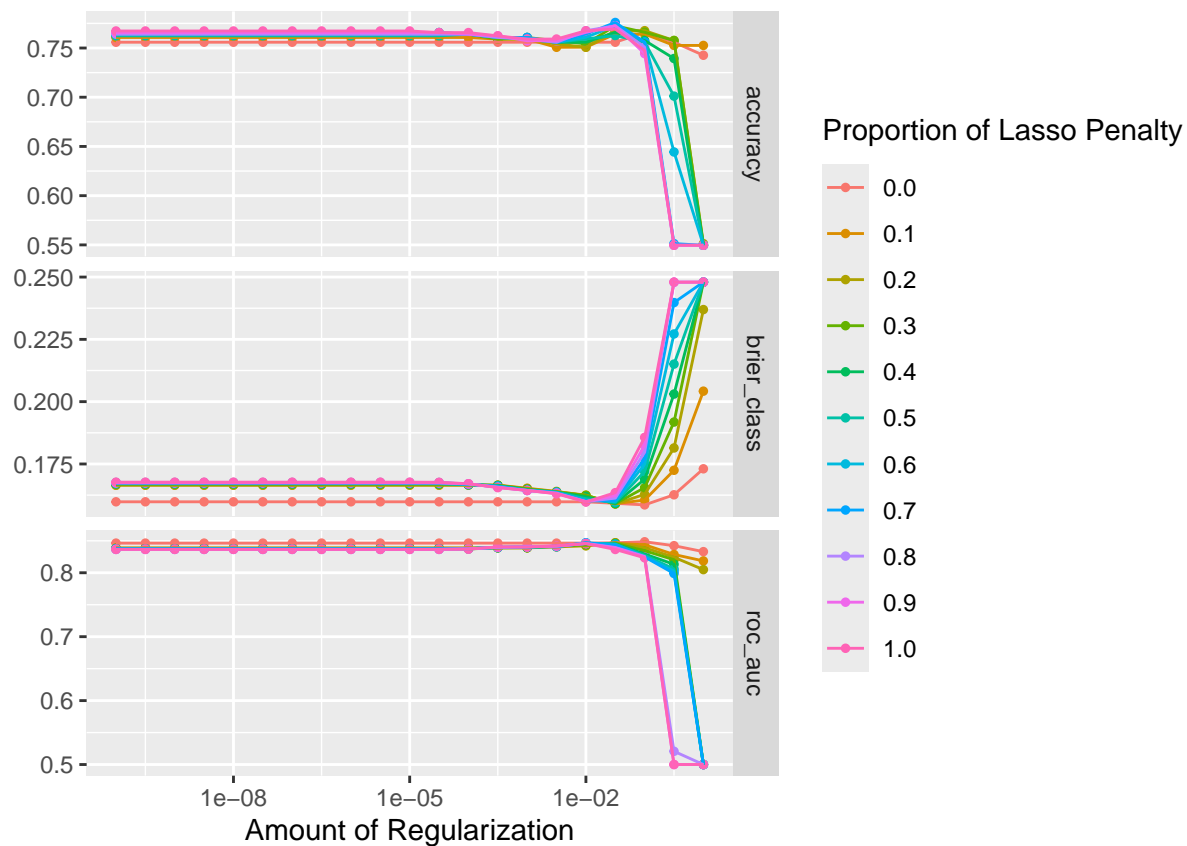
lin_wf <- lin_wf_base %>%
  add_model(logistic_reg(
    mode = "classification", engine = "glmnet",
    penalty = tune(), mixture = tune()
  ))

tune_grid <- grid_regular(
  penalty(), mixture(),
  levels = c(penalty = 21, mixture = 11)
)

tune_rs <- tune_grid(
  lin_wf, folds,
  grid = tune_grid,
  metrics = metric_set(yardstick::accuracy, brier_class, roc_auc)
)

autoplot(tune_rs)

```



```

choose_roc_auc <- tune_rs %>%
  select_by_one_std_err(metric = "roc_auc", -mixture, penalty)
choose_roc_auc

```

```

## # A tibble: 1 x 3
##   penalty mixture .config
##   <dbl>   <dbl> <chr>
## 1 0.0000000001     1 Preprocessor1_Model211

```

final

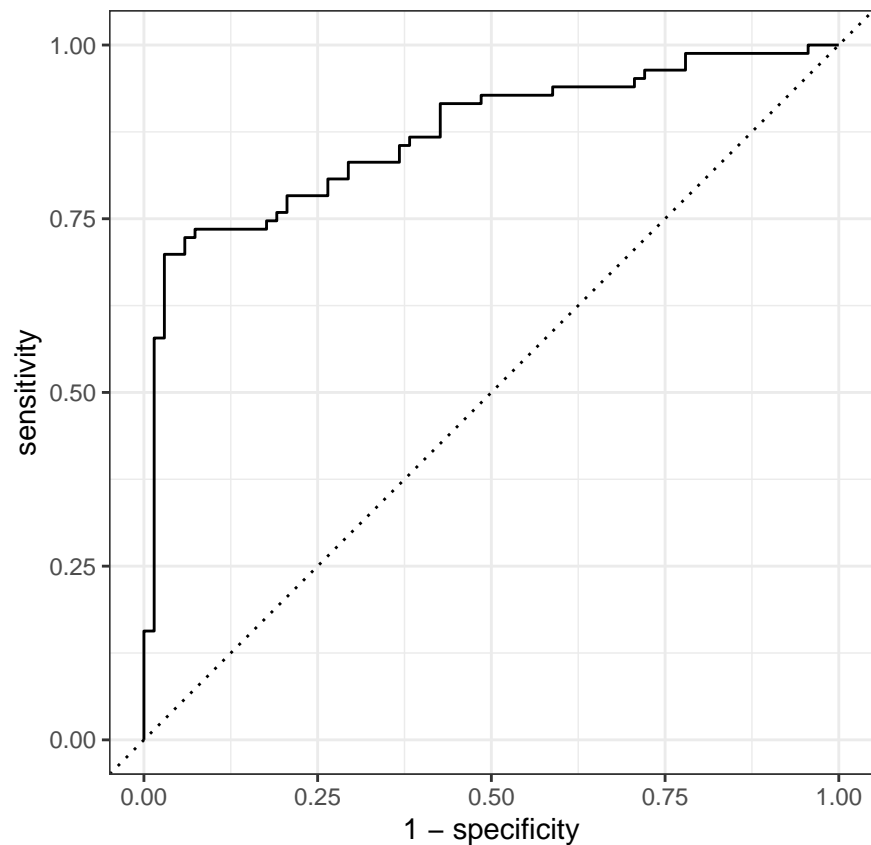
```
lin_final_wf <- finalize_workflow(lin_wf, choose_roc_auc)
lin_final_wf

## == Workflow =====
## Preprocessor: Recipe
## Model: logistic_reg()
##
## -- Preprocessor -----
## 1 Recipe Step
##
## * step_normalize()
##
## -- Model -----
## Logistic Regression Model Specification (classification)
##
## Main Arguments:
##   penalty = 1e-10
##   mixture = 1
##
## Computational engine: glmnet
lin_final_fitted <- last_fit(lin_final_wf, data_split)

collect_predictions(lin_final_fitted) %>%
  conf_mat(truth = class, estimate = .pred_class)

##           Truth
## Prediction bad good
##           bad   64   14
##           good   19   54

collect_predictions(lin_final_fitted) %>%
  roc_curve(truth = class, .pred_bad) %>%
  autoplot()
```



```
extract_fit_parsnip(lin_final_fitted) %>%
  vip::vi(lambda = choose_roc_auc$penalty) %>%
  print(n = 80)
```

```
## # A tibble: 61 x 3
##   Variable      Importance Sign
##   <chr>          <dbl> <chr>
## 1 sentlen.m      2.99  POS
## 2 ari            2.64  NEG
## 3 gf             1.96  NEG
## 4 sentcount      1.86  POS
## 5 atl            1.41  POS
## 6 activity        1.37  POS
## 7 VERBfrac.m     1.32  NEG
## 8 smog           1.17  POS
## 9 hpoint          1.13  NEG
## 10 wordcount      1.05  NEG
## 11 ttr            0.886  NEG
## 12 fre            0.806  NEG
## 13 entropy.v      0.720  POS
## 14 entropy        0.693  NEG
## 15 sentlen.v      0.580  POS
## 16 ttr.v          0.541  NEG
## 17 predsubjdist.m 0.493  NEG
## 18 anaphoricrefs  0.447  POS
## 19 cli            0.430  NEG
## 20 extrcaseexprs  0.411  POS
```

## 21	compoundVERBs	0.410	POS
## 22	passives	0.402	NEG
## 23	mattr	0.347	NEG
## 24	caserepcount.v	0.339	NEG
## 25	predobjdist.m	0.321	NEG
## 26	literary	0.314	NEG
## 27	verbdist	0.308	POS
## 28	caserepcount.m	0.307	POS
## 29	maentropy	0.285	POS
## 30	predorder.m	0.267	NEG
## 31	hapaxes	0.263	POS
## 32	VERBcomp	0.247	POS
## 33	NOUNcount.v	0.227	NEG
## 34	subj	0.223	POS
## 35	NOUNcount.m	0.212	POS
## 36	VERBcompdist.v	0.208	NEG
## 37	predobjdist.v	0.203	POS
## 38	rftpass_animsbj	0.197	NEG
## 39	NEGcount.m	0.188	POS
## 40	NOUNfrac.m	0.184	NEG
## 41	longexprs	0.179	POS
## 42	redundexprs	0.177	NEG
## 43	compoundVERBsdist.m	0.175	NEG
## 44	doubleADPs	0.168	NEG
## 45	VERBfrac.v	0.157	POS
## 46	relativisticexprs	0.157	NEG
## 47	NEGcount.v	0.145	NEG
## 48	compoundVERBsdist.v	0.139	POS
## 49	NEGfrac.v	0.126	POS
## 50	VERBcompdist.m	0.126	POS
## 51	GPs	0.105	NEG
## 52	predsubjdist.v	0.0944	NEG
## 53	mamr	0.0940	NEG
## 54	NOUNfrac.v	0.0857	POS
## 55	obj	0.0766	POS
## 56	weakmeaning	0.0758	NEG
## 57	predorder.v	0.0467	POS
## 58	verbalNOUNs	0.0348	NEG
## 59	abstractNOUNs	0.00983	POS
## 60	NEGfrac.m	0.000988	POS
## 61	fkgl	0	NEG

```
lin_final_fitted %>%
  extract_fit_parsnip() %>%
  tidy() %>%
  arrange(estimate) %>%
  print(n = 80)
```

```
## # A tibble: 62 x 3
##   term          estimate    penalty
##   <chr>         <dbl>      <dbl>
## 1 ari          -2.64      0.0000000001
## 2 gf           -1.96      0.0000000001
## 3 VERBfrac.m   -1.32      0.0000000001
## 4 hpoint       -1.13      0.0000000001
```

## 5 wordcount	-1.05	0.0000000001
## 6 ttr	-0.886	0.0000000001
## 7 fre	-0.806	0.0000000001
## 8 entropy	-0.693	0.0000000001
## 9 (Intercept)	-0.542	0.0000000001
## 10 ttr.v	-0.541	0.0000000001
## 11 predsubjdist.m	-0.493	0.0000000001
## 12 cli	-0.430	0.0000000001
## 13 passives	-0.402	0.0000000001
## 14 mattr	-0.347	0.0000000001
## 15 caserepcount.v	-0.339	0.0000000001
## 16 predobjdist.m	-0.321	0.0000000001
## 17 literary	-0.314	0.0000000001
## 18 predorder.m	-0.267	0.0000000001
## 19 NOUNcount.v	-0.227	0.0000000001
## 20 VERBcompdist.v	-0.208	0.0000000001
## 21 rfpass_animsubj	-0.197	0.0000000001
## 22 NOUNfrac.m	-0.184	0.0000000001
## 23 redundexprs	-0.177	0.0000000001
## 24 compoundVERBsdist.m	-0.175	0.0000000001
## 25 doubleADPs	-0.168	0.0000000001
## 26 relativisticexprs	-0.157	0.0000000001
## 27 NEGcount.v	-0.145	0.0000000001
## 28 GPs	-0.105	0.0000000001
## 29 predsubjdist.v	-0.0944	0.0000000001
## 30 mamr	-0.0940	0.0000000001
## 31 weakmeaning	-0.0758	0.0000000001
## 32 verbalNOUNs	-0.0348	0.0000000001
## 33 fkg1	0	0.0000000001
## 34 NEGfrac.m	0.000988	0.0000000001
## 35 abstractNOUNs	0.00983	0.0000000001
## 36 predorder.v	0.0467	0.0000000001
## 37 obj	0.0766	0.0000000001
## 38 NOUNfrac.v	0.0857	0.0000000001
## 39 VERBcompdist.m	0.126	0.0000000001
## 40 NEGfrac.v	0.126	0.0000000001
## 41 compoundVERBsdist.v	0.139	0.0000000001
## 42 VERBfrac.v	0.157	0.0000000001
## 43 longexprs	0.179	0.0000000001
## 44 NEGcount.m	0.188	0.0000000001
## 45 predobjdist.v	0.203	0.0000000001
## 46 NOUNcount.m	0.212	0.0000000001
## 47 subj	0.223	0.0000000001
## 48 VERBcomp	0.247	0.0000000001
## 49 hapaxes	0.263	0.0000000001
## 50 maentropy	0.285	0.0000000001
## 51 caserepcount.m	0.307	0.0000000001
## 52 verbdist	0.308	0.0000000001
## 53 compoundVERBs	0.410	0.0000000001
## 54 extrcaseexprs	0.411	0.0000000001
## 55 anaphoricrefs	0.447	0.0000000001
## 56 sentlen.v	0.580	0.0000000001
## 57 entropy.v	0.720	0.0000000001
## 58 smog	1.17	0.0000000001

```
## 59 activity          1.37      0.0000000001
## 60 atl               1.41      0.0000000001
## 61 sentcount         1.86      0.0000000001
## 62 sentlen.m         2.99      0.0000000001
```

Individual regressions

```
data_scaled <- data_clean %>%
  mutate(across(all_of(.firstnonmetacolumn:ncol(data_clean)), ~ scale(.x)[, 1]))

feature_importances <- tibble(
  feat_name = character(),
  p_value = numeric(),
  estimate = numeric(),
  wilcox_p = numeric(),
  wilcox_r = numeric(),
  kw_p = numeric(),
  kw_chi2 = numeric(),
  kw_epsilon2 = numeric(),
  kw_epsilon2_lci = numeric(),
  kw_epsilon2_uci = numeric(),
  med_sign = numeric(),
  mean_sign = numeric()
)

for (i in .firstnonmetacolumn:ncol(data_scaled)) {
  fname <- names(data_scaled)[i]
  message(fname)

  formula_single <- reformulate(fname, "class")
  formula_single_reversed <- reformulate("class", fname)

  glm_model <- glm(formula_single, data_scaled, family = "binomial")
  glm_coefficients <- summary(glm_model)$coefficients
  row_index <- which(rownames(glm_coefficients) == fname)
  p_value <- glm_coefficients[row_index, 4]
  beta <- glm_coefficients[row_index, 1]

  wilcox_p <- wilcox.test(formula_single_reversed, data_scaled)$p.value
  wilcox_r <- wilcox_effsize(data_scaled, formula_single_reversed)$effsize[[1]]

  kw <- kruskal.test(data_scaled[[fname]], data_scaled$class)
  kw_p <- kw$p.value
  kw_chi2 <- kw$statistic[[1]]
  kw_epsilon2_t <- epsilonSquared(
    data_scaled[[fname]], data_scaled$class,
    ci = TRUE
  )
  kw_epsilon2 <- kw_epsilon2_t[[1]]
  kw_epsilon2_lci <- kw_epsilon2_t[[2]]
  kw_epsilon2_uci <- kw_epsilon2_t[[3]]

  med_good <- filter(data_scaled, class == "good")[[fname]] %>% median()
```



```

med_bad <- filter(data_scaled, class == "bad")[[fname]] %>% median()
med_sign <- sign(med_good - med_bad)

mean_good <- filter(data_scaled, class == "good")[[fname]] %>% mean()
mean_bad <- filter(data_scaled, class == "bad")[[fname]] %>% mean()
mean_sign <- sign(mean_good - mean_bad)

feature_importances <- feature_importances %>%
  add_row(
    feat_name = fname,
    p_value = p_value,
    estimate = beta,
    wilcox_p = wilcox_p,
    wilcox_r = wilcox_r,
    kw_p = kw_p,
    kw_chi2 = kw_chi2,
    kw_epsilon2 = kw_epsilon2,
    kw_epsilon2_uci = kw_epsilon2_uci,
    kw_epsilon2_lci = kw_epsilon2_lci,
    med_sign = med_sign,
    mean_sign = mean_sign,
  )
}

```

```

## abstractNOUNs
## anaphoricrefs
## caserepcount.m
## caserepcount.v
## extrcaseexprs
## doubleADPs
## VERBcomp
## VERBcompdist.m
## VERBcompdist.v
## literary
## sentlen.m
## sentlen.v
## compoundVERBs
## compoundVERBsdist.m
## compoundVERBsdist.v
## passives
## predorder.m
## predorder.v
## obj
## predobjdist.m

```

```
## predobjdist.v
## subj
## predsubjdist.m
## predsubjdist.v
## redundexprs
## rfpass_animsubj
## relativisticexprs
## VERBfrac.m
## VERBfrac.v
## longexprs
## NEGcount.m
## NEGcount.v
## NEGfrac.m
## NEGfrac.v
## NOUNcount.m
## NOUNcount.v
## NOUNfrac.m
## NOUNfrac.v
## verbalNOUNs
## weakmeaning
## activity
## ari
## atl
## cli
## entropy
## fkg1
## fre
## gf
## hpoint
## maentropy
## entropy.v
## mamr
## mattr
## ttr.v
## hapaxes
## sentcount
```

```

## smog
## ttr
## verbdist
## wordcount
## GPs
feature_importances

## # A tibble: 61 x 12
##   feat_name      p_value estimate wilcox_p wilcox_r      kw_p kw_chi2 kw_epsilon2
##   <chr>          <dbl>    <dbl>    <dbl>    <dbl>    <dbl> <dbl>    <dbl>
## 1 abstractNOU~ 2.20e- 3    0.232 6.39e- 3    0.0994 6.39e- 3    7.44    0.00989
## 2 anaphoricre~ 6.73e- 1    0.0308 9.80e- 3    0.0941 9.79e- 3    6.67    0.00887
## 3 caserepcoun~ 6.59e- 2   -0.137 7.61e- 2    0.0647 7.60e- 2    3.15    0.00419
## 4 caserepcoun~ 4.54e- 3   -0.215 9.43e- 4    0.121 9.43e- 4   10.9    0.0145
## 5 extrcaseexp~ 1.08e- 1   -0.123 1.34e- 3    0.117 1.34e- 3   10.3    0.0137
## 6 doubleADPs   2.71e- 1   -0.0816 3.02e- 1    0.0376 3.02e- 1    1.06    0.00141
## 7 VERBcomp     5.24e-15    0.659 1.36e-16    0.301 1.36e-16   68.4    0.0909
## 8 VERBcompdis~ 5.48e- 2   -0.191 1.73e- 2    0.0868 1.73e- 2    5.67    0.00754
## 9 VERBcompdis~ 6.58e- 2   -0.137 7.90e- 2    0.0640 7.89e- 2    3.09    0.0041
## 10 literary    7.00e-21   -0.918 1.44e-26    0.389 1.44e-26   114.    0.151
## # i 51 more rows
## # i 4 more variables: kw_epsilon2_lci <dbl>, kw_epsilon2_uci <dbl>,
## #   med_sign <dbl>, mean_sign <dbl>

selected_features <- feature_importances %>%
  mutate(
    selected = p_value <= 0.05,
    wilcox_sel = wilcox_p < 0.05,
    kw_sel = kw_p < 0.05
  )

selected_features %>%
  select(selected, kw_sel) %>%
  table()

##           kw_sel
## selected FALSE TRUE
##   FALSE      8    4
##   TRUE       4   45

cor(-log(selected_features$p_value), selected_features$kw_epsilon2)

## [1] 0.952316

cor(-log(selected_features$p_value), -log(selected_features$kw_p))

## [1] 0.9524106

cor(selected_features$estimate, selected_features$kw_epsilon2)

## [1] -0.3662002

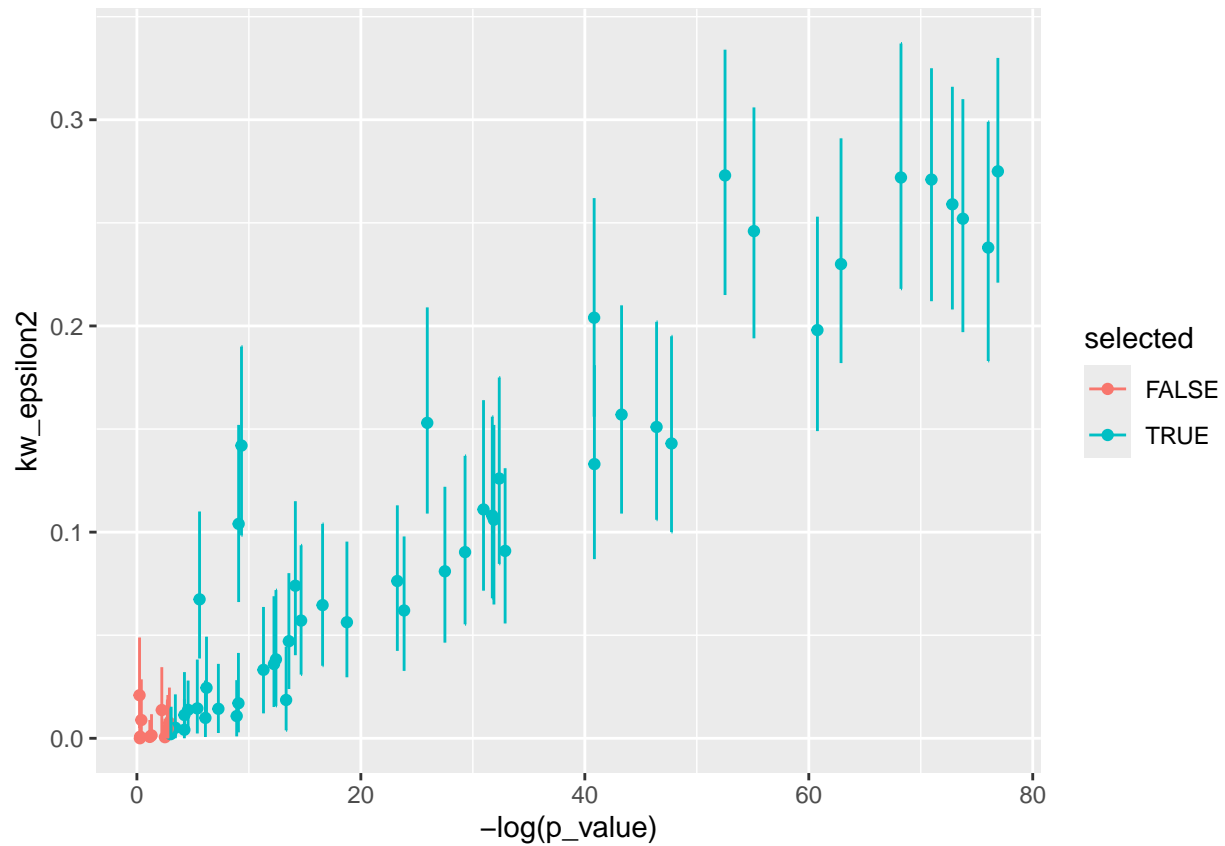
selected_features %>%
  ggplot(aes(
    x = -log(p_value), y = kw_epsilon2,

```

```

    ymin = kw_epsilon2_lci, ymax = kw_epsilon2_uci, color = selected
  )) +
  geom_point() +
  geom_errorbar()

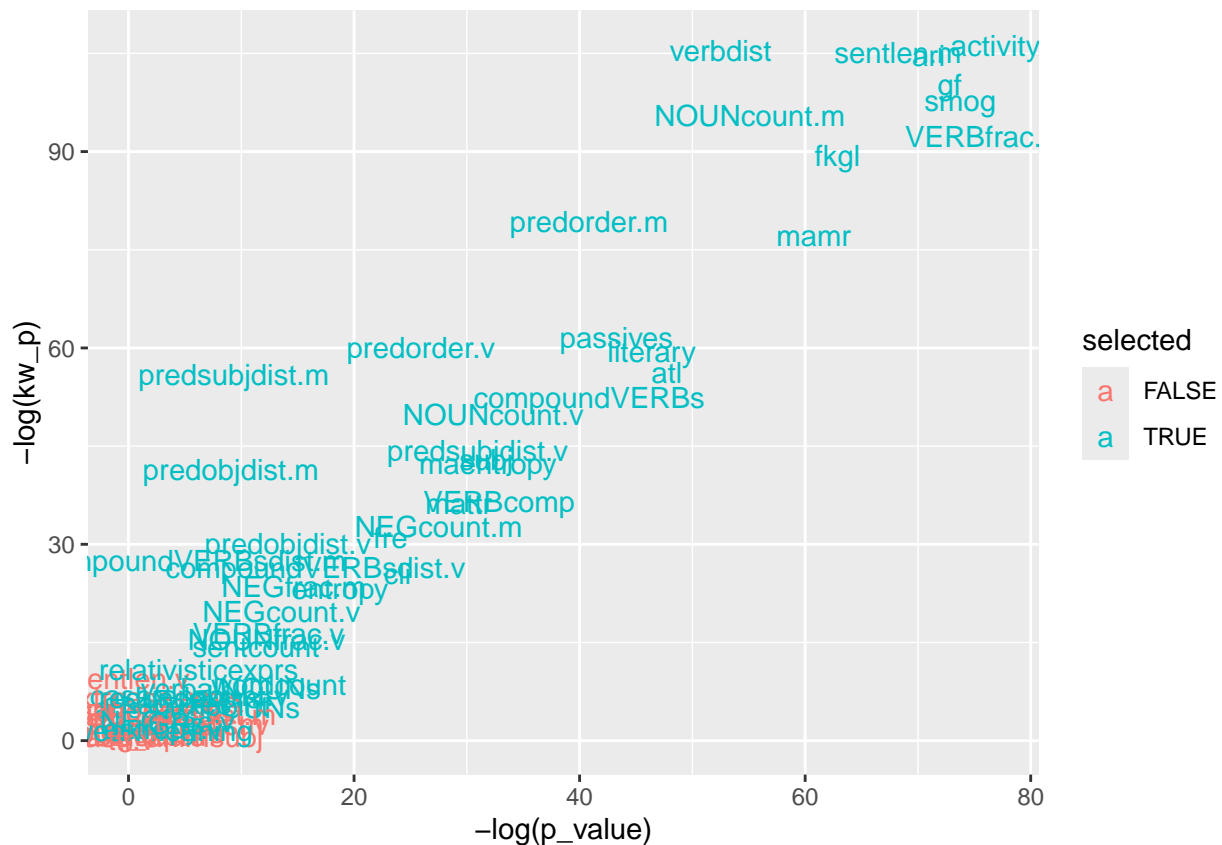
```



```

selected_features %>%
  ggplot(aes(
    x = -log(p_value), y = -log(kw_p), color = selected, label = feat_name
  )) +
  # geom_point() +
  geom_text()

```



```
selected_features_names <- selected_features %>%  
  filter(selected) %>%  
  pull(feats_name)
```

Compare the two

```
featcomp <- extract_fit_parsnip(lin_final_fitted) %>%
  vip::vi(lambda = choose_roc_auc$penalty) %>%
  full_join(
    selected_features %>% rename(Variable = feat_name),
    by = "Variable"
  ) %>%
  rename(selected_pval = selected) %>%
  mutate(
    log_p = -log(p_value),
    log_wilcox_p = -log(wilcox_p),
    log_kw_p = -log(kw_p),
    selected_reg = Importance > 0
  )

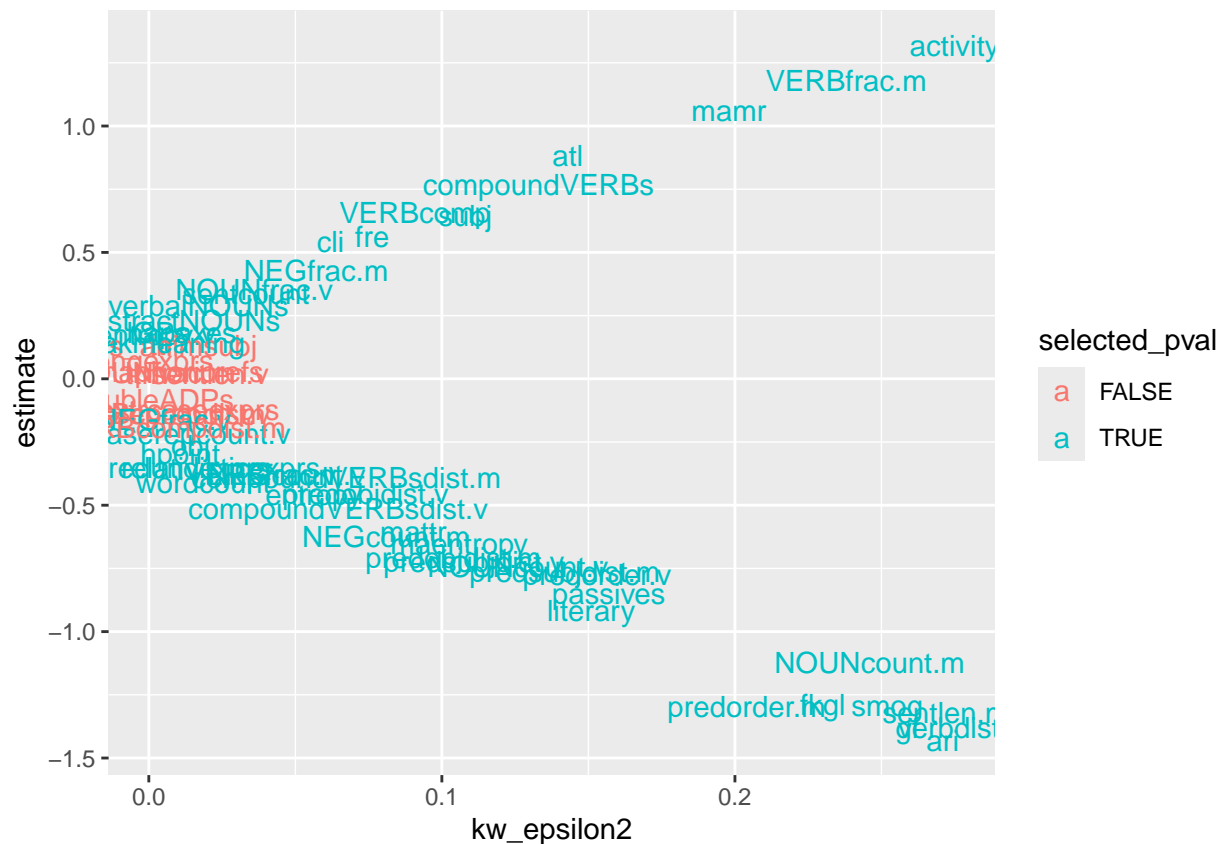
featcomp %>% write_csv("featcomp.csv")

featcomp %>%
  filter(!is.na(Importance)) %>%
  select(Importance, kw_epsilon2, log_p, log_kw_p) %>%
  cor() %>%
```

```
round(2)
```

```
##          Importance kw_epsilon2 log_p log_kw_p
## Importance      1.00      0.47  0.51   0.47
## kw_epsilon2     0.47      1.00  0.95   1.00
## log_p           0.51      0.95  1.00   0.95
## log_kw_p        0.47      1.00  0.95   1.00
```

```
featcomp %>%
  ggplot(aes(
    x = kw_epsilon2, y = estimate, color = selected_pval, label = Variable
  )) +
  geom_text()
```



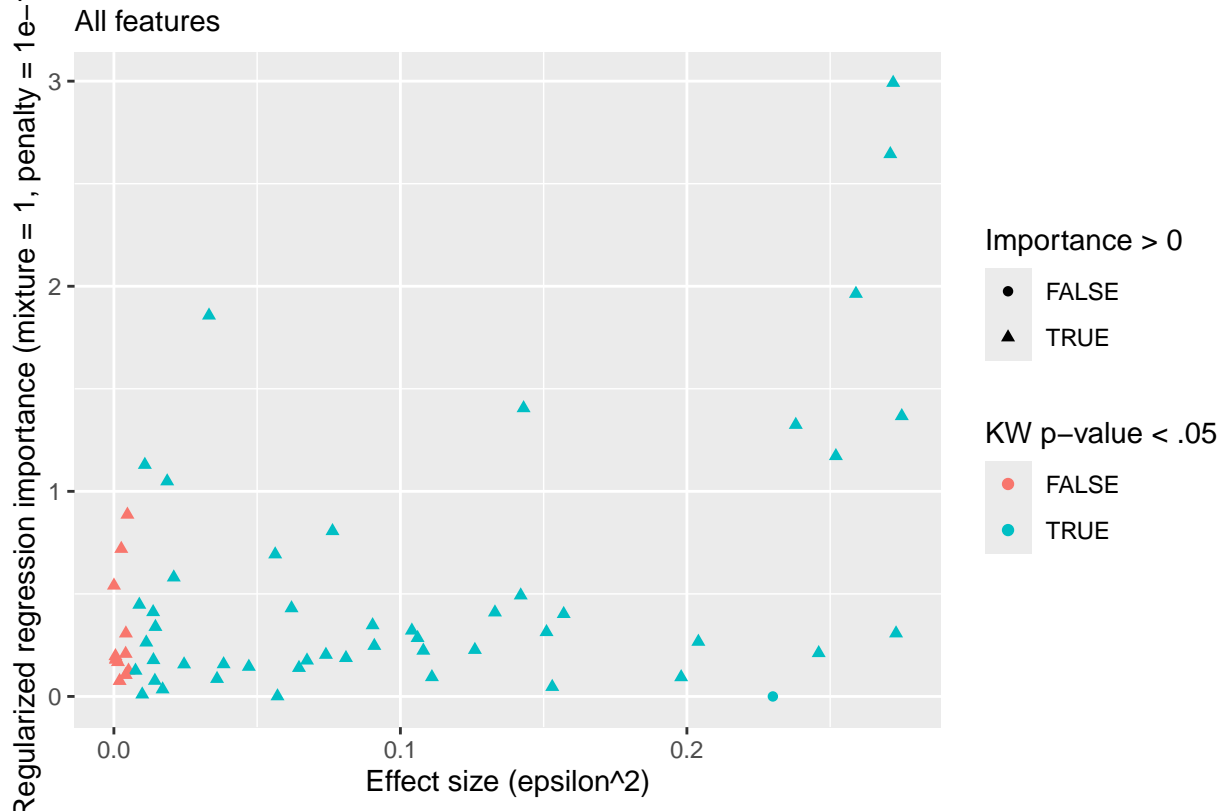
```
featcomp_plot <- featcomp %>% ggplot(aes(
  x = kw_epsilon2,
  y = Importance,
  # size = log_p,
  color = kw_sel,
  shape = selected_reg
)) +
  geom_point() +
  labs(
    title = "Feature importance measures",
    subtitle = "All features",
    # subtitle = "Features with |r| < 0.90",
    x = "Effect size (epsilon^2)",
    y = paste0(c(
```

```

    "Regularized regression importance (mixture = ",
    choose_roc_auc$mixture[1], ", penalty = ",
    choose_roc_auc$penalty[1], ")"
  ), collapse = ""),
  # size = "-log(p-value)",
  color = "KW p-value < .05",
  shape = "Importance > 0"
)
print(featcomp_plot)

```

Feature importance measures



```
ggsave("featcomp_all.png")
```

```
## Saving 6.5 x 4.5 in image
```

```
# ggsave("featcomp_nocorr.png")
```

Results

```

featcomp %>%
  filter(!kw_sel) %>%
  select(Variable, kw_chi2, kw_p) %>%
  arrange(Variable) %>%
  as.data.frame() %>%
  print(digits = 2)

```

```
##           Variable kw_chi2 kw_p
```

```
## 1          GPs      3.116 0.078
## 2      NEGfrac.v    3.835 0.050
## 3      NOUNfrac.m   0.582 0.446
## 4  VERBcompdist.v   3.087 0.079
## 5  caserepcount.m   3.148 0.076
## 6      doubleADPs   1.064 0.302
## 7      entropy.v    1.937 0.164
## 8      longexprs    0.513 0.474
## 9  rfpass_animsubj   0.414 0.520
## 10         ttr      3.550 0.060
## 11         ttr.v    0.022 0.882
## 12    weakmeaning    1.504 0.220
```

```
featcomp %>%
  filter(kw_sel) %>%
  mutate(signed_effect = kw_epsilon2 * mean_sign) %>%
  select(Variable, kw_epsilon2, kw_p, signed_effect) %>%
  arrange(-kw_epsilon2) %>%
  as.data.frame() %>%
  print(digits = 2)
```

```
##          Variable kw_epsilon2   kw_p signed_effect
## 1      activity    0.2750 6.9e-47     0.2750
## 2      verbdist    0.2730 1.7e-46    -0.2730
## 3      sentlen.m   0.2720 2.2e-46    -0.2720
## 4          ari     0.2710 3.2e-46    -0.2710
## 5          gf      0.2590 2.7e-44    -0.2590
## 6          smog     0.2520 3.4e-43    -0.2520
## 7      NOUNcount.m  0.2460 3.4e-42    -0.2460
## 8      VERBfrac.m   0.2380 7.7e-41     0.2380
## 9          fkg1     0.2300 1.4e-39    -0.2300
## 10     predorder.m  0.2040 3.5e-35    -0.2040
## 11          mamr     0.1980 2.9e-34     0.1980
## 12     passives     0.1570 1.9e-27    -0.1570
## 13     predorder.v  0.1530 7.8e-27    -0.1530
## 14     literary     0.1510 1.4e-26    -0.1510
## 15          atl     0.1430 3.6e-25     0.1430
## 16     predsubjdist.m 0.1420 5.2e-25    -0.1420
## 17     compoundVERBs 0.1330 1.8e-23     0.1330
## 18     NOUNcount.v   0.1260 2.2e-22    -0.1260
## 19     predsubjdist.v 0.1110 6.0e-20    -0.1110
## 20          subj     0.1080 2.2e-19     0.1080
## 21     maentropy     0.1060 4.3e-19    -0.1060
## 22     predobjdist.m 0.1040 1.1e-18    -0.1040
## 23     VERBcomp     0.0909 1.4e-16     0.0909
## 24          mattr     0.0903 1.7e-16    -0.0903
## 25     NEGcount.m    0.0810 5.9e-15    -0.0810
## 26          fre      0.0763 3.6e-14     0.0763
## 27     predobjdist.v 0.0740 8.6e-14    -0.0740
## 28 compoundVERBsdist.m 0.0674 1.1e-12    -0.0674
## 29 compoundVERBsdist.v 0.0646 3.2e-12    -0.0646
## 30          cli      0.0620 8.5e-12     0.0620
## 31     NEGfrac.m     0.0571 5.7e-11     0.0571
## 32     entropy      0.0563 7.6e-11    -0.0563
## 33     NEGcount.v    0.0471 2.6e-09    -0.0471
```



```
## 34      VERBfrac.v      0.0383 8.1e-08      -0.0383
## 35      NOUNfrac.v      0.0360 2.0e-07       0.0360
## 36      sentcount      0.0332 5.9e-07       0.0332
## 37      relativisticexprs 0.0245 1.8e-05      -0.0245
## 38      sentlen.v      0.0209 7.2e-05       0.0209
## 39      wordcount      0.0186 1.8e-04      -0.0186
## 40      verbalNOUNs     0.0170 3.6e-04       0.0170
## 41      caserepcount.v  0.0145 9.4e-04      -0.0145
## 42      obj            0.0143 1.0e-03      -0.0143
## 43      redundexprs     0.0138 1.3e-03      -0.0138
## 44      extrcaseexprs   0.0137 1.3e-03      -0.0137
## 45      hapaxes        0.0113 3.5e-03       0.0113
## 46      hpoint         0.0108 4.4e-03      -0.0108
## 47      abstractNOUNs   0.0099 6.4e-03       0.0099
## 48      anaphoricrefs   0.0089 9.8e-03       0.0089
## 49      VERBcompdist.m  0.0075 1.7e-02      -0.0075
```

```
featcomp %>%
  filter(kw_sel) %>%
  select(
    Variable,
    kw_chi2,
    kw_p,
    kw_epsilon2_lci,
    kw_epsilon2,
    kw_epsilon2_uci,
    mean_sign
  ) %>%
  arrange(-kw_epsilon2) %>%
  print(n = 100)
```

```
## # A tibble: 49 x 7
##   Variable      kw_chi2      kw_p kw_epsilon2_lci kw_epsilon2 kw_epsilon2_uci
##   <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 activity      207.    6.94e-47      0.221      0.275      0.33
## 2 verbdist      205.    1.70e-46      0.215      0.273      0.334
## 3 sentlen.m     205.    2.17e-46      0.218      0.272      0.337
## 4 ari           204.    3.23e-46      0.212      0.271      0.325
## 5 gf           195.    2.68e-44      0.208      0.259      0.316
## 6 smog          190.    3.42e-43      0.197      0.252      0.31
## 7 NOUNcount.m   185.    3.41e-42      0.194      0.246      0.306
## 8 VERBfrac.m    179.    7.72e-41      0.183      0.238      0.299
## 9 fkg1         173.    1.40e-39      0.182      0.23       0.291
## 10 predorder.m  153.    3.50e-35      0.156      0.204      0.262
## 11 mamr         149.    2.90e-34      0.149      0.198      0.253
## 12 passives     118.    1.87e-27      0.109      0.157      0.21
## 13 predorder.v  115.    7.80e-27      0.109      0.153      0.209
## 14 literary     114.    1.44e-26      0.106      0.151      0.202
## 15 atl          107.    3.57e-25      0.1       0.143      0.195
## 16 predsubjdist.m 107.    5.16e-25      0.0984     0.142      0.19
## 17 compoundVERBs 99.6    1.83e-23      0.0869     0.133      0.181
## 18 NOUNcount.v   94.7    2.18e-22      0.0846     0.126      0.175
## 19 predsubjdist.v 83.6    5.96e-20      0.0716     0.111      0.164
## 20 subj         81.0    2.20e-19      0.0679     0.108      0.156
## 21 maentropy     79.7    4.28e-19      0.0649     0.106      0.152
```

```
## 22 predobjdist.m      77.9 1.07e-18      0.0661      0.104      0.152
## 23 VERBcomp           68.4 1.36e-16      0.0557      0.0909      0.131
## 24 mattr              67.9 1.70e-16      0.0553      0.0903      0.137
## 25 NEGcount.m         60.9 5.91e-15      0.0464      0.081       0.122
## 26 fre                 57.4 3.55e-14      0.0424      0.0763      0.113
## 27 predobjdist.v      55.7 8.58e-14      0.0403      0.074       0.115
## 28 compoundVERBsdi~   50.7 1.08e-12      0.0387      0.0674      0.11
## 29 compoundVERBsdi~   48.5 3.22e-12      0.0352      0.0646      0.104
## 30 cli                 46.6 8.51e-12      0.0327      0.062       0.0979
## 31 NEGfrac.m           42.9 5.68e-11      0.0309      0.0571      0.0936
## 32 entropy             42.4 7.56e-11      0.0296      0.0563      0.0954
## 33 NEGcount.v          35.4 2.62e- 9      0.0239      0.0471      0.0801
## 34 VERBfrac.v          28.8 8.05e- 8      0.0157      0.0383      0.0719
## 35 NOUNfrac.v          27.1 1.95e- 7      0.0152      0.036       0.0689
## 36 sentcount           25.0 5.87e- 7      0.0121      0.0332      0.0637
## 37 relativisticexp~   18.4 1.78e- 5      0.00828     0.0245      0.0493
## 38 sentlen.v           15.8 7.22e- 5      0.00497     0.0209      0.0489
## 39 wordcount           14.0 1.84e- 4      0.00386     0.0186      0.0444
## 40 verbalNOUNs         12.8 3.56e- 4      0.00287     0.017       0.0414
## 41 caserepcount.v      10.9 9.43e- 4      0.00234     0.0145      0.0382
## 42 obj                  10.8 1.03e- 3      0.00258     0.0143      0.0361
## 43 redundexprs         10.4 1.29e- 3      0.00351     0.0138      0.028
## 44 extrcaseexprs       10.3 1.34e- 3      0.00258     0.0137      0.0345
## 45 hapaxes              8.53 3.50e- 3      0.00135     0.0113      0.0321
## 46 hpoint               8.12 4.38e- 3      0.000932    0.0108      0.0282
## 47 abstractNOUNs        7.44 6.39e- 3      0.000641    0.00989     0.028
## 48 anaphoricrefs        6.67 9.79e- 3      0.00037     0.00887     0.0286
## 49 VERBcompdist.m       5.67 1.73e- 2      0.000255    0.00754     0.0246
## # i 1 more variable: mean_sign <dbl>
```

```
featcomp %>%
  mutate(signed_effect = kw_epsilon2 * mean_sign) %>%
  ggplot(aes(x = estimate, y = signed_effect, label = Variable)) +
  geom_line(alpha = 0.25) +
  geom_text(aes(color = kw_sel))
```