

# Housing Prices Regression

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```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr  0.3.4
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.4.1
## v readr   2.1.3      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(caret)
```

```
## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
##     lift
```

```
library(glmnet)
```

```
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
##
## The following objects are masked from 'package:tidyr':
##
##     expand, pack, unpack
##
## Loaded glmnet 4.1-6
```

## Import data

```
train<- read_csv("train.csv")
```

```
## Rows: 1460 Columns: 81
## -- Column specification -----
## Delimiter: ","
```

```
## chr (43): MSZoning, Street, Alley, LotShape, LandContour, Utilities, LotConf...
## dbl (38): Id, MSSubClass, LotFrontage, LotArea, OverallQual, OverallCond, Ye...
##
## 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.
```

```
test <- read_csv("test.csv")
```

```
## Rows: 1459 Columns: 80
## -- Column specification -----
## Delimiter: ","
## chr (43): MSZoning, Street, Alley, LotShape, LandContour, Utilities, LotConf...
## dbl (37): Id, MSSubClass, LotFrontage, LotArea, OverallQual, OverallCond, Ye...
##
## 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.
```

```
head(train)
```

```
## # A tibble: 6 x 81
##       Id MSSubClass MSZoning LotFr~1 LotArea Street Alley LotSh~2 LandC~3 Utili~4
##   <dbl>      <dbl> <chr>      <dbl>   <dbl> <chr>  <chr> <chr>   <chr>   <chr>
## 1     1         60 RL          65    8450 Pave  <NA>  Reg    Lvl     AllPub
## 2     2         20 RL          80    9600 Pave  <NA>  Reg    Lvl     AllPub
## 3     3         60 RL          68   11250 Pave  <NA>  IR1    Lvl     AllPub
## 4     4         70 RL          60    9550 Pave  <NA>  IR1    Lvl     AllPub
## 5     5         60 RL          84   14260 Pave  <NA>  IR1    Lvl     AllPub
## 6     6         50 RL          85   14115 Pave  <NA>  IR1    Lvl     AllPub
## # ... with 71 more variables: LotConfig <chr>, LandSlope <chr>,
## #   Neighborhood <chr>, Condition1 <chr>, Condition2 <chr>, BldgType <chr>,
## #   HouseStyle <chr>, OverallQual <dbl>, OverallCond <dbl>, YearBuilt <dbl>,
## #   YearRemodAdd <dbl>, RoofStyle <chr>, RoofMatl <chr>, Exterior1st <chr>,
## #   Exterior2nd <chr>, MasVnrType <chr>, MasVnrArea <dbl>, ExterQual <chr>,
## #   ExterCond <chr>, Foundation <chr>, BsmtQual <chr>, BsmtCond <chr>,
## #   BsmtExposure <chr>, BsmtFinType1 <chr>, BsmtFinSF1 <dbl>, ...
```

```
test$Id <- NULL
train$Id <- NULL
test$SalePrice <- NA
```

Remove id variable in both data and add SalePrice variable to test\_data

```
all <- rbind(train,test)
```

combine train and test data

```
missing_percentage <- function(df){
  colSums(is.na(df))/nrow(df)
}
missing_percentage(all)
```

##	MSSubClass	MSZoning	LotFrontage	LotArea	Street
##	0.0000000000	0.0013703323	0.1664953751	0.0000000000	0.0000000000
##	Alley	LotShape	LandContour	Utilities	LotConfig
##	0.9321685509	0.0000000000	0.0000000000	0.0006851662	0.0000000000
##	LandSlope	Neighborhood	Condition1	Condition2	BldgType
##	0.0000000000	0.0000000000	0.0000000000	0.0000000000	0.0000000000
##	HouseStyle	OverallQual	OverallCond	YearBuilt	YearRemodAdd
##	0.0000000000	0.0000000000	0.0000000000	0.0000000000	0.0000000000
##	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType
##	0.0000000000	0.0000000000	0.0003425831	0.0003425831	0.0082219938
##	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual
##	0.0078794108	0.0000000000	0.0000000000	0.0000000000	0.0277492292
##	BsmtCond	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2
##	0.0280918123	0.0280918123	0.0270640630	0.0003425831	0.0274066461
##	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC
##	0.0003425831	0.0003425831	0.0003425831	0.0000000000	0.0000000000
##	CentralAir	Electrical	1stFlrSF	2ndFlrSF	LowQualFinSF
##	0.0000000000	0.0003425831	0.0000000000	0.0000000000	0.0000000000
##	GrLivArea	BsmtFullBath	BsmtHalfBath	FullBath	HalfBath
##	0.0000000000	0.0006851662	0.0006851662	0.0000000000	0.0000000000
##	BedroomAbvGr	KitchenAbvGr	KitchenQual	TotRmsAbvGrd	Functional
##	0.0000000000	0.0000000000	0.0003425831	0.0000000000	0.0006851662
##	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	GarageFinish
##	0.0000000000	0.4864679685	0.0537855430	0.0544707091	0.0544707091
##	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive
##	0.0003425831	0.0003425831	0.0544707091	0.0544707091	0.0000000000
##	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch
##	0.0000000000	0.0000000000	0.0000000000	0.0000000000	0.0000000000
##	PoolArea	PoolQC	Fence	MiscFeature	MiscVal
##	0.0000000000	0.9965741692	0.8043850634	0.9640287770	0.0000000000
##	MoSold	YrSold	SaleType	SaleCondition	SalePrice
##	0.0000000000	0.0000000000	0.0003425831	0.0000000000	0.4998287085

```
all <- mutate_if(all,is.character,as.factor)
```

```
all <- all %>% mutate_if(is.factor, ~ ifelse(is.na(.), 0, .))
# replace missing values NA with 0 for all categorical variables
all <- all %>% mutate_if(is.numeric, ~ ifelse(is.na(.), mean(., na.rm = TRUE), .))
# replcae missing values NA with mean for all numeric variables
sum(is.na(all))
```

```
## [1] 0
```

```
# no missing values in the data anymore
```

```
# Split the data back to training and test sets
train_data <- all[1:nrow(train),]
test_data <- all[(nrow(train)+1):nrow(all),]
test_data$SalePrice <- NA
```

## Ridge regression

```
lambda <- 10^seq(-3, 3, length = 100)
```

```
# Build the model
set.seed(123)
ridge <- train(
  SalePrice ~., data = train_data, method = "glmnet",
  trControl = trainControl("cv", number = 10),
  tuneGrid = expand.grid(alpha = 0, lambda = lambda)
)
```

```
# Model coefficients
coef(ridge$finalModel, ridge$bestTune$lambda)
```

```
## 80 x 1 sparse Matrix of class "dgCMatrix"
##              s1
## (Intercept)  1.590079e+06
## MSSubClass   -7.574294e+01
## MSZoning     -1.844319e+03
## LotFrontage  -9.841805e+01
## LotArea       3.734902e-01
## Street       3.032689e+04
## Alley        -2.813798e+03
## LotShape     -9.757778e+02
## LandContour   2.073182e+03
## Utilities    -4.185069e+04
## LotConfig    -7.806494e+01
## LandSlope     4.862401e+03
## Neighborhood  2.660318e+02
## Condition1   -5.690022e+02
## Condition2   -8.223009e+03
## BldgType     -2.344855e+03
## HouseStyle   -6.583748e+02
## OverallQual   1.003744e+04
## OverallCond   4.432523e+03
## YearBuilt     1.357571e+02
## YearRemodAdd  7.911962e+01
## RoofStyle     1.852354e+03
## RoofMatl      4.723247e+03
## Exterior1st  -6.856764e+02
## Exterior2nd   2.502328e+02
## MasVnrType    3.615820e+03
## MasVnrArea    3.005154e+01
## ExterQual     -9.857949e+03
## ExterCond      6.589677e+02
## Foundation    8.568175e+02
## BsmtQual     -6.822881e+03
## BsmtCond      2.850036e+03
## BsmtExposure  -2.723548e+03
## BsmtFinType1  -4.268888e+02
## BsmtFinSF1    9.808956e+00
```

```
## BsmtFinType2    1.846611e+03
## BsmtFinSF2      1.331640e+01
## BsmtUnfSF       6.243585e-01
## TotalBsmtSF     1.299931e+01
## Heating         -1.747944e+03
## HeatingQC       -7.918285e+02
## CentralAir      2.534156e+03
## Electrical      -4.588454e+02
## '1stFlrSF'      1.577211e+01
## '2ndFlrSF'      1.807983e+01
## LowQualFinSF    -1.495945e+01
## GrLivArea       2.073197e+01
## BsmtFullBath    6.348293e+03
## BsmtHalfBath    3.517593e+02
## FullBath        4.834663e+03
## HalfBath        1.821559e+03
## BedroomAbvGr   -2.706703e+03
## KitchenAbvGr   -1.633139e+04
## KitchenQual     -8.110780e+03
## TotRmsAbvGrd    3.658318e+03
## Functional      3.531496e+03
## Fireplaces      7.304832e+03
## FireplaceQu     -7.982269e+02
## GarageType      1.609407e+02
## GarageYrBltd   -2.854904e+01
## GarageFinish    -2.211134e+03
## GarageCars      9.092250e+03
## GarageArea      1.238980e+01
## GarageQual      -1.298970e+03
## GarageCond      -1.108356e+02
## PavedDrive      2.735088e+03
## WoodDeckSF      2.080849e+01
## OpenPorchSF     -4.903567e+00
## EnclosedPorch   3.774804e+00
## '3SsnPorch'     2.456111e+01
## ScreenPorch     4.672345e+01
## PoolArea        2.883107e+02
## PoolQC          -9.073622e+04
## Fence           3.012748e+02
## MiscFeature     -1.666935e+03
## MiscVal         2.634597e-01
## MoSold          -9.728437e+01
## YrSold           -9.670621e+02
## SaleType        -5.210729e+02
## SaleCondition   2.832771e+03
```

```
# Make predictions
```

```
ridge_predictions <- ridge %>% predict(test_data)
ridge_predictions <- list(unname(ridge_predictions))[[1]]
head(ridge_predictions)
```

```
## [1] 112625.4 162275.5 172171.1 189024.2 190119.3 173952.5
```

## Lasso regression

```
# Build the model
set.seed(123)
lasso <- train(
  SalePrice ~., data = train_data, method = "glmnet",
  trControl = trainControl("cv", number = 10),
  tuneGrid = expand.grid(alpha = 1, lambda = lambda)
)
```

```
# Model coefficients
coef(lasso$finalModel, lasso$bestTune$lambda)
```

```
## 80 x 1 sparse Matrix of class "dgCMatrix"
##              s1
## (Intercept)  9.977898e+05
## MSSubClass   -7.683223e+01
## MSZoning     -2.043587e+03
## LotFrontage  -1.080784e+02
## LotArea       4.071036e-01
## Street       3.054043e+04
## Alley        -1.952668e+03
## LotShape     -6.947967e+02
## LandContour   1.580685e+03
## Utilities    -3.728149e+04
## LotConfig     .
## LandSlope     3.831658e+03
## Neighborhood  2.109665e+02
## Condition1   -4.698519e+02
## Condition2   -8.296996e+03
## BldgType     -2.555701e+03
## HouseStyle    -3.878252e+02
## OverallQual   1.072505e+04
## OverallCond   5.444117e+03
## YearBuilt     2.011788e+02
## YearRemodAdd  1.530859e+01
## RoofStyle     9.606591e+02
## RoofMatl      4.221183e+03
## Exterior1st  -7.280742e+02
## Exterior2nd   3.150269e+02
## MasVnrType    3.974552e+03
## MasVnrArea    3.139741e+01
## ExterQual     -1.004066e+04
## ExterCond     4.923942e+02
## Foundation    .
## BsmtQual      -7.120689e+03
## BsmtCond       2.546859e+03
## BsmtExposure  -2.567291e+03
## BsmtFinType1  -2.161994e+02
## BsmtFinSF1     1.049722e+01
## BsmtFinType2   1.749219e+03
## BsmtFinSF2     1.215882e+01
## BsmtUnfSF      .
```

```
## TotalBsmtSF      1.438338e+01
## Heating          -1.108243e+03
## HeatingQC        -4.386006e+02
## CentralAir       1.097620e+03
## Electrical       -2.356310e+02
## '1stFlrSF'       .
## '2ndFlrSF'       1.103718e+00
## LowQualFinSF     -4.180105e+01
## GrLivArea        4.215983e+01
## BsmtFullBath     5.654138e+03
## BsmtHalfBath     -2.169250e+02
## FullBath         1.283185e+03
## HalfBath         .
## BedroomAbvGr    -3.631023e+03
## KitchenAbvGr    -1.630775e+04
## KitchenQual      -7.622710e+03
## TotRmsAbvGrd    3.796544e+03
## Functional       3.474986e+03
## Fireplaces       6.945320e+03
## FireplaceQu     -9.284502e+02
## GarageType       5.402206e+01
## GarageYrBlt      .
## GarageFinish    -1.841193e+03
## GarageCars       1.005255e+04
## GarageArea       4.374141e+00
## GarageQual      -1.467178e+03
## GarageCond       .
## PavedDrive       2.033219e+03
## WoodDeckSF       2.009377e+01
## OpenPorchSF     -2.662077e+00
## EnclosedPorch   -9.375567e-01
## '3SsnPorch'     2.176285e+01
## ScreenPorch      4.588821e+01
## PoolArea        6.689632e+02
## PoolQC          -1.894544e+05
## Fence           .
## MiscFeature     -1.247885e+03
## MiscVal         .
## MoSold          -3.487299e+01
## YrSold          -7.049526e+02
## SaleType        -4.622954e+02
## SaleCondition   3.383327e+03
```

```
# Make predictions
```

```
lasso_predictions <- lasso %>% predict(test_data)
lasso_predictions <- list(unname(lasso_predictions))[[1]]
head(lasso_predictions)
```

```
## [1] 113048.1 160854.5 169713.7 187489.0 189715.1 172390.3
```

## elastic net regression

```
# Build the model using the training set
set.seed(123)
elastic <- train(
  SalePrice ~., data = train_data, method = "glmnet",
  trControl = trainControl("cv", number = 10),
  tuneLength = 10
)
```

```
# Model coefficients
coef(elastic$finalModel, elastic$bestTune$lambda)
```

```
## 80 x 1 sparse Matrix of class "dgCMatrix"
##               s1
## (Intercept)  -2.222545e+04
## MSSubClass    -6.017465e+01
## MSZoning      -1.147227e+02
## LotFrontage    .
## LotArea       3.086747e-01
## Street        1.485696e+04
## Alley         -1.669626e+02
## LotShape      -8.007844e+02
## LandContour    5.167040e+02
## Utilities     -1.200510e+02
## LotConfig      .
## LandSlope      1.805793e+03
## Neighborhood  1.167375e+02
## Condition1     .
## Condition2    -3.122362e+03
## BldgType       -1.504731e+03
## HouseStyle      .
## OverallQual    1.083895e+04
## OverallCond    2.791447e+03
## YearBuilt      1.040328e+02
## YearRemodAdd   1.092165e+02
## RoofStyle      1.181253e+03
## RoofMatl       3.916179e+03
## Exterior1st    -1.144936e+02
## Exterior2nd     .
## MasVnrType     2.009820e+03
## MasVnrArea     2.466299e+01
## ExterQual      -9.872008e+03
## ExterCond       .
## Foundation      .
## BsmtQual       -5.811176e+03
## BsmtCond       2.137000e+03
## BsmtExposure   -1.667971e+03
## BsmtFinType1    .
## BsmtFinSF1     9.999052e+00
## BsmtFinType2    .
## BsmtFinSF2     8.251030e-02
## BsmtUnfSF       .
```



```
## TotalBsmtSF      1.419577e+01
## Heating          .
## HeatingQC        -6.978261e+02
## CentralAir        2.879013e+03
## Electrical        .
## '1stFlrSF'        1.222391e+01
## '2ndFlrSF'        1.121039e+01
## LowQualFinSF      .
## GrLivArea         2.288009e+01
## BsmtFullBath       5.441548e+03
## BsmtHalfBath       .
## FullBath           3.709265e+03
## HalfBath           1.472197e+03
## BedroomAbvGr      .
## KitchenAbvGr      -1.312546e+04
## KitchenQual        -8.632634e+03
## TotRmsAbvGrd       2.473421e+03
## Functional         2.736324e+03
## Fireplaces         5.938355e+03
## FireplaceQu        .
## GarageType         .
## GarageYrBlt        .
## GarageFinish       -1.747602e+03
## GarageCars         8.321729e+03
## GarageArea         1.196946e+01
## GarageQual         .
## GarageCond         .
## PavedDrive         1.646628e+03
## WoodDeckSF         1.938248e+01
## OpenPorchSF        .
## EnclosedPorch      .
## '3SsnPorch'        .
## ScreenPorch        3.195121e+01
## PoolArea           7.667022e+01
## PoolQC             -3.737457e+04
## Fence              .
## MiscFeature        -1.342096e+01
## MiscVal            .
## MoSold             .
## YrSold             -1.992799e+02
## SaleType           .
## SaleCondition      1.765328e+03
```

```
# Make predictions
```

```
elastic_predictions <- elastic %>% predict(test_data)
elastic_predictions <- list(unname(elastic_predictions))[[1]]
head(elastic_predictions)
```

```
## [1] 113815.0 164734.5 175006.6 192678.4 192390.1 175869.5
```

## Comparing models

```
models <- list(ridge = ridge, lasso = lasso, elastic = elastic)
resamples(models) %>% summary( metric = "RMSE")
```

```
##
## Call:
## summary.resamples(object = ., metric = "RMSE")
##
## Models: ridge, lasso, elastic
## Number of resamples: 10
##
## RMSE
##           Min.   1st Qu.   Median     Mean   3rd Qu.     Max. NA's
## ridge   25043.56 26561.72 31352.83 33808.79 36573.45 61047.36    0
## lasso   24407.04 26443.11 32237.38 34461.08 37888.28 60703.50    0
## elastic 25732.45 26415.56 30501.21 33821.52 36788.75 61461.48    0
```

```
submission<-read_csv("sample_submission.csv")
```

```
## Rows: 1459 Columns: 2
## -- Column specification -----
## Delimiter: ","
## dbl (2): Id, SalePrice
##
## 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.
```

```
submission<-mutate(submission, SalePrice=elastic_predictions)
write.csv(submission, file = "submission.csv",row.names = FALSE)
```