Nomad2018 Predicting Transparent Conductors

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Load the libraries

```
library(caret)
## Loading required package: lattice
## Loading required package: ggplot2
library(ggplot2)
library(dplyr)
##
## Attaching package: 'dplyr'
  The following objects are masked from 'package:stats':
##
##
       filter, lag
  The following objects are masked from 'package:base':
##
##
##
       intersect, setdiff, setequal, union
library(randomForest)
## randomForest 4.6-12
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:dplyr':
##
##
       combine
```

```
## The following object is masked from 'package:ggplot2':
##
## margin
```

Load the Data

```
Training_Data<-read.csv("C:/Users/ddddd/Comptitions/Kaggle/Nomad2018 Predicting/train.csv", stri
ngsAsFactors = F)
Test_Data<-read.csv("C:/Users/ddddd/Comptitions/Kaggle/Nomad2018 Predicting/test.csv", stringsAs
Factors = F)</pre>
```

Data Exploration

```
names(Training_Data)
```

```
str(Training_Data)
```

```
## 'data.frame':
                   2400 obs. of 14 variables:
## $ id
                               : int 1 2 3 4 5 6 7 8 9 10 ...
                               : int 33 194 227 167 194 227 206 12 206 194 ...
  $ spacegroup
   $ number of total atoms
                                      80 80 40 30 80 40 80 20 80 80 ...
  $ percent atom al
                               : num
                                      0.625 0.625 0.812 0.75 0 ...
   $ percent_atom_ga
                               : num
                                      0.375 0.375 0.188 0 0.625 ...
##
  $ percent_atom_in
                               : num 0 0 0 0.25 0.375 0 0.875 0.5 0.25 0 ...
  $ lattice_vector_1_ang
                               : num 9.95 6.18 9.75 5 6.66 ...
                               : num 8.55 6.18 5.66 5 6.66 ...
## $ lattice vector 2 ang
   $ lattice_vector_3_ang
                               : num 9.18 23.63 13.96 13.53 24.58 ...
   $ lattice angle alpha degree: num 90 90 91 90 90 ...
   $ lattice angle beta degree : num 90 90 91.1 90 90 ...
   $ lattice angle gamma degree: num 90 120 30.5 120 120 ...
   $ formation_energy_ev_natom : num   0.068   0.249   0.1821   0.2172   0.0505   ...
                               : num 3.44 2.92 2.74 3.35 1.38 ...
   $ bandgap energy ev
```

```
Training_Data$id <- as.numeric(Training_Data$id )
Training_Data$spacegroup <- as.numeric(Training_Data$spacegroup )</pre>
```

```
str(Training_Data)
```

```
## 'data.frame':
                  2400 obs. of 14 variables:
   $ id
                              : num
                                    1 2 3 4 5 6 7 8 9 10 ...
##
  $ spacegroup
                              : num 33 194 227 167 194 227 206 12 206 194 ...
   $ number of total atoms
##
                                    80 80 40 30 80 40 80 20 80 80 ...
  $ percent atom al
                              : num 0.625 0.625 0.812 0.75 0 ...
   $ percent atom ga
                                     0.375 0.375 0.188 0 0.625 ...
##
                              : num
##
  $ percent_atom_in
                             : num 0 0 0 0.25 0.375 0 0.875 0.5 0.25 0 ...
   $ lattice_vector_1_ang
                             : num 9.95 6.18 9.75 5 6.66 ...
##
  $ lattice_vector_2_ang : num 8.55 6.18 5.00 5 0.00 ...

d lattice_vector 3_ang : num 9.18 23.63 13.96 13.53 24.58 ...
##
   $ lattice angle alpha degree: num 90 90 91 90 90 ...
   $ lattice angle beta degree : num 90 90 91.1 90 90 ...
   $ lattice angle gamma degree: num 90 120 30.5 120 120 ...
   $ bandgap energy ev
                              : num 3.44 2.92 2.74 3.35 1.38 ...
```

Research question

We are already given a research question by this competition on Kaggle itself.

The prediction of two target properties: the formation energy (which is an indication of the stability of a new material) and the bandgap energy (which is an indication of the potential for transparency over the visible range) to facilitate the discovery of new transparent conductors and allow for advancements in the above-mentioned technologies.

For each id in the test set, we must predict a value for both formation energy ev natom and bandgap energy ev.

Solution Starts from here...

Data Preprocessing

```
Training_Data %>%
  filter( id != "NA", spacegroup != "NA", number_of_total_atoms != "NA", percent_atom_al != "NA",
  percent_atom_ga != "NA", percent_atom_in != "NA", lattice_vector_1_ang != "NA", lattice_vector
  _2_ang != "NA", lattice_vector_3_ang != "NA", lattice_angle_alpha_degree != "NA", lattice_angle_
  beta_degree != "NA", lattice_angle_gamma_degree != "NA", formation_energy_ev_natom != "NA", band
  gap_energy_ev != "NA") %>%
    select( id, spacegroup, number_of_total_atoms, percent_atom_al, percent_atom_ga, percent_atom_
    in    , lattice_vector_1_ang, lattice_vector_2_ang, lattice_vector_3_ang, lattice_angle_alpha_deg
  ree, lattice_angle_beta_degree, lattice_angle_gamma_degree, formation_energy_ev_natom, bandgap_
  energy_ev ) %>%
    str()
```

```
## 'data.frame':
                  2400 obs. of 14 variables:
   $ id
                             : num 1 2 3 4 5 6 7 8 9 10 ...
##
##
   $ spacegroup
                             : num 33 194 227 167 194 227 206 12 206 194 ...
                                   80 80 40 30 80 40 80 20 80 80 ...
   $ number of total atoms
##
                             : num
##
  $ percent atom al
                                   0.625 0.625 0.812 0.75 0 ...
                            : num
   $ percent atom ga
                                   0.375 0.375 0.188 0 0.625 ...
##
                            : num
##
  $ percent atom in
                            : num 0 0 0 0.25 0.375 0 0.875 0.5 0.25 0 ...
                            : num 9.95 6.18 9.75 5 6.66 ...
##
   $ lattice vector 1 ang
## $ lattice vector 2 ang
                             : num 8.55 6.18 5.66 5 6.66 ...
  $ lattice_vector_3_ang
                            : num 9.18 23.63 13.96 13.53 24.58 ...
##
  $ lattice_angle_alpha_degree: num 90 90 91 90 90 ...
   $ lattice angle beta degree : num 90 90 91.1 90 90 ...
   $ lattice angle gamma degree: num 90 120 30.5 120 120 ...
   $ bandgap_energy_ev
                             : num 3.44 2.92 2.74 3.35 1.38 ...
##
```

Training Data %>%

filter(id != "NA", spacegroup != "NA", number_of_total_atoms != "NA", percent_atom_al != "NA"
, percent_atom_ga != "NA", percent_atom_in != "NA", lattice_vector_1_ang != "NA", lattice_vector
_2_ang != "NA", lattice_vector_3_ang != "NA", lattice_angle_alpha_degree != "NA", lattice_angle_
beta_degree != "NA", lattice_angle_gamma_degree != "NA", formation_energy_ev_natom != "NA", band
gap_energy_ev != "NA") %>%

select(id, spacegroup, number_of_total_atoms, percent_atom_al, percent_atom_ga, percent_atom_
in , lattice_vector_1_ang, lattice_vector_2_ang, lattice_vector_3_ang, lattice_angle_alpha_deg
ree, lattice_angle_beta_degree, lattice_angle_gamma_degree, formation_energy_ev_natom, bandgap_
energy_ev) %>%

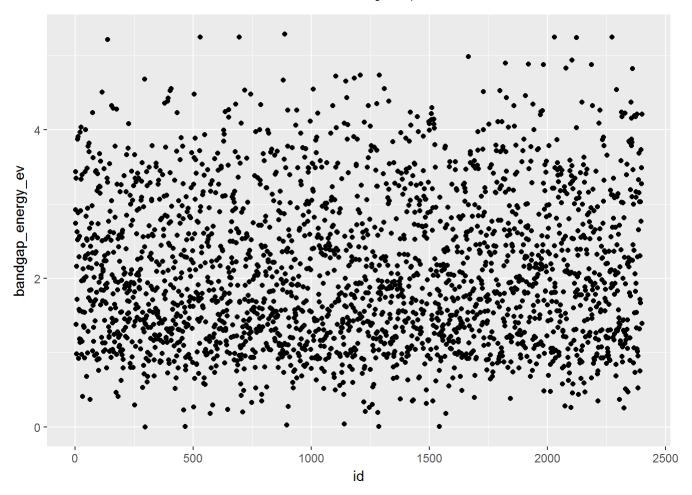
group_by(id, spacegroup, number_of_total_atoms, percent_atom_al, percent_atom_ga, percent_atom _in , lattice_vector_1_ang, lattice_vector_2_ang, lattice_vector_3_ang, lattice_angle_alpha_de gree, lattice_angle_beta_degree, lattice_angle_gamma_degree, formation_energy_ev_natom, bandgap _energy_ev)

```
## # A tibble: 2,400 x 14
               id, spacegroup, number of total atoms, percent atom al,
## # Groups:
## #
       percent atom ga, percent atom in, lattice vector 1 ang,
## #
       lattice vector 2 ang, lattice vector 3 ang,
## #
       lattice_angle_alpha_degree, lattice_angle_beta_degree,
       lattice_angle_gamma_degree, formation_energy_ev_natom,
## #
## #
       bandgap_energy_ev [2,400]
##
         id spacegroup number of total atoms percent atom al percent atom ga
                  <dbl>
##
      <dbl>
                                         <dbl>
                                                          <dbl>
                                                                          <dbl>
   1
          1
                                                         0.6250
##
                     33
                                            80
                                                                         0.3750
   2
          2
                   194
##
                                            80
                                                         0.6250
                                                                         0.3750
   3
          3
##
                    227
                                            40
                                                         0.8125
                                                                         0.1875
##
          4
                    167
                                            30
                                                         0.7500
                                                                         0.0000
##
                   194
                                                         0.0000
                                                                         0.6250
          6
                   227
##
                                            40
                                                         0.5625
                                                                         0.4375
##
          7
                    206
                                            80
                                                         0.0312
                                                                         0.0938
##
          8
                    12
                                            20
                                                         0.5000
                                                                         0.0000
##
   9
          9
                    206
                                            80
                                                         0.5312
                                                                         0.2188
         10
                   194
                                            80
                                                         0.4062
                                                                         0.5938
## 10
     ... with 2,390 more rows, and 9 more variables: percent atom in <dbl>,
       lattice_vector_1_ang <dbl>, lattice_vector_2_ang <dbl>,
## #
       lattice_vector_3_ang <dbl>, lattice_angle_alpha_degree <dbl>,
## #
## #
       lattice_angle_beta_degree <dbl>, lattice_angle_gamma_degree <dbl>,
## #
       formation energy ev natom <dbl>, bandgap energy ev <dbl>
```

Scatter plots to see the correlation between variable of x axis and y axis

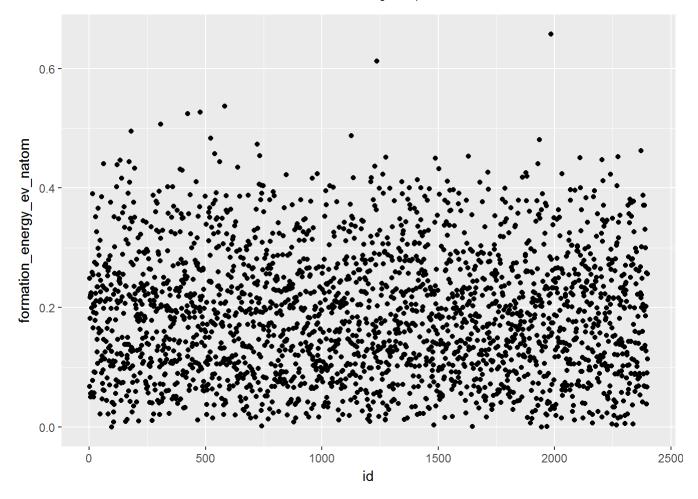
Plot for bandgap energy versus id

```
qplot(id, bandgap_energy_ev,data=Training_Data)
```



Plot for Formation Energy versus id

qplot(id, formation_energy_ev_natom,data=Training_Data)



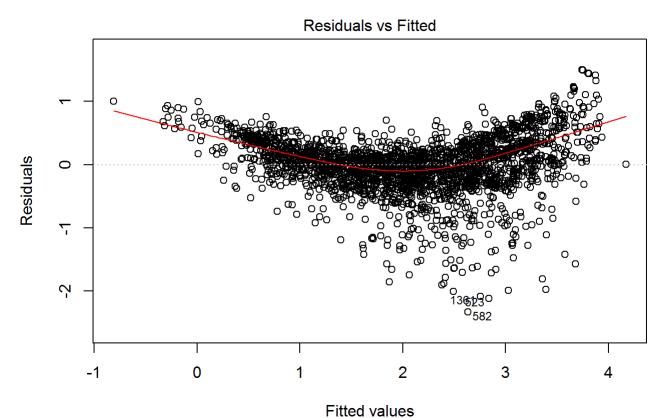
Modelling

This is for prediction of two target properties: the formation energy (which is an indication of the stability of a new material) and the bandgap energy (which is an indication of the potential for transparency over the visible range)

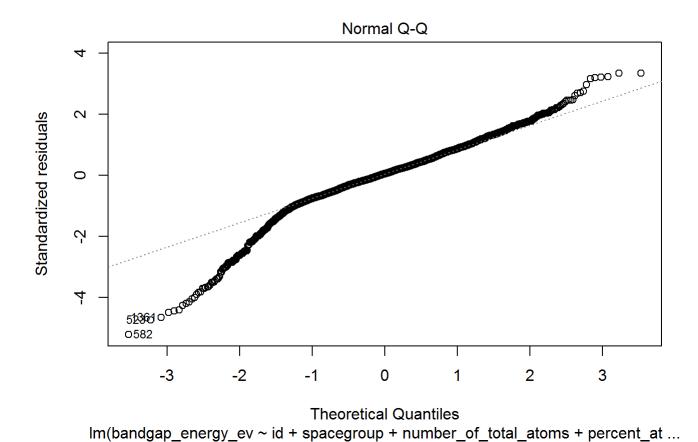
Model for bandgape energy

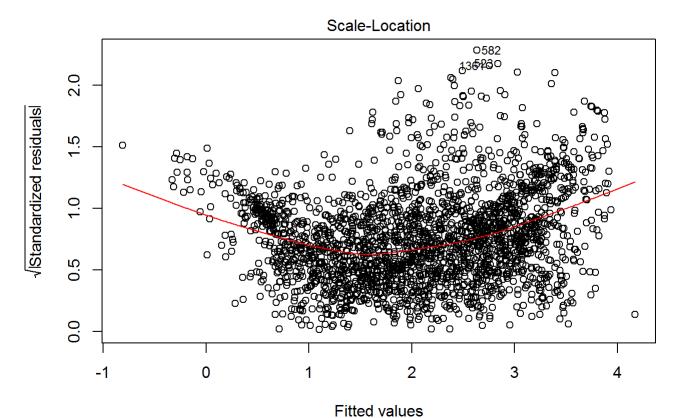
```
##
## Call:
## lm(formula = bandgap energy ev ~ id + spacegroup + number of total atoms +
##
       percent atom al + percent atom ga + percent atom in + lattice vector 1 ang +
##
       lattice_vector_2_ang + lattice_vector_3_ang + lattice_angle_alpha_degree +
##
       lattice angle beta degree + lattice angle gamma degree, data = Training Data)
##
## Residuals:
       Min
##
                      Median
                                   3Q
                 1Q
                                           Max
  -2.33608 -0.22049 0.02426 0.26132 1.50190
##
##
## Coefficients:
##
                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                              5.484e+02 2.227e+02 2.462 0.01388 *
                              3.762e-05 1.324e-05
                                                     2.841 0.00454 **
## id
## spacegroup
                             -3.123e-04 1.601e-04 -1.951 0.05119 .
## number of total atoms
                              1.280e-02 1.582e-03 8.091 9.30e-16 ***
## percent atom al
                             -5.466e+02 2.227e+02 -2.454 0.01419 *
                                                           0.01392 *
## percent atom ga
                             -5.481e+02 2.227e+02 -2.461
                             -5.499e+02 2.227e+02 -2.469
## percent atom in
                                                           0.01362 *
                             -5.580e-02 7.279e-03 -7.666 2.57e-14 ***
## lattice vector 1 ang
## lattice vector 2 ang
                             -1.379e-01 1.897e-02 -7.266 4.98e-13 ***
## lattice_vector_3_ang
                             -9.701e-02 4.841e-03 -20.041 < 2e-16 ***
                                                    6.014 2.09e-09 ***
## lattice angle alpha degree 5.855e-02 9.736e-03
## lattice_angle_beta_degree -2.374e-02 5.799e-03 -4.094 4.39e-05 ***
## lattice angle gamma degree 6.942e-03 5.944e-04 11.679 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4486 on 2387 degrees of freedom
## Multiple R-squared: 0.8025, Adjusted R-squared: 0.8015
## F-statistic: 808.1 on 12 and 2387 DF, p-value: < 2.2e-16
```

```
plot(bandgap_energy_Model)
```

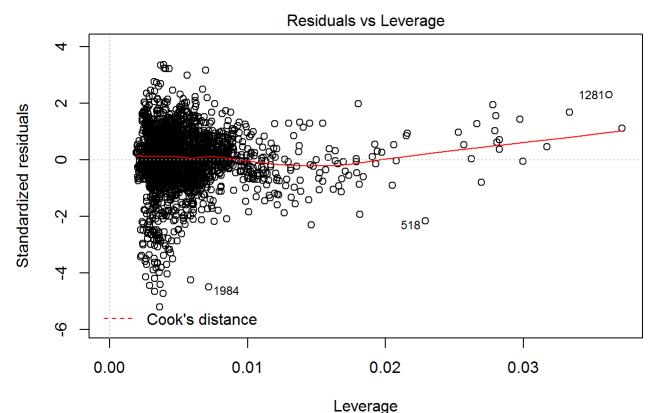


Im(bandgap_energy_ev ~ id + spacegroup + number_of_total_atoms + percent_at ...





Im(bandgap_energy_ev ~ id + spacegroup + number_of_total_atoms + percent_at ...

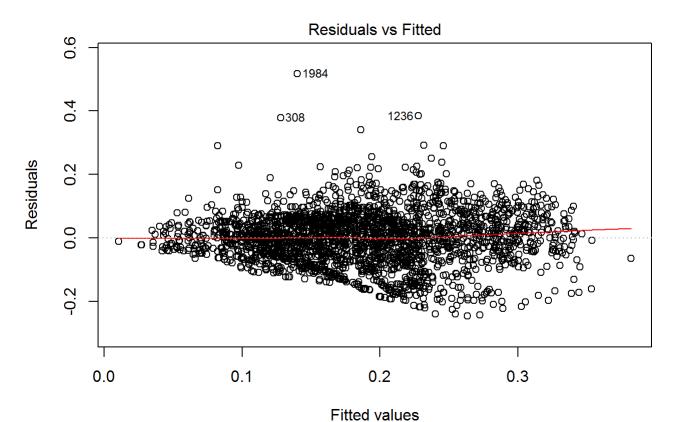


Im(bandgap_energy_ev ~ id + spacegroup + number_of_total_atoms + percent_at ...

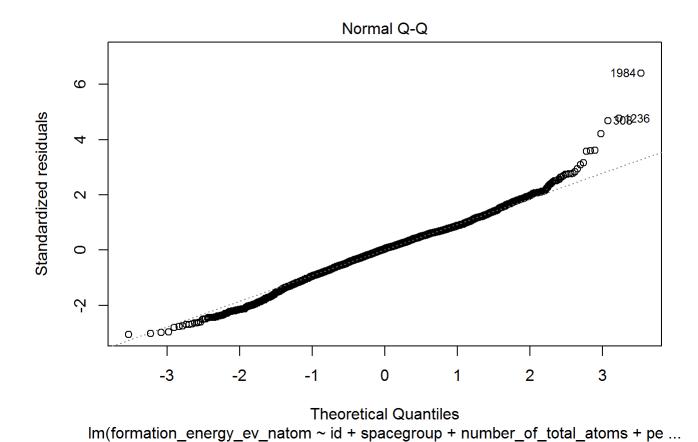
Model for formation energy

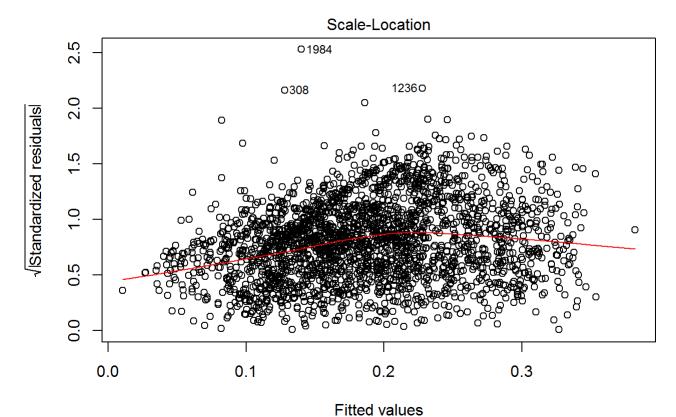
```
##
## Call:
## lm(formula = formation energy ev natom ~ id + spacegroup + number of total atoms +
##
       percent_atom_al + percent_atom_ga + percent_atom_in + lattice_vector_1_ang +
##
       lattice vector 2 ang + lattice vector 3 ang + lattice angle alpha degree +
##
       lattice_angle_beta_degree + lattice_angle_gamma_degree, data = Training_Data)
##
## Residuals:
##
        Min
                 1Q
                      Median
                                   3Q
                                           Max
##
  -0.24675 -0.04983 0.00377 0.05110 0.51706
##
## Coefficients:
##
                               Estimate Std. Error t value Pr(>|t|)
                             -1.560e+02 4.020e+01 -3.879 0.000108 ***
## (Intercept)
## id
                             -4.912e-06 2.391e-06 -2.055 0.040033 *
## spacegroup
                              6.404e-05 2.890e-05
                                                     2.216 0.026780 *
                             -7.824e-04 2.857e-04 -2.739 0.006208 **
## number of total atoms
## percent atom al
                              1.565e+02 4.020e+01
                                                     3.892 0.000102 ***
## percent atom ga
                              1.563e+02 4.020e+01 3.889 0.000104 ***
## percent_atom_in
                              1.565e+02 4.020e+01 3.893 0.000102 ***
## lattice vector 1 ang
                              4.266e-03 1.314e-03 3.246 0.001186 **
## lattice_vector_2_ang
                              2.498e-03 3.425e-03 0.729 0.465858
## lattice vector 3 ang
                              1.126e-02 8.739e-04 12.882 < 2e-16 ***
## lattice_angle_alpha_degree -3.040e-03 1.758e-03 -1.730 0.083791 .
## lattice angle beta degree -1.200e-03 1.047e-03 -1.147 0.251592
## lattice angle gamma degree -6.706e-04 1.073e-04 -6.249 4.87e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.08099 on 2387 degrees of freedom
## Multiple R-squared: 0.3978, Adjusted R-squared: 0.3948
## F-statistic: 131.4 on 12 and 2387 DF, p-value: < 2.2e-16
```

```
plot(formation_energy_Model)
```

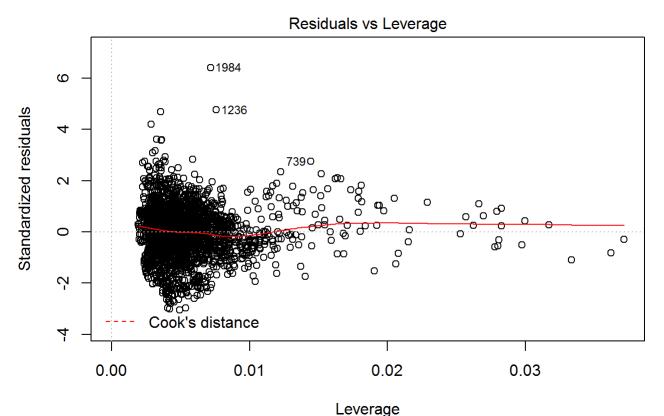


Im(formation_energy_ev_natom ~ id + spacegroup + number_of_total_atoms + pe ...





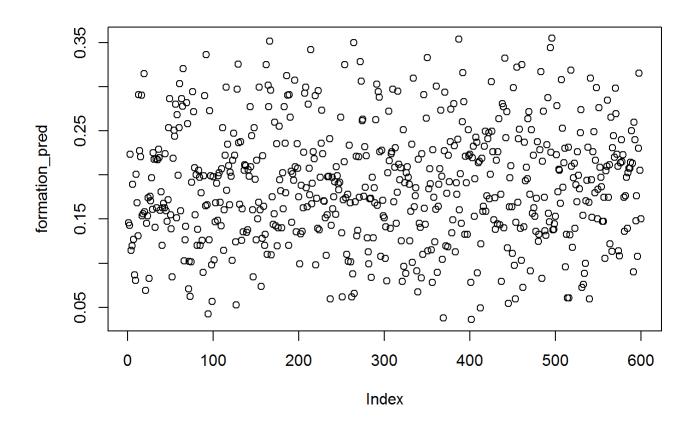
Im(formation_energy_ev_natom ~ id + spacegroup + number_of_total_atoms + pe ...



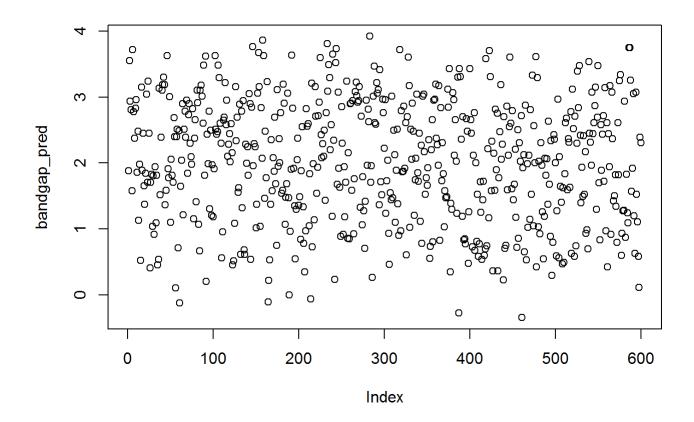
Im(formation_energy_ev_natom ~ id + spacegroup + number_of_total_atoms + pe ...

Predictions on Test Data

formation_pred <- predict(formation_energy_Model, Test_Data)
plot(formation pred)</pre>



bandgap_pred <- predict(bandgap_energy_Model, Test_Data)
plot(bandgap_pred)</pre>



Data for submission

```
Predicted_Outcome <- data.frame(id = 1:600, formation_energy_ev_natom = formation_pred, bandgap_
energy_ev = bandgap_pred)
```

```
colnames(Predicted_Outcome) <- c("id","formation_energy_ev_natom","bandgap_energy_ev")
write.csv(Predicted_Outcome,"Predicted_Outcome.csv",row.names = FALSE)</pre>
```

This Prediction helps to avoid costly and inefficient trial-and-error of synthetic methods.