

Analysis and Prediction of Energy Performance on Residential Buildings

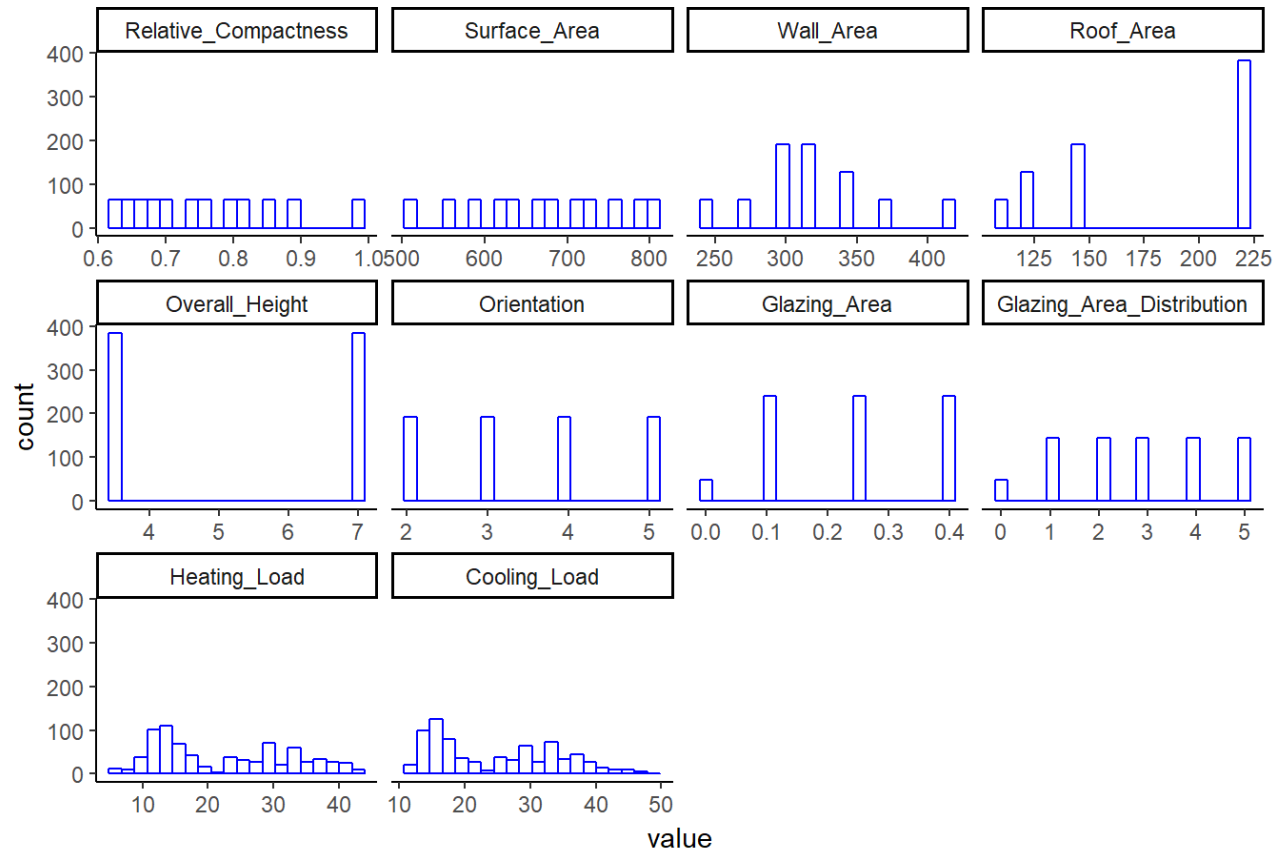


Tajudeen Abdulazeez and Greg Miller

SYRACUSE UNIVERSITY
School of Information Studies

Data Collected

Histograms distribution

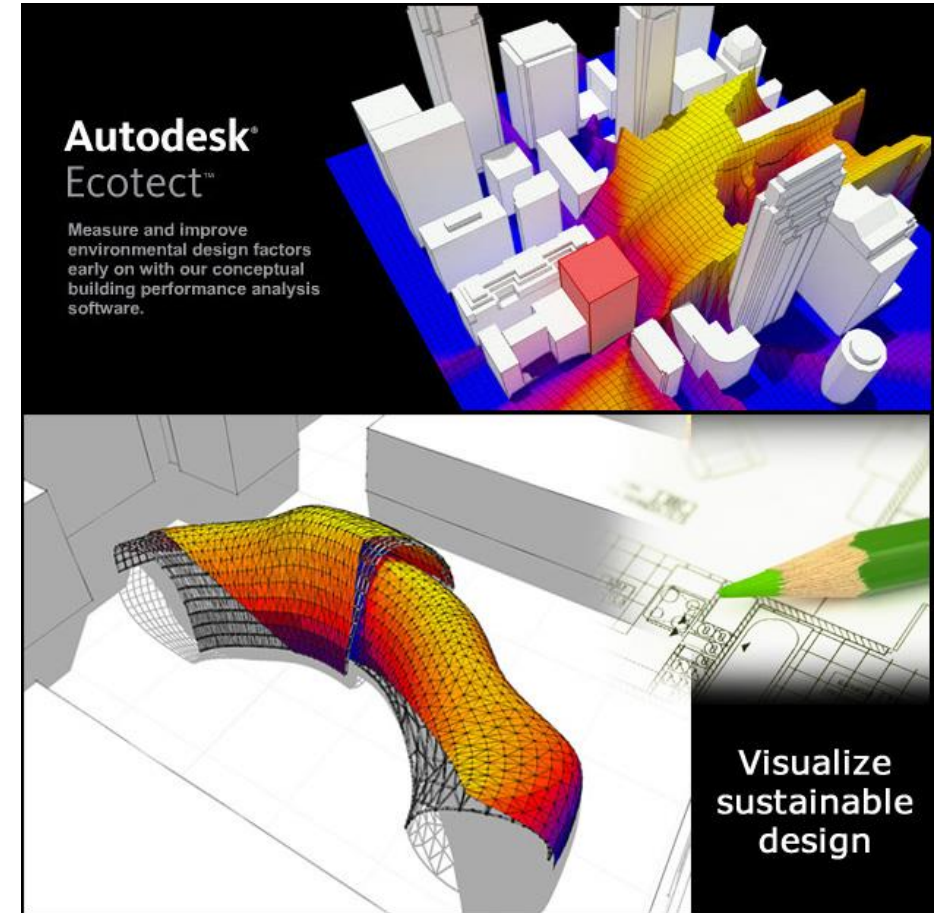


Input/Output Variable	Possible Values
Relative Compactness	12
Surface Area	12
Wall Area	7
Roof Area	4
Overall Height	2
Orientation	4
Glazing Area	4
Glazing Area Distribution	6
Heating Load	586
Cooling	636

<https://archive.ics.uci.edu/ml/datasets/Energy+efficiency>

Scope

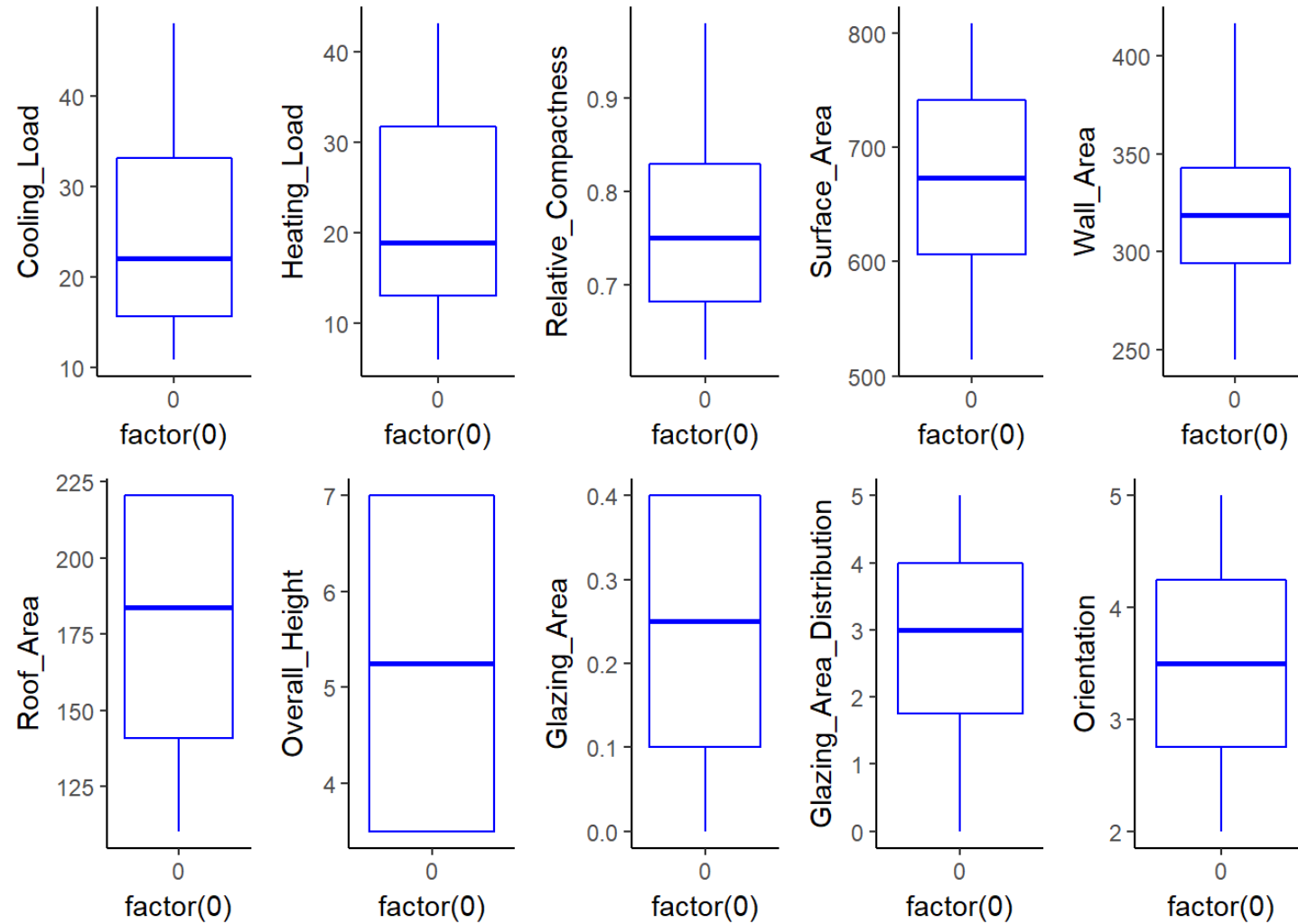
- Ecotect is a simulation tool for designing buildings
- Heating and Cooling Loads indicate the energy efficiency of the building
- Ecotect simulates these values through a difficult interface
- Create a model to better understand input variables and attempt a model that would predict the Heating and Cooling Loads



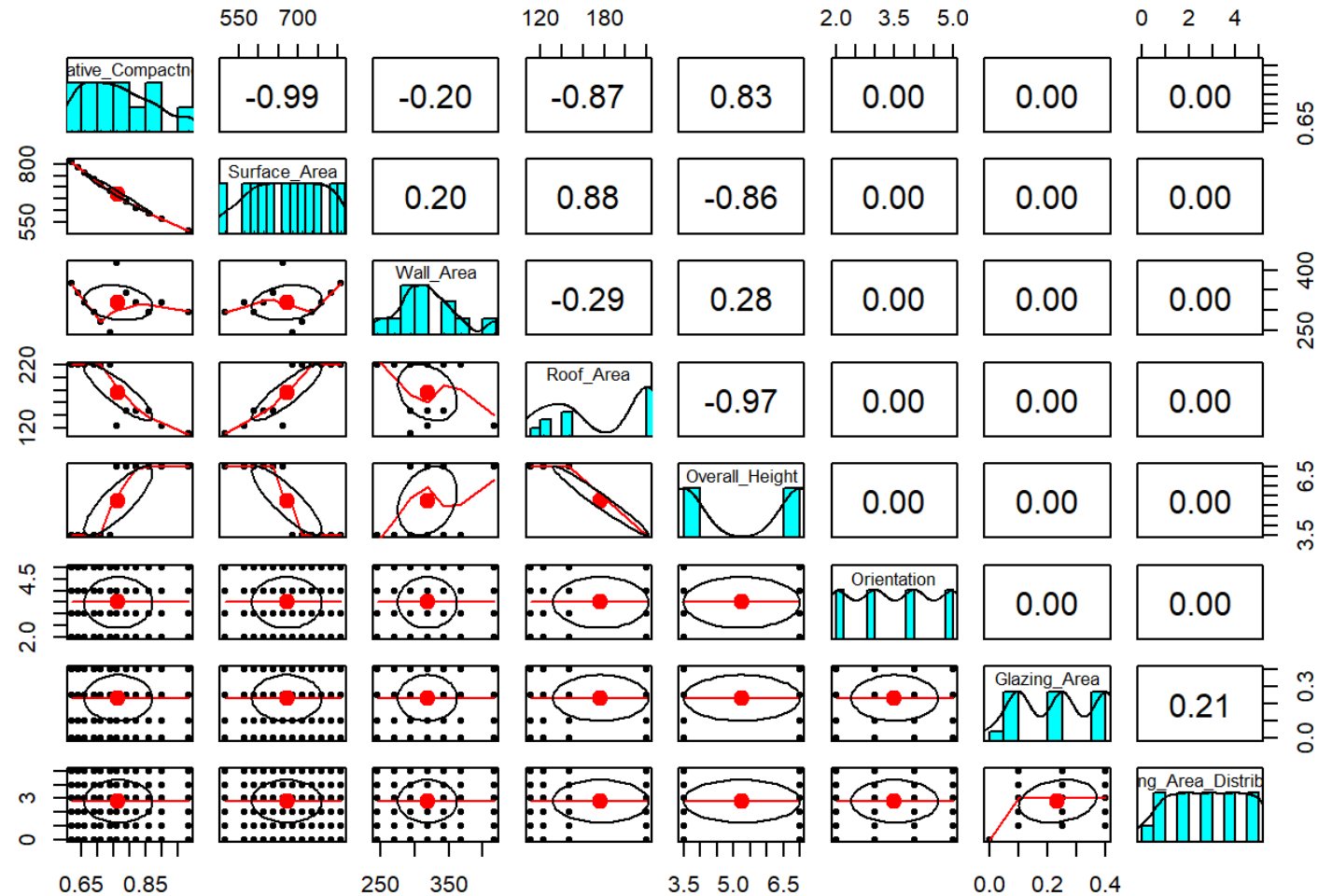
Summary

```
## Relative_Compactness Surface_Area Wall_Area Roof_Area
## Min. :0.6200 Min. :514.5 Min. :245.0 Min. :110.2
## 1st Qu.:0.6825 1st Qu.:606.4 1st Qu.:294.0 1st Qu.:140.9
## Median :0.7500 Median :673.8 Median :318.5 Median :183.8
## Mean :0.7642 Mean :671.7 Mean :318.5 Mean :176.6
## 3rd Qu.:0.8300 3rd Qu.:741.1 3rd Qu.:343.0 3rd Qu.:220.5
## Max. :0.9800 Max. :808.5 Max. :416.5 Max. :220.5
## Overall_Height Orientation Glazing_Area Glazing_Area_Distribution
## Min. :3.50 Min. :2.00 Min. :0.0000 Min. :0.000
## 1st Qu.:3.50 1st Qu.:2.75 1st Qu.:0.1000 1st Qu.:1.750
## Median :5.25 Median :3.50 Median :0.2500 Median :3.000
## Mean :5.25 Mean :3.50 Mean :0.2344 Mean :2.812
## 3rd Qu.:7.00 3rd Qu.:4.25 3rd Qu.:0.4000 3rd Qu.:4.000
## Max. :7.00 Max. :5.00 Max. :0.4000 Max. :5.000
## Heating_Load Cooling_Load
## Min. : 6.01 Min. :10.90
## 1st Qu.:12.99 1st Qu.:15.62
## Median :18.95 Median :22.08
## Mean :22.31 Mean :24.59
## 3rd Qu.:31.67 3rd Qu.:33.13
## Max. :43.10 Max. :48.03
```

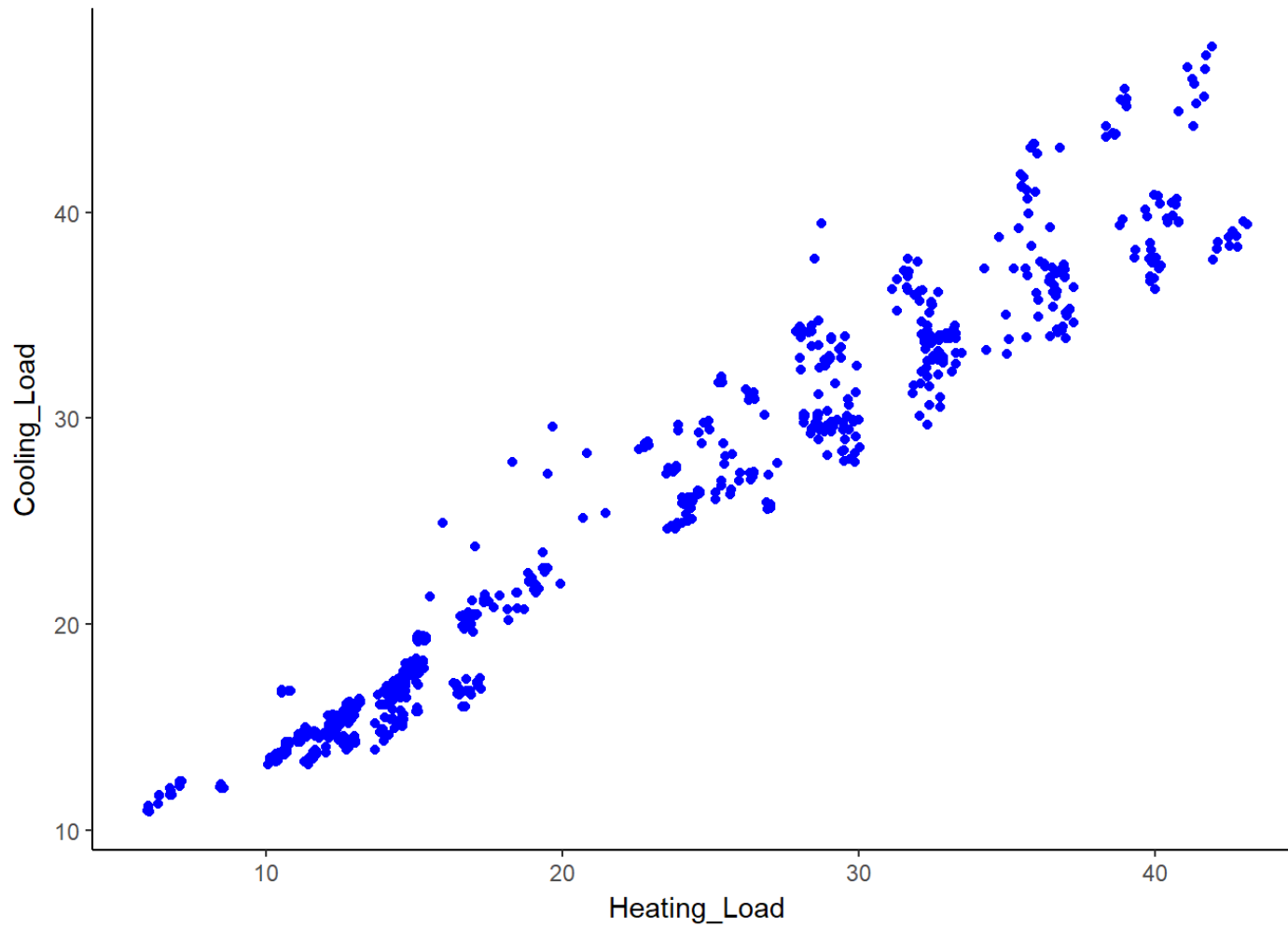
Data Distribution



Input Correlation



Output Correlation



Linear Model

```
##
## Call:
## lm(formula = Heating_Load ~ ., data = Mh)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.3862  -1.3667  -0.0142   1.3162   7.5555
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    84.386471   19.111765   4.415 1.15e-05 ***
## Relative_Compactness -64.773432  10.333611  -6.268 6.11e-10 ***
## Surface_Area     -0.087289   0.017149  -5.090 4.51e-07 ***
## Wall_Area         0.060813   0.006676   9.109 < 2e-16 ***
## Roof_Area         NA          NA         NA      NA
## Overall_Height     4.169954   0.339441  12.285 < 2e-16 ***
## Glazing_Area      20.437968   0.798727  25.588 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.947 on 762 degrees of freedom
## Multiple R-squared:  0.9153, Adjusted R-squared:  0.9147
## F-statistic: 1646 on 5 and 762 DF,  p-value: < 2.2e-16
```

```
##
## Call:
## lm(formula = Cooling_Load ~ ., data = Mc)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
##  -8.7240  -1.6017  -0.2631   1.3417  11.3251
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    97.761848   20.756339   4.710 2.94e-06 ***
## Relative_Compactness -70.787707  11.222822  -6.307 4.80e-10 ***
## Surface_Area     -0.088245   0.018624  -4.738 2.57e-06 ***
## Wall_Area         0.044682   0.007251   6.162 1.16e-09 ***
## Roof_Area         NA          NA         NA      NA
## Overall_Height     4.283843   0.368650  11.620 < 2e-16 ***
## Glazing_Area      14.817971   0.867458  17.082 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.2 on 762 degrees of freedom
## Multiple R-squared:  0.8876, Adjusted R-squared:  0.8868
## F-statistic: 1203 on 5 and 762 DF,  p-value: < 2.2e-16
```


Heating Load Random Forest

```
#heating load

rf_heating <- randomForest(Heating_Load ~ ., data = M1, mtry=3, importance = TRUE, na.action = na.omit)

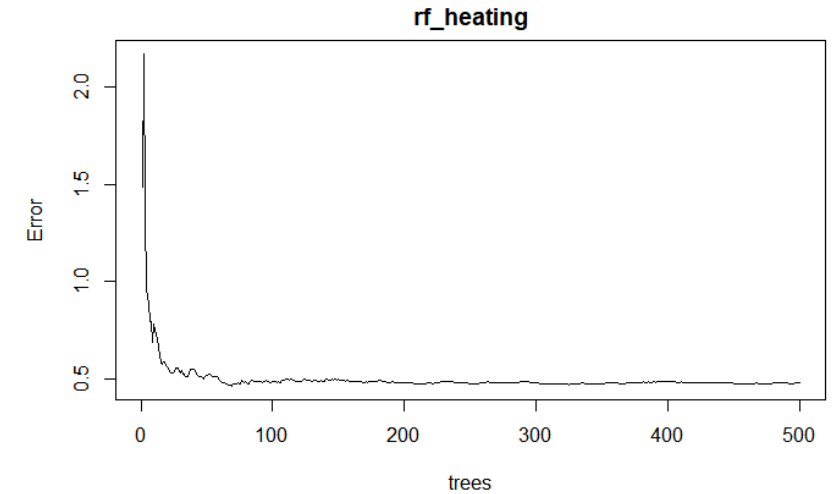
rf_heating
```

```
##
## Call:
## randomForest(formula = Heating_Load ~ ., data = M1, mtry = 3,      importance = TRUE, na.action = na.omit)
##              Type of random forest: regression
##              Number of trees: 500
## No. of variables tried at each split: 3
##
##              Mean of squared residuals: 0.4447002
##              % Var explained: 99.56
```

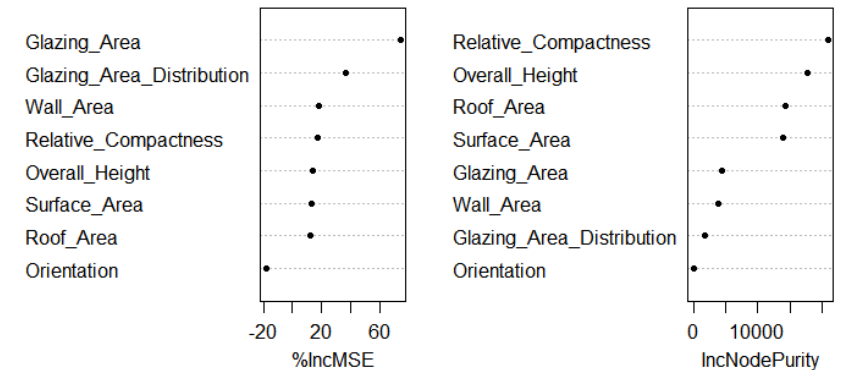
Feature importance

```
round( importance( rf_heating ), 2 )
```

```
##              %IncMSE IncNodePurity
## Relative_Compactness      17.04      20747.03
## Surface_Area              13.48      13741.94
## Wall_Area                  16.62       3631.61
## Roof_Area                  13.26      16200.41
## Overall_Height             13.32      16616.68
## Orientation                -18.72        56.31
## Glazing_Area               83.13       4440.37
## Glazing_Area_Distribution   37.85       1840.84
```



Importance of Variables



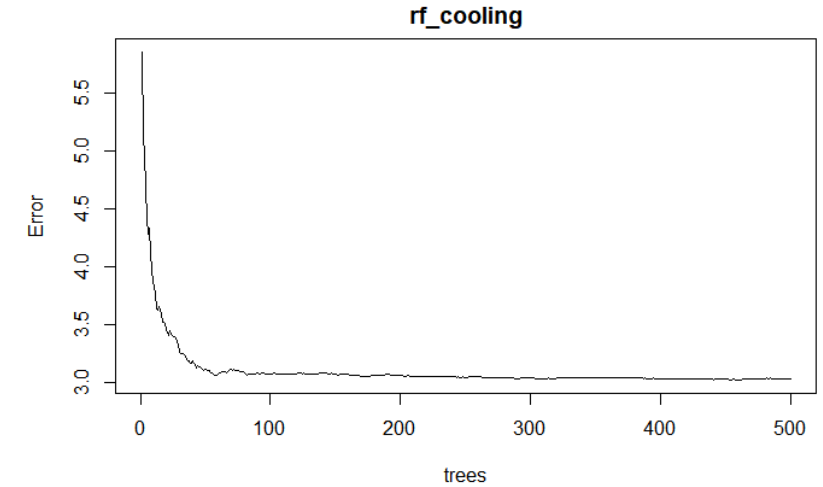
Cooling Load Random Forest

```
rf_cooling <- randomForest(Cooling_Load ~ ., data = m2, mtry=3, importance = TRUE, na.action = na.omit)
rf_cooling
```

```
##
## Call:
## randomForest(formula = Cooling_Load ~ ., data = m2, mtry = 3,      importance = TRUE, na.action = na.omit)
##              Type of random forest: regression
##              Number of trees: 500
## No. of variables tried at each split: 3
##
##              Mean of squared residuals: 3.046427
##              % Var explained: 96.63
```

```
round( importance( rf_cooling ) ,2 )
```

	%IncMSE	IncNodePurity
## Relative_Compactness	15.40	15751.47
## Surface_Area	16.22	17650.40
## Wall_Area	19.68	3318.23
## Roof_Area	13.18	14323.40
## Overall_Height	12.06	12810.72
## Orientation	2.76	251.59
## Glazing_Area	84.32	2456.18
## Glazing_Area_Distribution	25.02	826.77



Importance of Variables

