

# Assets model

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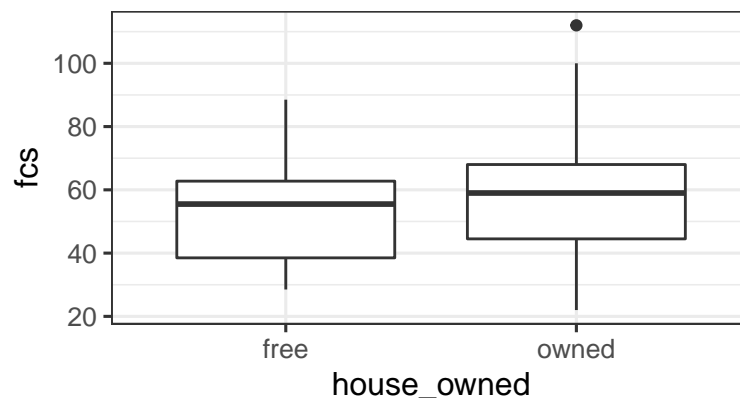
## Data Prep

```
# reading in the data
bihs <- read.csv('bihs_final.csv')
bihs_original <- read_excel('../tidy_dataset/BIHS_household_2011_15.xlsx')

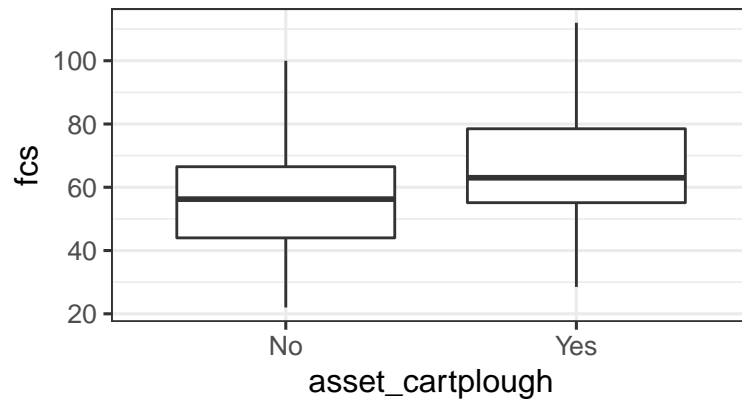
# adding house_owned variable to bihs dataset
bihs <- bihs %>% mutate(house_owned = bihs_original$house_owned) %>%
  select(
    fcs,
    year,
    house_owned,
    asset_radio,
    asset_telephone,
    asset_tractor,
    asset_television,
    asset_motorbike,
    asset_cartplough,
    asset_qty_poultry,
    asset_qty_cattle,
    asset_qty_otherlivestock,
    asset_qty_sheepgoat
  ) %>%
  na.omit()
```

## Exploratory plots

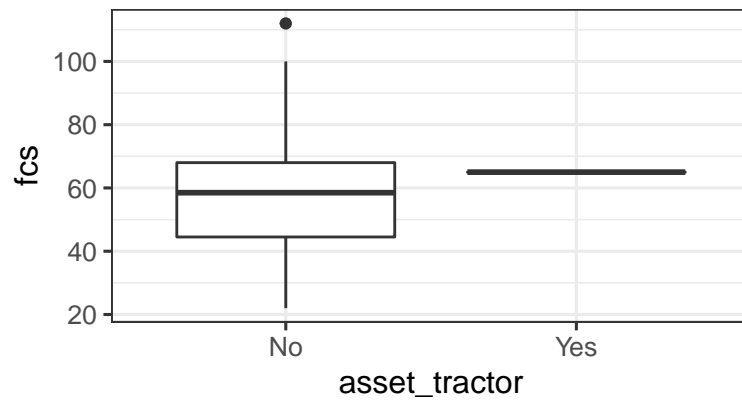
```
# whether they owned/rented a house
gf_boxplot(data = bihs, fcs ~ house_owned)
```



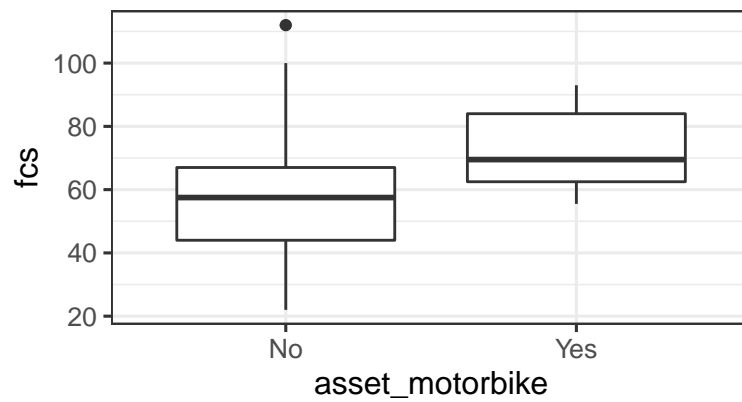
```
# tech assets
gf_boxplot(data = bihs, fcs ~ asset_cartplough)
```



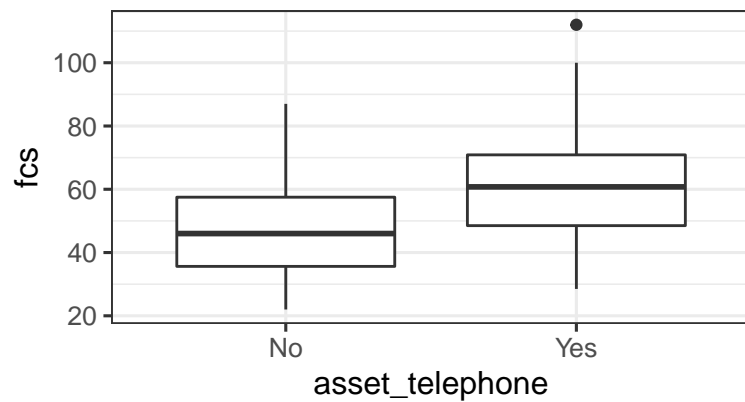
```
gf_boxplot(data = bihs, fcs ~ asset_tractor)
```



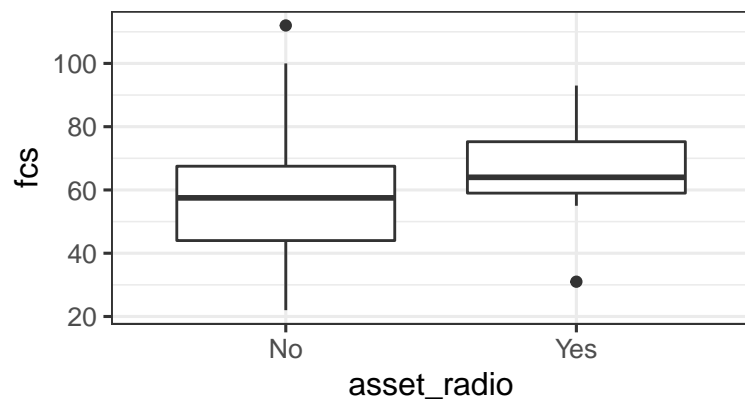
```
gf_boxplot(data = bihs, fcs ~ asset_motorbike)
```



```
gf_boxplot(data = bihs, fcs ~ asset_telephone)
```



```
gf_boxplot(data = bihs, fcs ~ asset_radio)
```



## Fitting the model

try changing categorical to 1 and 0

```
head(bihs)
```

```
##      fcs year house_owned asset_radio asset_telephone asset_tractor
## 1  59.0 2011      owned        No          Yes           No
## 2  24.5 2011      owned        No          No            No
## 3  79.5 2011      owned        No          Yes           No
## 4  59.5 2011      owned        Yes         Yes           No
## 5  40.5 2011      owned        No          Yes           No
## 6  87.5 2011      owned        No          Yes           No
##      asset_television asset_motorbike asset_cartplough asset_qty_poultry
## 1                Yes          Yes          No              0
## 2                Yes          No           No              0
## 3                No           No           No              0
## 4                No           No           No              0
## 5                No           No           No              0
## 6                No           No           No              0
##      asset_qty_cattle asset_qty_otherlivestock asset_qty_sheepgoat
## 1                0                0              0
## 2                0                0              0
## 3                0                0              0
## 4                0                0              0
## 5                0                0              0
```

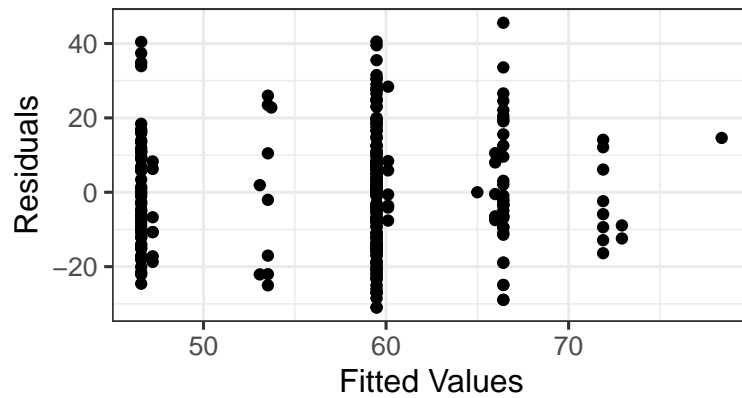
```
## 6          0          0          0
asset_lm <- lm(data = bihs, fcs ~ house_owned + asset_cartplough + asset_tractor + asset_motorbike + as
summary(asset_lm)

##
## Call:
## lm(formula = fcs ~ house_owned + asset_cartplough + asset_tractor +
##      asset_motorbike + asset_telephone + asset_radio, data = bihs,
##      na.action = "na.fail")
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -30.982 -11.982  -0.767   10.484   45.577
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      47.2158     4.3543  10.843 < 2e-16 ***
## house_ownedowned   -0.6293     4.4048  -0.143  0.8865
## asset_cartploughYes  6.9415     2.7819   2.495  0.0132 *
## asset_tractorYes    5.5183    16.3227   0.338  0.7356
## asset_motorbikeYes  12.4075     5.5977   2.217  0.0275 *
## asset_telephoneYes  12.8951     2.3477   5.493 9.64e-08 ***
## asset_radioYes      6.4970     5.0267   1.292  0.1974
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16.27 on 253 degrees of freedom
## Multiple R-squared:  0.1637, Adjusted R-squared:  0.1439
## F-statistic: 8.254 on 6 and 253 DF,  p-value: 3.588e-08
# collinearity check
vif(asset_lm)

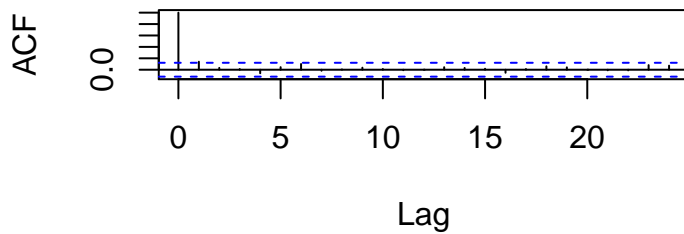
##      house_owned asset_cartplough  asset_tractor  asset_motorbike
##      1.036406      1.029941      1.002997      1.028848
##  asset_telephone  asset_radio
##      1.045979      1.005963
```

## Model assessment

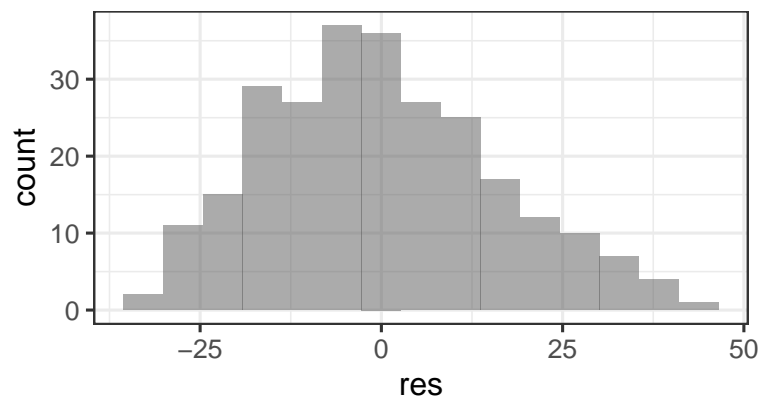
```
# linearity and constant variance - predictors are categorical tho...?
bihs <- bihs %>%
  mutate(res = resid(asset_lm),
         fitted = predict(asset_lm))
gf_point(res ~ fitted, data = bihs) %>%
  gf_labs(x = 'Fitted Values', y = 'Residuals')
```



```
# independence
acf(resid(asset_lm), main = '')
```



```
#normality of residuals
gf_histogram(~res, data = bihs, bins = 15) # they look a bit right skewed...
```



## Model Assessment

```
AIC_results <- dredge(asset_lm, rank = 'AIC')
head(AIC_results, 7)
```

```
## Global model call: lm(formula = fcs ~ house_owned + asset_cartplough + asset_tractor +
##   asset_motorbike + asset_telephone + asset_radio, data = bihs,
##   na.action = "na.fail")
## ---
## Model selection table
##   (Int) ass_crt ass_mtr ass_rad ass_tlp ass_trc hos_own df    logLik
## 12 46.95      +      +              +              5 -1091.476
## 16 46.67      +      +      +          +              6 -1090.619
## 28 46.94      +      +              +      +          6 -1091.424
```

```

## 44 47.65      +      +      +      +      + 6 -1091.459
## 32 46.66      +      +      +      +      + 7 -1090.561
## 48 47.20      +      +      +      +      + 7 -1090.609
## 14 46.69      +      +      +      +      + 5 -1093.096
##      AIC delta weight
## 12 2193.0 0.00 0.287
## 16 2193.2 0.29 0.249
## 28 2194.8 1.89 0.111
## 44 2194.9 1.97 0.107
## 32 2195.1 2.17 0.097
## 48 2195.2 2.27 0.092
## 14 2196.2 3.24 0.057
## Models ranked by AIC(x)

BIC_results <- dredge(asset_lm, rank = 'BIC')
head(BIC_results, 7)

## Global model call: lm(formula = fcs ~ house_owned + asset_cartplough + asset_tractor +
##      asset_motorbike + asset_telephone + asset_radio, data = bihs,
##      na.action = "na.fail")
## ---
## Model selection table
##      (Int) ass_crt ass_mtr ass_rad ass_tlp df      logLik      BIC delta weight
## 9  47.66          +          +          + 3 -1096.764 2210.2 0.00 0.280
## 10 47.01          +          +          + 4 -1094.145 2210.5 0.32 0.238
## 12 46.95          +          +          + 5 -1091.476 2210.8 0.55 0.213
## 11 47.66          +          +          + 4 -1094.647 2211.5 1.33 0.144
## 13 47.34          +          +          + 4 -1095.696 2213.6 3.42 0.051
## 14 46.69          +          +          + 5 -1093.096 2214.0 3.79 0.042
## 16 46.67          +          +          + 6 -1090.619 2214.6 4.39 0.031
## Models ranked by BIC(x)

# best bic model...?
asset_lm_bic <- lm(data = bihs, fcs ~ asset_telephone, na.action = 'na.fail')
summary(asset_lm_bic)

##
## Call:
## lm(formula = fcs ~ asset_telephone, data = bihs, na.action = "na.fail")
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -33.151 -13.151  -1.151   9.838  50.349
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    47.662     2.001  23.823 < 2e-16 ***
## asset_telephoneYes 13.989     2.328   6.009 6.34e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16.5 on 258 degrees of freedom
## Multiple R-squared:  0.1228, Adjusted R-squared:  0.1194
## F-statistic: 36.1 on 1 and 258 DF, p-value: 6.342e-09

```

```
asset_lm_all <- lm(data = bihs_original, fcs ~ factor(survey_year) + asset_qty_poultry + asset_qty_cattle +
summary(asset_lm_all)
```

```
##
## Call:
## lm(formula = fcs ~ factor(survey_year) + asset_qty_poultry +
##      asset_qty_cattle + asset_qty_otherlivestock + asset_qty_sheepgoat +
##      memb_total + memb_und15 + memb_15_44 + hhs_total + bio_bio_1 +
##      bio_bio_12, data = bihs_original)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -37.119 -12.054  -0.865   10.595   51.877
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      -27.596536   270.655923  -0.102  0.918827
## factor(survey_year)2015    8.403096    1.748418   4.806 2.02e-06 ***
## asset_qty_poultry    0.098423    0.136666    0.720 0.471744
## asset_qty_cattle     2.423539    0.698881    3.468 0.000569 ***
## asset_qty_otherlivestock -0.162446    0.315243  -0.515 0.606561
## asset_qty_sheepgoat   -0.014030    1.046830  -0.013 0.989312
## memb_total          0.976431    0.658199    1.483 0.138553
## memb_und15         -0.045341    0.802431  -0.057 0.954962
## memb_15_44          1.150961    0.849113    1.355 0.175853
## hhs_total            7.646841    1.358407    5.629 2.97e-08 ***
## bio_bio_1           2.180512   10.094620    0.216 0.829067
## bio_bio_12          -0.008513    0.006454  -1.319 0.187751
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16.24 on 518 degrees of freedom
## (14 observations deleted due to missingness)
## Multiple R-squared:  0.2848, Adjusted R-squared:  0.2696
## F-statistic: 18.76 on 11 and 518 DF,  p-value: < 2.2e-16
```

```
vif(asset_lm_all)
```

```
##      factor(survey_year)      asset_qty_poultry      asset_qty_cattle
##      1.534704              1.387393              1.469670
## asset_qty_otherlivestock      asset_qty_sheepgoat      memb_total
##      1.041566              1.099741              3.321780
##      memb_und15              memb_15_44              hhs_total
##      1.885714              2.026934              1.175876
##      bio_bio_1              bio_bio_12
##      9.987758              9.892606
```

```
asset_lm_all_two <- lm(data = bihs_original, fcs ~ factor(survey_year) + asset_qty_poultry + asset_qty_cattle +
summary(asset_lm_all_two)
```

```
##
## Call:
## lm(formula = fcs ~ factor(survey_year) + asset_qty_poultry +
##      asset_qty_cattle + asset_qty_otherlivestock + asset_qty_sheepgoat +
```

```
##      memb_total + memb_und15 + memb_15_44 + hhs_total + bio_bio_1 +
##      bio_bio_12 + factor(house_owned) + factor(asset_cartplough) +
##      factor(asset_telephone), data = bihs_original)
##
## Residuals:
##      Min        1Q    Median        3Q        Max
## -31.560 -10.627  -0.697   9.491  44.950
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      132.268232  353.041500   0.375 0.708197
## factor(survey_year)2015      2.824916   4.567763   0.618 0.536775
## asset_qty_poultry      0.002366   0.240925   0.010 0.992170
## asset_qty_cattle      2.758238   1.300787   2.120 0.034833 *
## asset_qty_otherlivestock -0.365057   0.339079  -1.077 0.282563
## asset_qty_sheepgoat    -1.411200   2.072145  -0.681 0.496403
## memb_total      2.758834   1.104307   2.498 0.013044 *
## memb_und15      -1.840420   1.175480  -1.566 0.118534
## memb_15_44      -1.158268   1.261428  -0.918 0.359281
## hhs_total      4.812035   1.552080   3.100 0.002126 **
## bio_bio_1      -3.415063  13.176284  -0.259 0.795681
## bio_bio_12      -0.010803   0.008462  -1.277 0.202776
## factor(house_owned)owned    -4.103573   4.314379  -0.951 0.342340
## factor(asset_cartplough)Yes  4.990623   2.556769   1.952 0.051926 .
## factor(asset_telephone)Yes  8.504240   2.513287   3.384 0.000815 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.56 on 285 degrees of freedom
## (244 observations deleted due to missingness)
## Multiple R-squared:  0.2986, Adjusted R-squared:  0.2641
## F-statistic: 8.665 on 14 and 285 DF,  p-value: 1.192e-15
```

```
vif(asset_lm_all_two)
```

```
##      factor(survey_year)      asset_qty_poultry      asset_qty_cattle
##              3.175399              1.785341              2.411222
## asset_qty_otherlivestock      asset_qty_sheepgoat      memb_total
##              1.094819              1.175269              5.405350
##              memb_und15              memb_15_44              hhs_total
##              2.550880              2.752706              1.334021
##              bio_bio_1              bio_bio_12      factor(house_owned)
##              10.335599              10.288256              1.095878
## factor(asset_cartplough) factor(asset_telephone)
##              1.312503              1.372369
```