

keşifçi veri analizi ve görselleştirme

ilke

2022-08-05

veri setinin 6 boyutlu gösterimi 1.boyut: değişkenler 2.'': değişkenlerin dağılımı(histogram) 3.boyut : yoğunluk 4.boyut: ilişki 5.boyut: korelasyona ait p-value 6.boyut:scatter plot

library(tidyverse)

```
## — Attaching packages — tidyverse 1.3.2 —
## ✔ ggplot2 3.3.6   ✔ purrr 0.3.4
## ✔ tibble 3.1.8   ✔ dplyr 1.0.9
## ✔ tidyr 1.2.0    ✔ stringr 1.4.0
## ✔ readr 2.1.2    ✔ forcats 0.5.1
## — Conflicts — tidyverse_conflicts() —
## ✖ dplyr::filter() masks stats::filter()
## ✖ dplyr::lag() masks stats::lag()
```

df <- mpg

```
df$class <- factor(df$class)
glimpse(df)
```

```
## Rows: 234
## Columns: 11
## $ manufacturer <chr> "audi", "audi", "audi", "audi", "audi", "audi", "audi", "...
## $ model <chr> "a4", "a4", "a4", "a4", "a4", "a4", "a4", "a4 quattro", "...
## $ displ <dbl> 1.8, 1.8, 2.0, 2.0, 2.8, 2.8, 3.1, 1.8, 1.8, 2.0, 2.0, 2....
## $ year <int> 1999, 1999, 2008, 2008, 1999, 1999, 2008, 1999, 1999, 200...
## $ cyl <int> 4, 4, 4, 4, 6, 6, 6, 4, 4, 4, 4, 6, 6, 6, 6, 6, 8, 8, ...
## $ trans <chr> "auto(l5)", "manual(m5)", "manual(m6)", "auto(av)", "auto...
## $ drv <chr> "f", "f", "f", "f", "f", "f", "f", "4", "4", "4", "4", "4...
## $ cty <int> 18, 21, 20, 21, 16, 18, 18, 18, 16, 20, 19, 15, 17, 17, 1...
## $ hwy <int> 29, 29, 31, 30, 26, 26, 27, 26, 25, 28, 27, 25, 25, 25, 2...
## $ fl <chr> "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p...
## $ class <fct> compact, compact, compact, compact, compact, compact, com...
```

```
df$disp_yeni <- ifelse(df$displ > 2, "bir", "sifir") #disp değerleri ikiden büyükse 1 değilse 0
df$disp_yeni <- factor(df$disp_yeni)
df$disp_iki <- cut(df$displ, breaks = c(0,2, max(df$displ)))
```

ilk bakış

colnames(mpg)

```
## [1] "manufacturer" "model" "displ" "year" "cyl"
## [6] "trans" "drv" "cty" "hwy" "fl"
## [11] "class"
```

rownames(mpg)

```
## [1] "1" "2" "3" "4" "5" "6" "7" "8" "9" "10" "11" "12"
## [13] "13" "14" "15" "16" "17" "18" "19" "20" "21" "22" "23" "24"
## [25] "25" "26" "27" "28" "29" "30" "31" "32" "33" "34" "35" "36"
## [37] "37" "38" "39" "40" "41" "42" "43" "44" "45" "46" "47" "48"
## [49] "49" "50" "51" "52" "53" "54" "55" "56" "57" "58" "59" "60"
## [61] "61" "62" "63" "64" "65" "66" "67" "68" "69" "70" "71" "72"
## [73] "73" "74" "75" "76" "77" "78" "79" "80" "81" "82" "83" "84"
## [85] "85" "86" "87" "88" "89" "90" "91" "92" "93" "94" "95" "96"
## [97] "97" "98" "99" "100" "101" "102" "103" "104" "105" "106" "107" "108"
## [109] "109" "110" "111" "112" "113" "114" "115" "116" "117" "118" "119" "120"
## [121] "121" "122" "123" "124" "125" "126" "127" "128" "129" "130" "131" "132"
## [133] "133" "134" "135" "136" "137" "138" "139" "140" "141" "142" "143" "144"
## [145] "145" "146" "147" "148" "149" "150" "151" "152" "153" "154" "155" "156"
## [157] "157" "158" "159" "160" "161" "162" "163" "164" "165" "166" "167" "168"
## [169] "169" "170" "171" "172" "173" "174" "175" "176" "177" "178" "179" "180"
## [181] "181" "182" "183" "184" "185" "186" "187" "188" "189" "190" "191" "192"
## [193] "193" "194" "195" "196" "197" "198" "199" "200" "201" "202" "203" "204"
## [205] "205" "206" "207" "208" "209" "210" "211" "212" "213" "214" "215" "216"
## [217] "217" "218" "219" "220" "221" "222" "223" "224" "225" "226" "227" "228"
## [229] "229" "230" "231" "232" "233" "234"
```

```
nrow(mpg)
```

```
## [1] 234
```

```
ncol(mpg)
```

```
## [1] 11
```

```
head(mpg)
```

```
## # A tibble: 6 × 11
##   manufacturer model displ  year  cyl trans   drv   cty   hwy fl   class
##   <chr>         <chr> <dbl> <int> <int> <chr>   <chr> <int> <int> <chr> <chr>
## 1 audi         a4     1.8 1999    4 auto(l5) f    18   29 p   compa...
## 2 audi         a4     1.8 1999    4 manual(m5) f    21   29 p   compa...
## 3 audi         a4     2   2008    4 manual(m6) f    20   31 p   compa...
## 4 audi         a4     2   2008    4 auto(av) f    21   30 p   compa...
## 5 audi         a4     2.8 1999    6 auto(l5) f    16   26 p   compa...
## 6 audi         a4     2.8 1999    6 manual(m5) f    18   26 p   compa...
```

```
str(mpg)
```

```
## tibble [234 × 11] (S3: tbl_df/tbl/data.frame)
## $ manufacturer: chr [1:234] "audi" "audi" "audi" "audi" ...
## $ model       : chr [1:234] "a4" "a4" "a4" "a4" ...
## $ displ       : num [1:234] 1.8 1.8 2 2 2.8 2.8 3.1 1.8 1.8 2 ...
## $ year        : int [1:234] 1999 1999 2008 2008 1999 1999 2008 1999 1999 2008 ...
## $ cyl         : int [1:234] 4 4 4 4 6 6 6 4 4 4 ...
## $ trans       : chr [1:234] "auto(l5)" "manual(m5)" "manual(m6)" "auto(av)" ...
## $ drv         : chr [1:234] "f" "f" "f" "f" ...
## $ cty         : int [1:234] 18 21 20 21 16 18 18 18 16 20 ...
## $ hwy         : int [1:234] 29 29 31 30 26 26 27 26 25 28 ...
## $ fl          : chr [1:234] "p" "p" "p" "p" ...
## $ class       : chr [1:234] "compact" "compact" "compact" "compact" ...
```

```
glimpse(mpg) #veri seti genel yapı hakkında bilgi
```

```
## Rows: 234
## Columns: 11
## $ manufacturer <chr> "audi", "audi", "audi", "audi", "audi", "audi", "audi", "...
## $ model      <chr> "a4", "a4", "a4", "a4", "a4", "a4", "a4", "a4 quattro", "...
## $ displ      <dbl> 1.8, 1.8, 2.0, 2.0, 2.8, 2.8, 3.1, 1.8, 1.8, 2.0, 2.0, 2....
## $ year       <int> 1999, 1999, 2008, 2008, 1999, 1999, 2008, 1999, 1999, 200...
## $ cyl        <int> 4, 4, 4, 4, 6, 6, 6, 4, 4, 4, 4, 6, 6, 6, 6, 6, 8, 8, ...
## $ trans      <chr> "auto(l5)", "manual(m5)", "manual(m6)", "auto(av)", "auto...
## $ drv        <chr> "f", "f", "f", "f", "f", "f", "f", "4", "4", "4", "4", "4...
## $ cty        <int> 18, 21, 20, 21, 16, 18, 18, 18, 16, 20, 19, 15, 17, 17, 1...
## $ hwy        <int> 29, 29, 31, 30, 26, 26, 27, 26, 25, 28, 27, 25, 25, 25, 2...
## $ fl         <chr> "p", "p", "p", "p", "p", "p", "p", "p", "p", "p", "p...
## $ class      <chr> "compact", "compact", "compact", "compact", "compact", "c...
```

```
psych::headTail(mpg) #ilk 4 son 4 gözlem
```

```
##  manufacturer model displ year cyl  trans drv cty hwy  fl  class
## 1      audi   a4   1.8 1999   4 auto(l5)  f  18  29   p compact
## 2      audi   a4   1.8 1999   4 manual(m5) f  21  29   p compact
## 3      audi   a4   2.0 2008   4 manual(m6)  f  20  31   p compact
## 4      audi   a4   2.0 2008   4 auto(av)   f  21  30   p compact
## 5      <NA>  <NA>   ... ..   <NA> <NA>   ... .. <NA>  <NA>
## 6 volkswagen passat  2.0 2008   4 manual(m6) f  21  29   p midsize
## 7 volkswagen passat  2.8 1999   6 auto(l5)   f  16  26   p midsize
## 8 volkswagen passat  2.8 1999   6 manual(m5) f  18  26   p midsize
## 9 volkswagen passat  3.6 2008   6 auto(s6)   f  17  26   p midsize
```

```
summary(df)
```

```
##  manufacturer      model      displ      year
## Length:234      Length:234      Min.   :1.600 Min.   :1999
## Class :character Class :character 1st Qu.:2.400 1st Qu.:1999
## Mode  :character Mode  :character Median :3.300 Median :2004
##              Mean  :3.472 Mean  :2004
##              3rd Qu.:4.600 3rd Qu.:2008
##              Max.   :7.000 Max.   :2008
##
##      cyl      trans      drv      cty
## Min.   :4.000 Length:234 Length:234 Min.   :9.00
## 1st Qu.:4.000 Class :character Class :character 1st Qu.:14.00
## Median :6.000 Mode  :character Mode  :character Median :17.00
## Mean   :5.889              Mean  :16.86
## 3rd Qu.:8.000              3rd Qu.:19.00
## Max.   :8.000              Max.   :35.00
##
##      hwy      fl      class disp_yeni disp_iki
## Min.   :12.00 Length:234 2seater  :5  bir  :191 (0,2]: 43
## 1st Qu.:18.00 Class :character compact :47  sifir: 43 (2,7]:191
## Median :24.00 Mode  :character midsize  :41
## Mean   :23.44              minivan  :11
## 3rd Qu.:27.00              pickup   :33
## Max.   :44.00              subcompact:35
##              suv          :62
```

```
fivenum(df$displ)
```

```
## [1] 1.6 2.4 3.3 4.6 7.0
```

```
levels(df$class)
```

```
## [1] "2seater" "compact" "midsize" "minivan" "pickup"
## [6] "subcompact" "suv"
```

hızlı detaylı genel bakış

```
library(funModeling)
```

```
## Loading required package: Hmisc
```

```
## Loading required package: lattice
```

```
## Loading required package: survival
```

```
## Loading required package: Formula
```

```
##  
## Attaching package: 'Hmisc'
```

```
## The following objects are masked from 'package:dplyr':  
##  
##   src, summarize
```

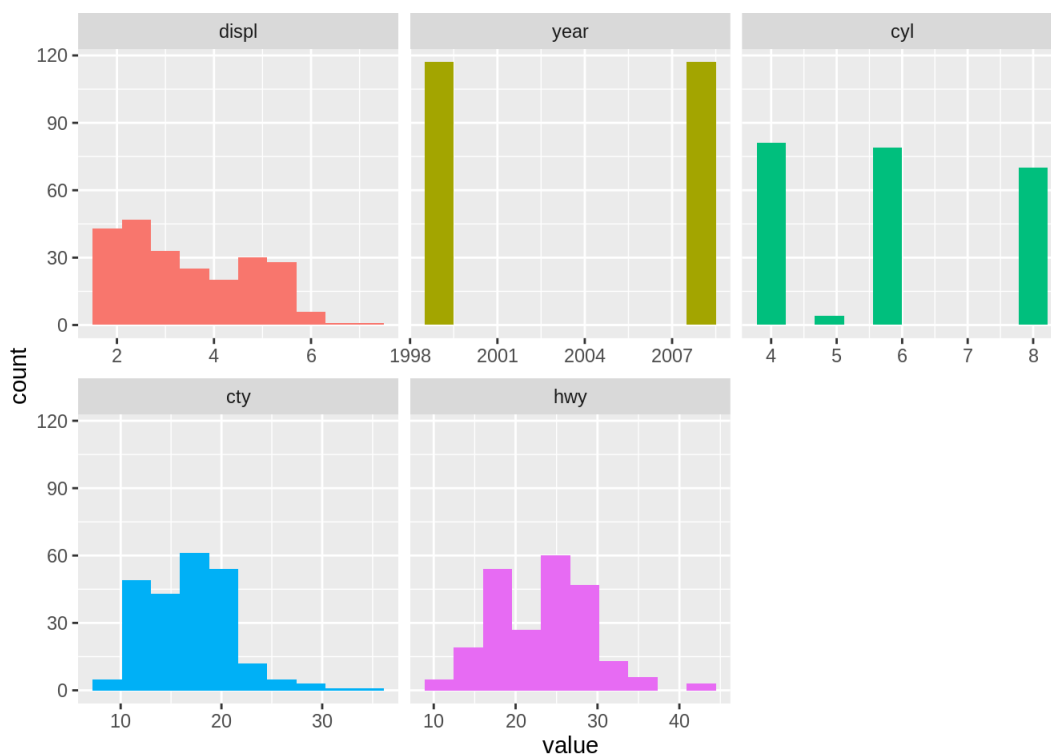
```
## The following objects are masked from 'package:base':  
##  
##   format.pval, units
```

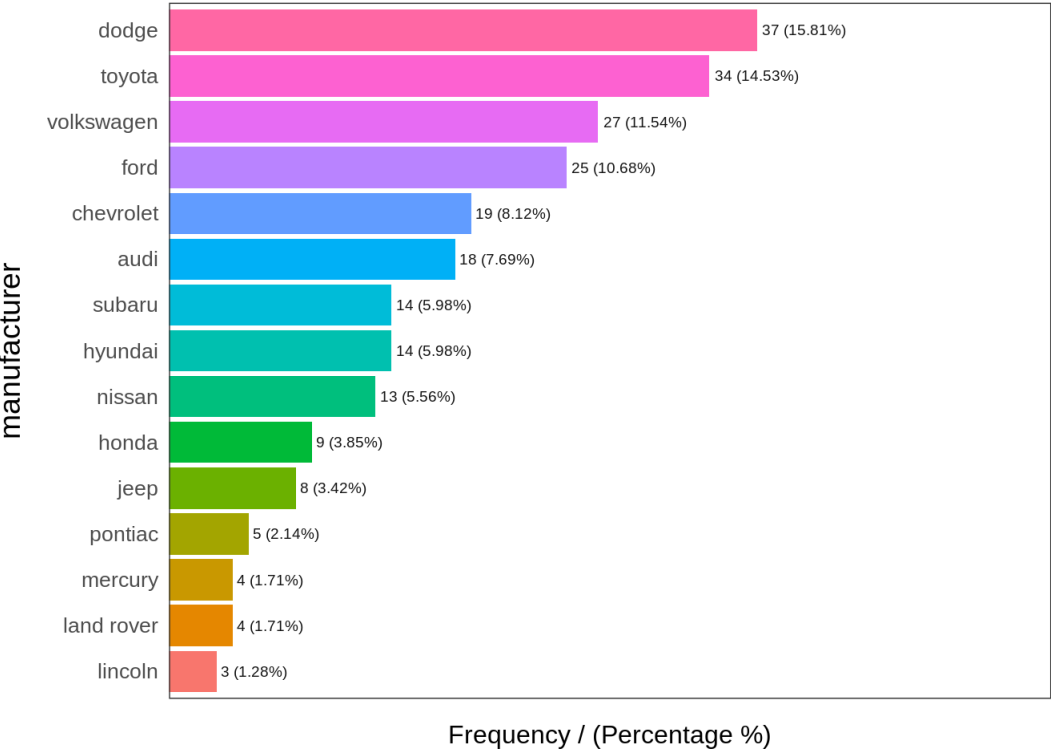
```
## funModeling v.1.9.4 :)  
## Examples and tutorials at livebook.datascienceheroes.com  
## / Now in Spanish: librovivodecienciadedatos.ai
```

```
profiling_num(df) #sürekli değişkenlerin özet bazı istatistiklerini verir
```

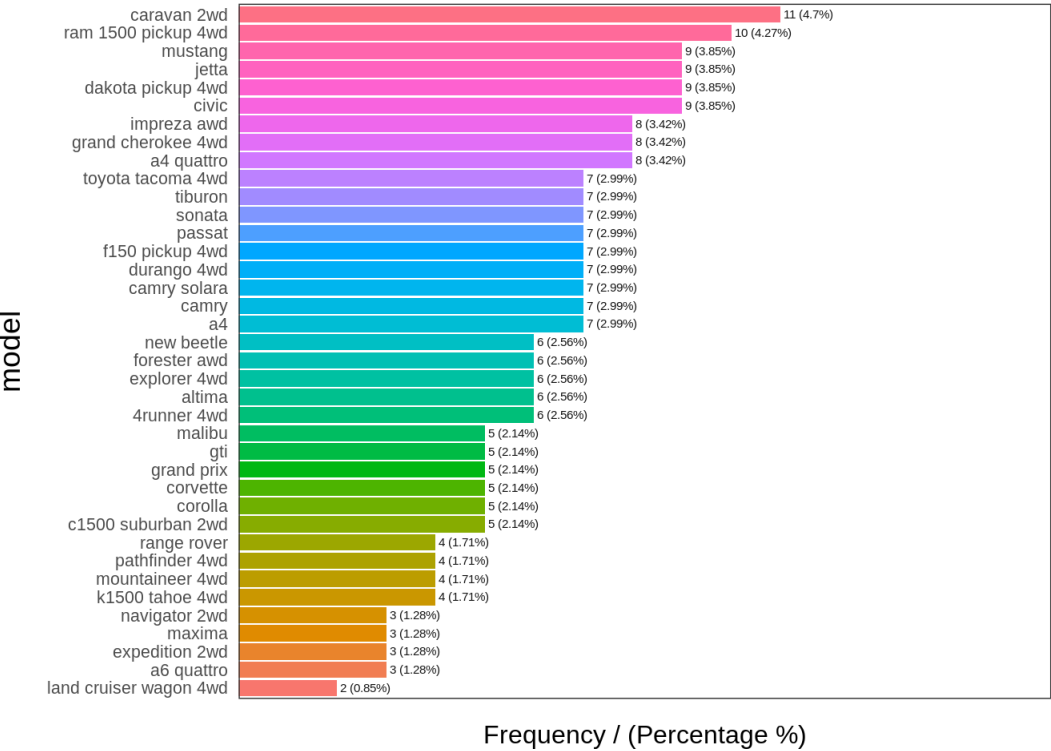
```
## variable mean std_dev variation_coef p_01 p_05 p_25 p_50  
## 1 displ 3.471795 1.291959 0.372130002 1.6 1.8 2.4 3.3  
## 2 year 2003.500000 4.509646 0.002250884 1999.0 1999.0 1999.0 2003.5  
## 3 cyl 5.888889 1.611534 0.273656799 4.0 4.0 4.0 6.0  
## 4 cty 16.858974 4.255946 0.252443926 9.0 11.0 14.0 17.0  
## 5 hwy 23.440171 5.954643 0.254035837 12.0 15.0 18.0 24.0  
## p_75 p_95 p_99 skewness kurtosis iqr range_98  
## 1 4.6 5.7 6.20 0.4414630 2.107412 2.2 [1.6, 6.2]  
## 2 2008.0 2008.0 2008.00 0.0000000 1.000000 9.0 [1999, 2008]  
## 3 8.0 8.0 8.00 0.1130695 1.549122 4.0 [4, 8]  
## 4 19.0 24.0 28.67 0.7914453 4.468651 5.0 [9, 28.67]  
## 5 27.0 32.0 39.68 0.3668650 3.163929 9.0 [12, 39.67999999999999]  
## range_80  
## 1 [2, 5.4]  
## 2 [1999, 2008]  
## 3 [4, 8]  
## 4 [11, 21]  
## 5 [16.3, 30]
```

```
plot_num(df) #genel histogram
```

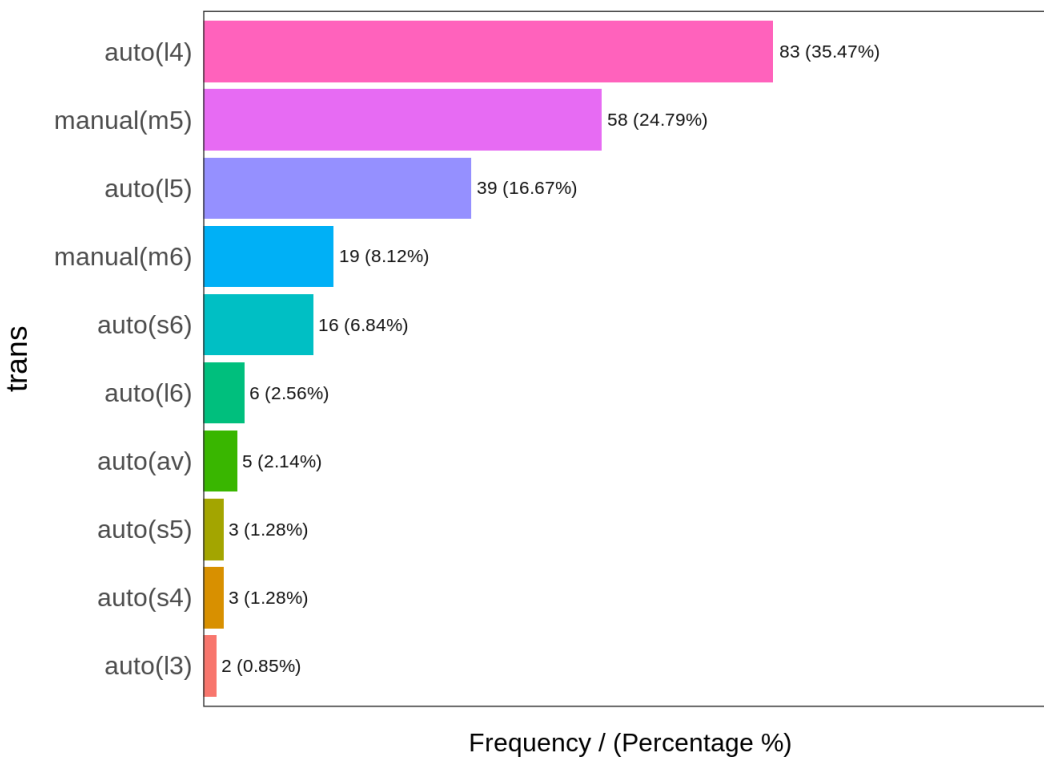




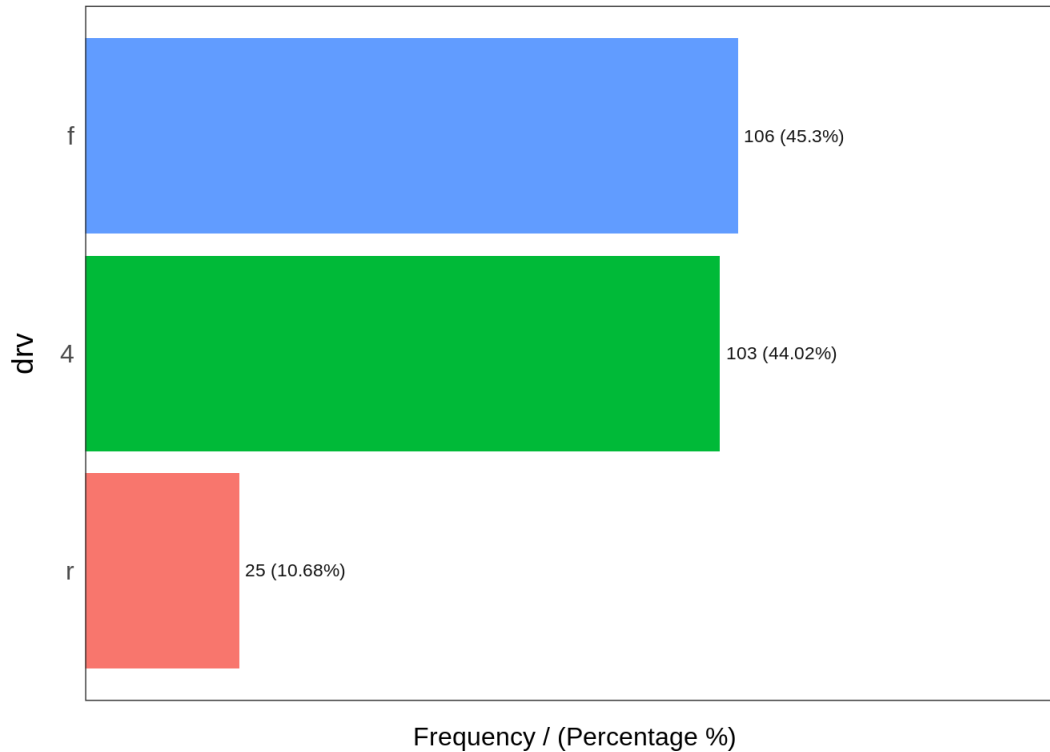
##	manufacturer	frequency	percentage	cumulative_perc
## 1	dodge	37	15.81	15.81
## 2	toyota	34	14.53	30.34
## 3	volkswagen	27	11.54	41.88
## 4	ford	25	10.68	52.56
## 5	chevrolet	19	8.12	60.68
## 6	audi	18	7.69	68.37
## 7	hyundai	14	5.98	74.35
## 8	subaru	14	5.98	80.33
## 9	nissan	13	5.56	85.89
## 10	honda	9	3.85	89.74
## 11	jeep	8	3.42	93.16
## 12	pontiac	5	2.14	95.30
## 13	land rover	4	1.71	97.01
## 14	mercury	4	1.71	98.72
## 15	lincoln	3	1.28	100.00



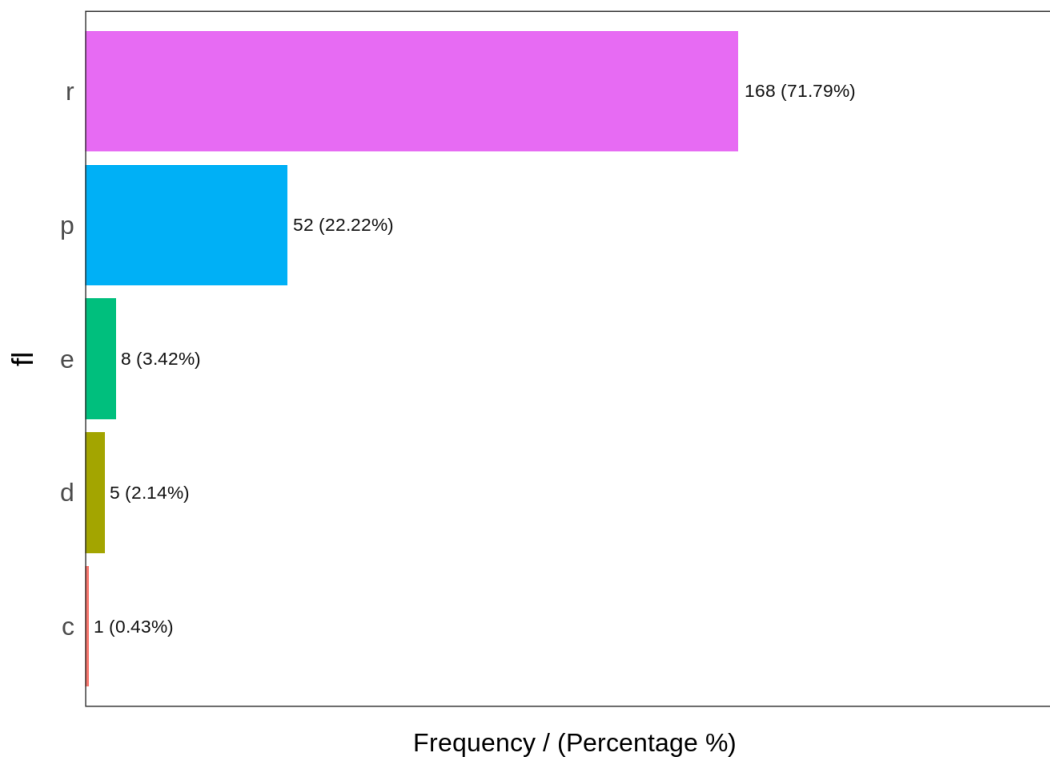
##	model	frequency	percentage	cumulative_perc
## 1	caravan 2wd	11	4.70	4.70
## 2	ram 1500 pickup 4wd	10	4.27	8.97
## 3	civic	9	3.85	12.82
## 4	dakota pickup 4wd	9	3.85	16.67
## 5	jetta	9	3.85	20.52
## 6	mustang	9	3.85	24.37
## 7	a4 quattro	8	3.42	27.79
## 8	grand cherokee 4wd	8	3.42	31.21
## 9	impreza awd	8	3.42	34.63
## 10	a4	7	2.99	37.62
## 11	camry	7	2.99	40.61
## 12	camry solara	7	2.99	43.60
## 13	durango 4wd	7	2.99	46.59
## 14	f150 pickup 4wd	7	2.99	49.58
## 15	passat	7	2.99	52.57
## 16	sonata	7	2.99	55.56
## 17	tiburon	7	2.99	58.55
## 18	toyota tacoma 4wd	7	2.99	61.54
## 19	4runner 4wd	6	2.56	64.10
## 20	altima	6	2.56	66.66
## 21	explorer 4wd	6	2.56	69.22
## 22	forester awd	6	2.56	71.78
## 23	new beetle	6	2.56	74.34
## 24	c1500 suburban 2wd	5	2.14	76.48
## 25	corolla	5	2.14	78.62
## 26	corvette	5	2.14	80.76
## 27	grand prix	5	2.14	82.90
## 28	gti	5	2.14	85.04
## 29	malibu	5	2.14	87.18
## 30	k1500 tahoe 4wd	4	1.71	88.89
## 31	mountaineer 4wd	4	1.71	90.60
## 32	pathfinder 4wd	4	1.71	92.31
## 33	range rover	4	1.71	94.02
## 34	a6 quattro	3	1.28	95.30
## 35	expedition 2wd	3	1.28	96.58
## 36	maxima	3	1.28	97.86
## 37	navigator 2wd	3	1.28	99.14
## 38	land cruiser wagon 4wd	2	0.85	100.00



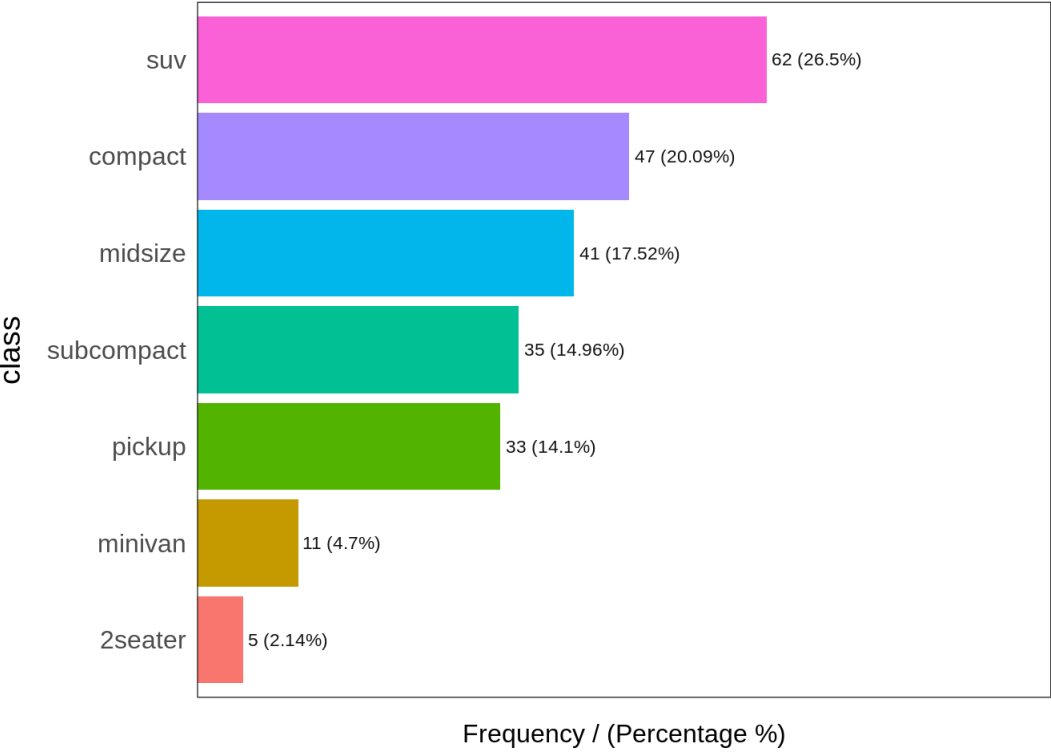
##	trans	frequency	percentage	cumulative_perc
## 1	auto(l4)	83	35.47	35.47
## 2	manual(m5)	58	24.79	60.26
## 3	auto(l5)	39	16.67	76.93
## 4	manual(m6)	19	8.12	85.05
## 5	auto(s6)	16	6.84	91.89
## 6	auto(l6)	6	2.56	94.45
## 7	auto(av)	5	2.14	96.59
## 8	auto(s4)	3	1.28	97.87
## 9	auto(s5)	3	1.28	99.15
## 10	auto(l3)	2	0.85	100.00



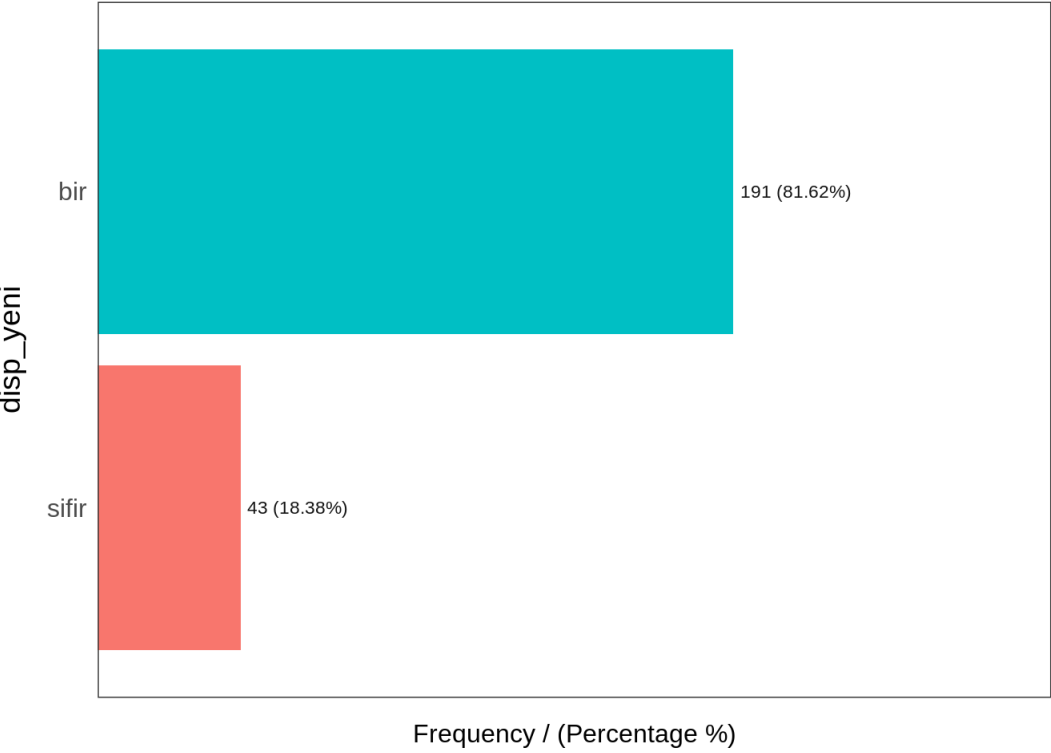
##	drv	frequency	percentage	cumulative_perc
## 1	f	106	45.30	45.30
## 2	4	103	44.02	89.32
## 3	r	25	10.68	100.00



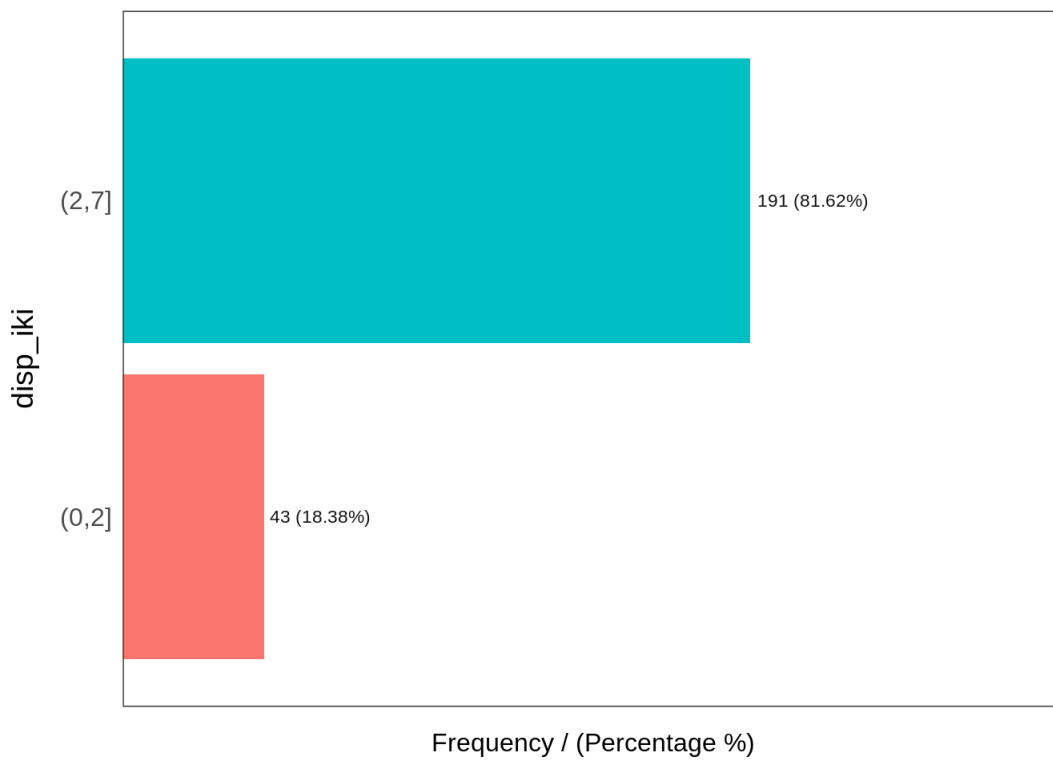
##	fl	frequency	percentage	cumulative_perc
## 1	r	168	71.79	71.79
## 2	p	52	22.22	94.01
## 3	e	8	3.42	97.43
## 4	d	5	2.14	99.57
## 5	c	1	0.43	100.00



##	class	frequency	percentage	cumulative_perc
## 1	suv	62	26.50	26.50
## 2	compact	47	20.09	46.59
## 3	midsize	41	17.52	64.11
## 4	subcompact	35	14.96	79.07
## 5	pickup	33	14.10	93.17
## 6	minivan	11	4.70	97.87
## 7	2seater	5	2.14	100.00



##	disp_yeni	frequency	percentage	cumulative_perc
## 1	bir	191	81.62	81.62
## 2	sifir	43	18.38	100.00



```
## disp_iki frequency percentage cumulative_perc
## 1 (2,7] 191 81.62 81.62
## 2 (0,2] 43 18.38 100.00
```

```
## [1] "Variables processed: manufacturer, model, trans, drv, fl, class, disp_yeni, disp_iki"
```

Detaylı

```
library(psych)
```

```
##
## Attaching package: 'psych'
```

```
## The following object is masked from 'package:Hmisc':
##
## describe
```

```
## The following objects are masked from 'package:ggplot2':
##
## %+%, alpha
```

```
describe(mpg$displ) #displ bazı istatistikleri
```

```
## vars n mean sd median trimmed mad min max range skew kurtosis se
## X1 1 234 3.47 1.29 3.3 3.39 1.33 1.6 7 5.4 0.44 -0.91 0.08
```

```
library(Hmisc)
Hmisc::describe(mpg$displ) #en küçük en yüksek değerleri verir. aykırı gözlem tespiti için kullanılabilir
```

```
## mpg$displ
## n missing distinct Info Mean Gmd .05 .10
## 234 0 35 0.997 3.472 1.471 1.8 2.0
## .25 .50 .75 .90 .95
## 2.4 3.3 4.6 5.4 5.7
##
## lowest : 1.6 1.8 1.9 2.0 2.2, highest: 6.0 6.1 6.2 6.5 7.0
```

```
library(pastecs)
```

```
##
## Attaching package: 'pastecs'
```

```
## The following objects are masked from 'package:dplyr':
```

```
##
```

```
## first, last
```

```
## The following object is masked from 'package:tidyr':
```

```
##
```

```
## extract
```

```
stat.desc(mpg) #sürekli değişkenlerin tümünün min,median,range vs vs detaylı data.frame olarak verir
```

```
##      manufacturer model   displ   year   cyl trans drv
## nbr.val      NA    NA 234.000000 2.340000e+02 234.0000000  NA NA
## nbr.null      NA    NA  0.000000 0.000000e+00  0.0000000  NA NA
## nbr.na        NA    NA  0.000000 0.000000e+00  0.0000000  NA NA
## min           NA    NA  1.600000 1.999000e+03  4.0000000  NA NA
## max           NA    NA  7.000000 2.008000e+03  8.0000000  NA NA
## range         NA    NA  5.400000 9.000000e+00  4.0000000  NA NA
## sum           NA    NA 812.400000 4.688190e+05 1378.0000000  NA NA
## median        NA    NA  3.300000 2.003500e+03  6.0000000  NA NA
## mean          NA    NA  3.471795 2.003500e+03  5.8888889  NA NA
## SE.mean       NA    NA  0.084458 2.948048e-01  0.1053493  NA NA
## CI.mean       NA    NA  0.166399 5.808237e-01  0.2075589  NA NA
## var           NA    NA  1.669158 2.033691e+01  2.5970434  NA NA
## std.dev       NA    NA  1.291959 4.509646e+00  1.6115345  NA NA
## coef.var      NA    NA  0.372130 2.250884e-03  0.2736568  NA NA
##      cty      hwy fl class
## nbr.val 234.0000000 234.0000000 NA  NA
## nbr.null 0.0000000  0.0000000 NA  NA
## nbr.na   0.0000000  0.0000000 NA  NA
## min      9.0000000 12.0000000 NA  NA
## max      35.0000000 44.0000000 NA  NA
## range     26.0000000 32.0000000 NA  NA
## sum      3945.0000000 5485.0000000 NA  NA
## median    17.0000000 24.0000000 NA  NA
## mean     16.8589744 23.4401709 NA  NA
## SE.mean   0.2782199  0.3892672 NA  NA
## CI.mean   0.5481481  0.7669333 NA  NA
## var       18.1130736 35.4577785 NA  NA
## std.dev   4.2559457  5.9546434 NA  NA
## coef.var  0.2524439  0.2540358 NA  NA
```

```
library(d3Tree)
```

```
d3tree(list(root = df2tree(rootname = 'Titanic',
                             struct = as.data.frame(Titanic)),
          layout = 'collapse')) #anlamlandırmayı kolaylaştırır
```

bar plot (sütun grafiği)

```
#örnek1
```

```
df <- diamonds
```

```
glimpse(df)
```

```
## Rows: 53,940
```

```
## Columns: 10
```

```
## $ carat <dbl> 0.23, 0.21, 0.23, 0.29, 0.31, 0.24, 0.24, 0.26, 0.22, 0.23, 0....
```

```
## $ cut <ord> Ideal, Premium, Good, Premium, Good, Very Good, Very Good, Ver...
```

```
## $ color <ord> E, E, E, I, J, J, I, H, E, H, J, J, F, J, E, E, I, J, J, J, I,...
```

```
## $ clarity <ord> SI2, SI1, VS1, VS2, SI2, VVS2, VVS1, SI1, VS2, VS1, SI1, VS1, ...
```

```
## $ depth <dbl> 61.5, 59.8, 56.9, 62.4, 63.3, 62.8, 62.3, 61.9, 65.1, 59.4, 64...
```

```
## $ table <dbl> 55, 61, 65, 58, 58, 57, 57, 55, 61, 61, 55, 56, 61, 54, 62, 58...
```

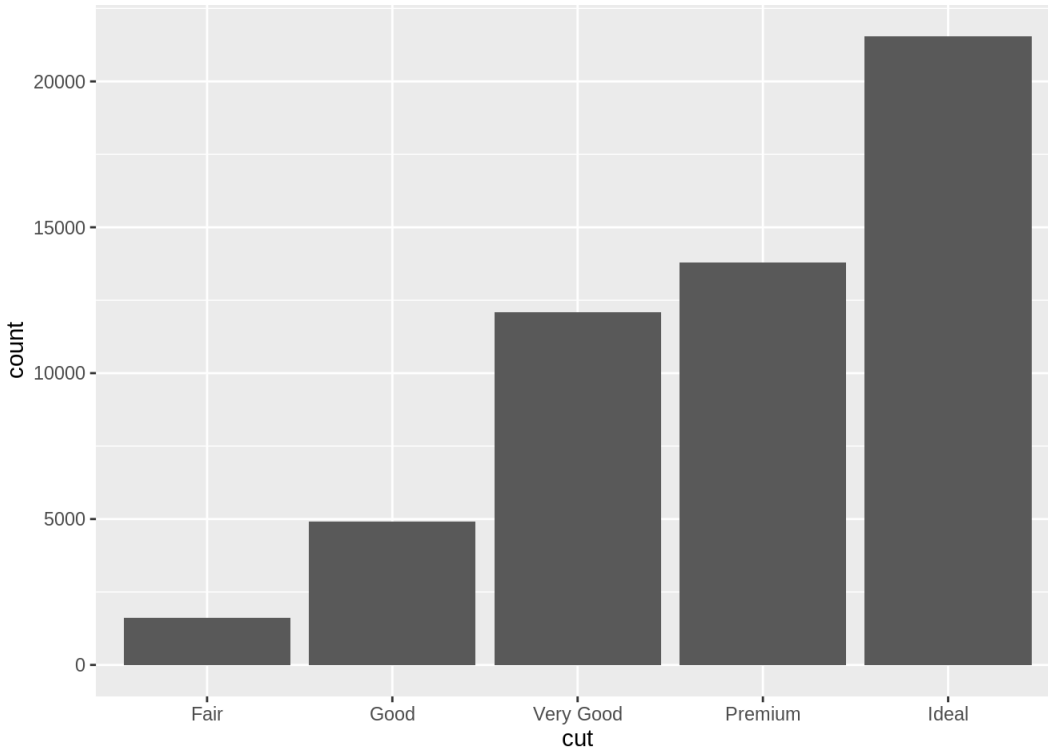
```
## $ price <int> 326, 326, 327, 334, 335, 336, 336, 337, 337, 338, 339, 340, 34...
```

```
## $ x <dbl> 3.95, 3.89, 4.05, 4.20, 4.34, 3.94, 3.95, 4.07, 3.87, 4.00, 4....
```

```
## $ y <dbl> 3.98, 3.84, 4.07, 4.23, 4.35, 3.96, 3.98, 4.11, 3.78, 4.05, 4....
```

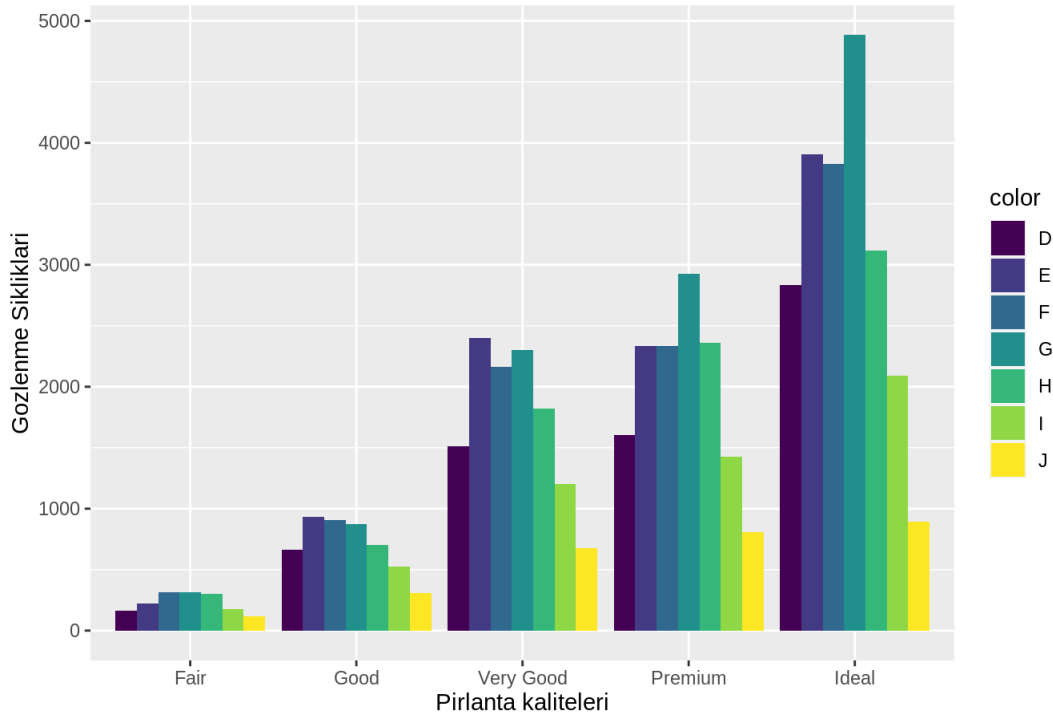
```
## $ z <dbl> 2.43, 2.31, 2.31, 2.63, 2.75, 2.48, 2.47, 2.53, 2.49, 2.39, 2....
```

```
ggplot(df, aes(cut)) +  
  geom_bar()
```



```
ggplot(df, aes(cut, fill = color)) +  
  geom_bar(position = position_dodge()) +  
  ggtitle("Bu yazı ana başlık yazıdır") +  
  xlab("Pirlanta kaliteleri") +  
  ylab("Gözlenme Sıklıkları")
```

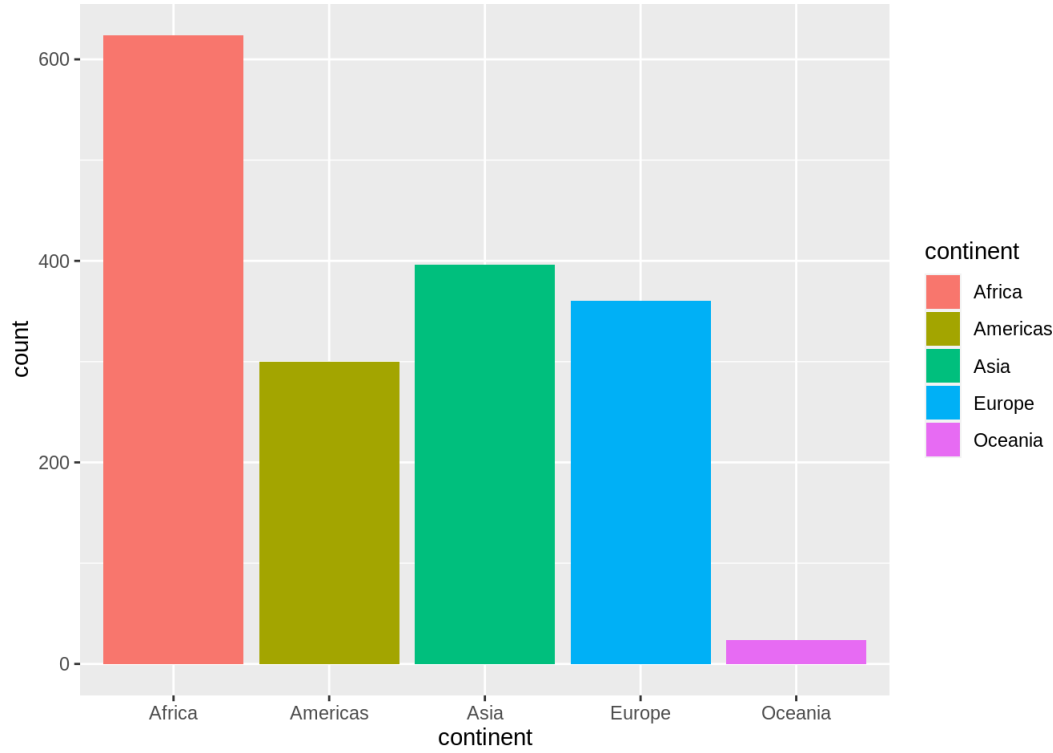
Bu yazı ana baslık yazididir



#örnek2

```
library(gapminder)
data("gapminder")
```

```
ggplot(gapminder, aes(x=continent, fill=continent)) +
  geom_bar()
```

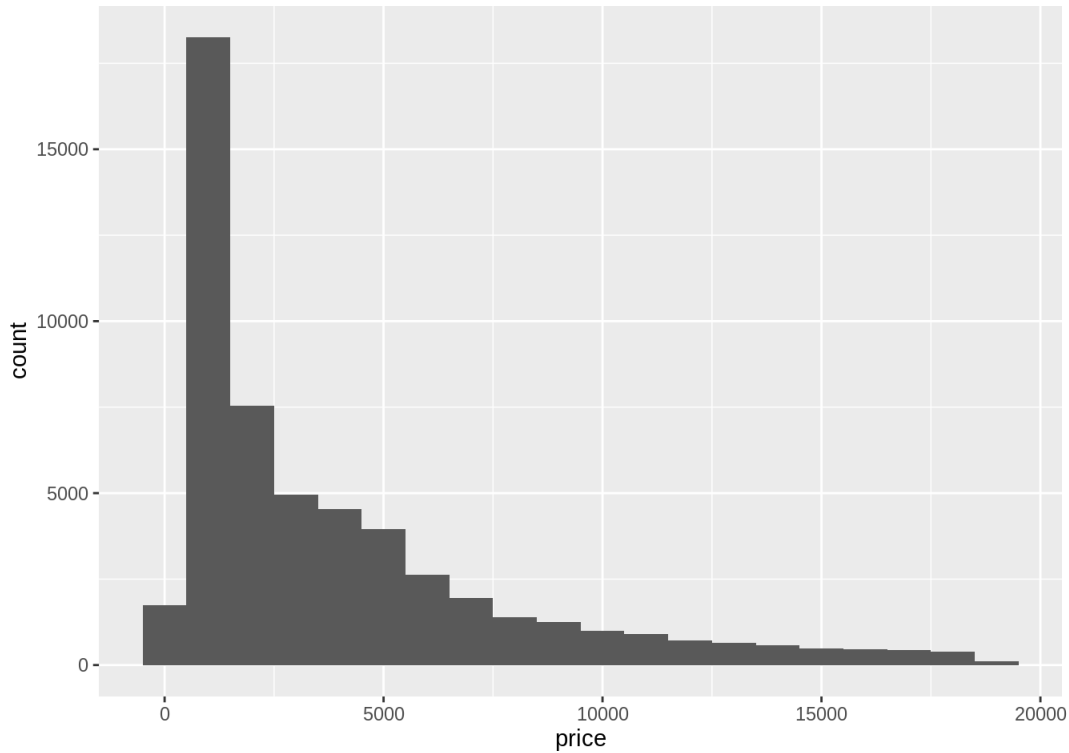


histogram

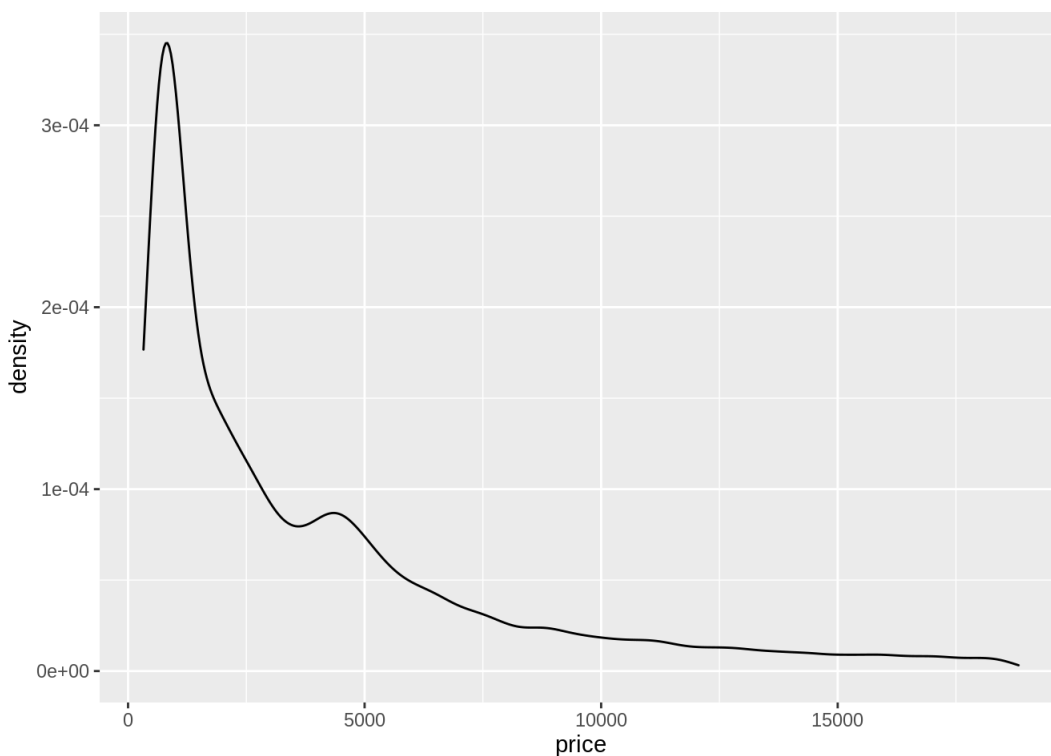
```
glimpse(df)
```

```
## Rows: 53,940
## Columns: 10
## $ carat <dbl> 0.23, 0.21, 0.23, 0.29, 0.31, 0.24, 0.24, 0.26, 0.22, 0.23, 0....
## $ cut <ord> Ideal, Premium, Good, Premium, Good, Very Good, Very Good, Ver...
## $ color <ord> E, E, E, I, J, J, I, H, E, H, J, J, F, J, E, E, I, J, J, J, I,...
## $ clarity <ord> SI2, SI1, VS1, VS2, SI2, VVS2, VVS1, SI1, VS2, VS1, SI1, VS1, ...
## $ depth <dbl> 61.5, 59.8, 56.9, 62.4, 63.3, 62.8, 62.3, 61.9, 65.1, 59.4, 64...
## $ table <dbl> 55, 61, 65, 58, 58, 57, 57, 55, 61, 61, 55, 56, 61, 54, 62, 58...
## $ price <int> 326, 326, 327, 334, 335, 336, 336, 337, 337, 338, 339, 340, 34...
## $ x <dbl> 3.95, 3.89, 4.05, 4.20, 4.34, 3.94, 3.95, 4.07, 3.87, 4.00, 4....
## $ y <dbl> 3.98, 3.84, 4.07, 4.23, 4.35, 3.96, 3.98, 4.11, 3.78, 4.05, 4....
## $ z <dbl> 2.43, 2.31, 2.31, 2.63, 2.75, 2.48, 2.47, 2.53, 2.49, 2.39, 2....
```

```
ggplot(df, aes(price)) + #histogram
  geom_histogram(binwidth = 1000)
```



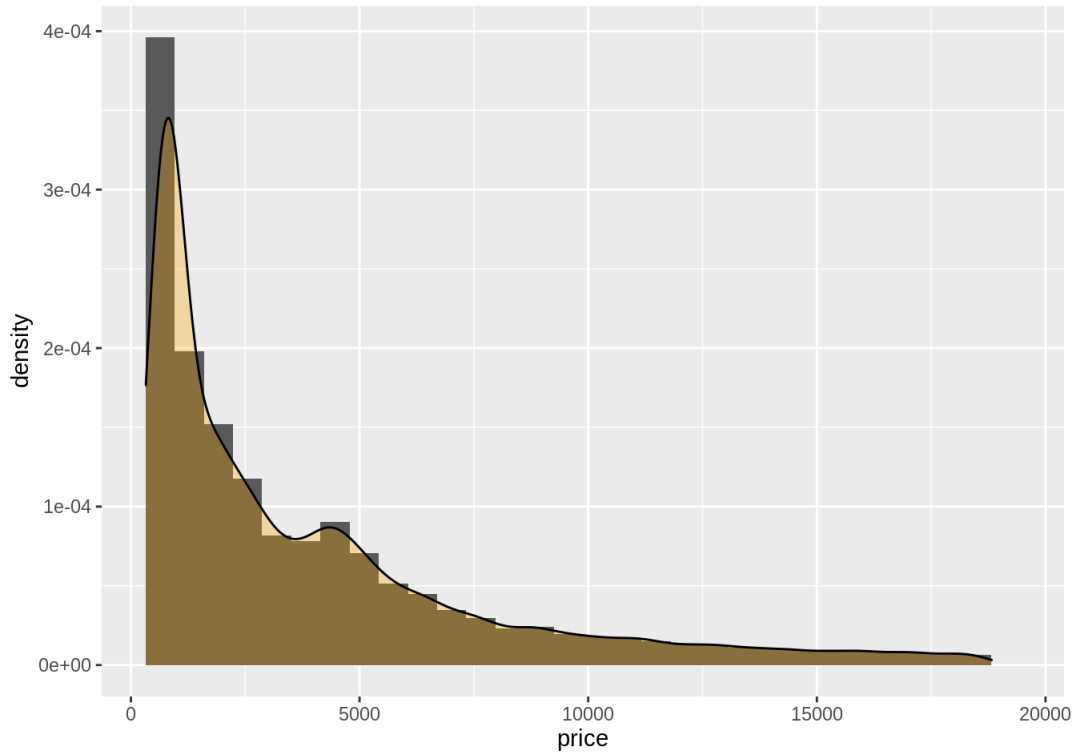
```
ggplot(df, aes(price)) + #yoğunluk
  geom_density()
```



```
ggplot(df, aes(price)) +  
  geom_histogram(aes(y = ..density..)) +  
  geom_density(alpha = .3, fill = "orange")
```

#beraber

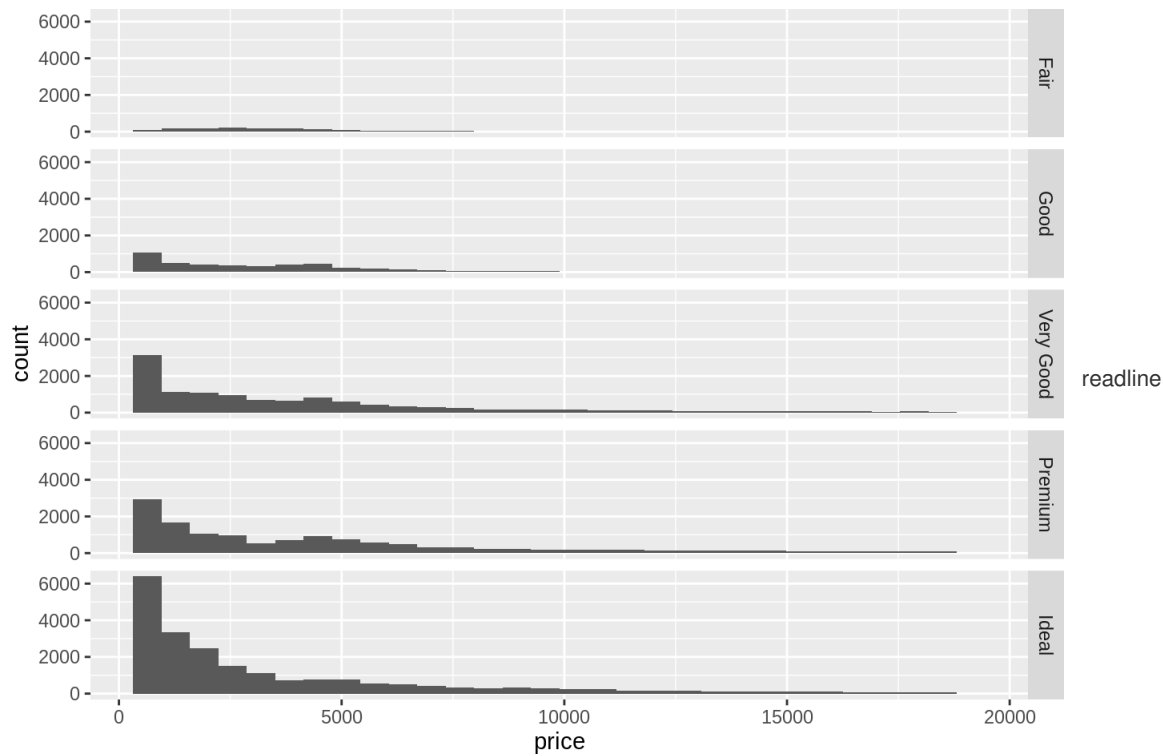
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
ggplot(df, aes(price)) +  
  geom_histogram() +  
  facet_grid(cut ~ .)
```

#ayrı ayrı verir

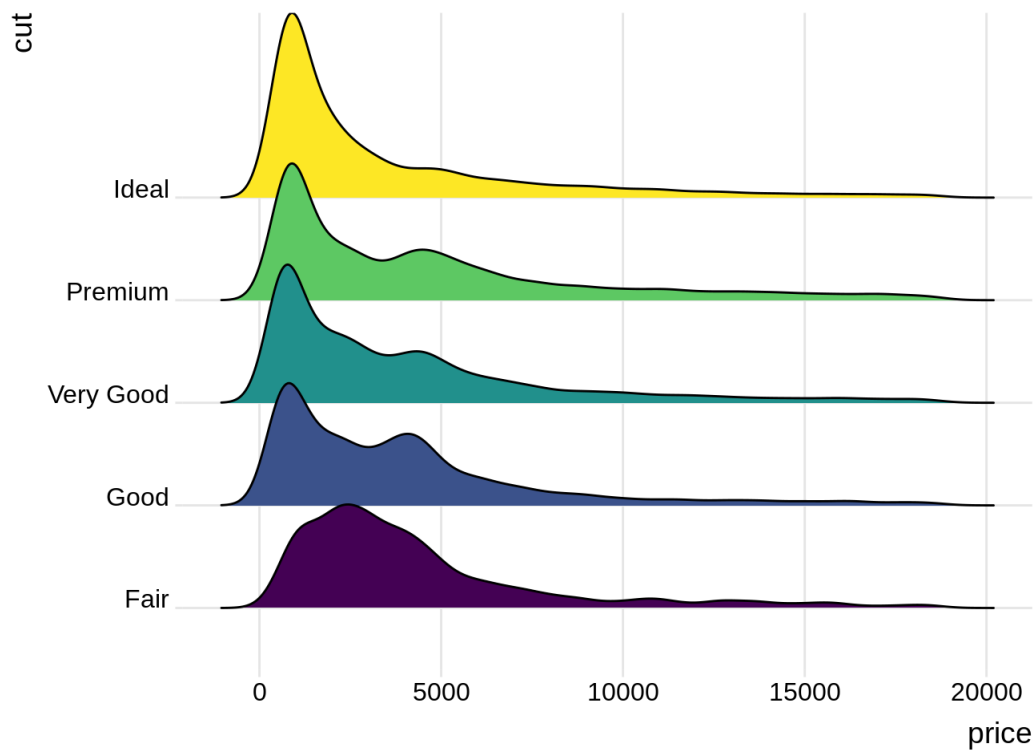
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
library(ggribes)
```

```
ggplot(df, aes(x = price, y = cut, fill = cut)) +  
  geom_density_ridges() +  
  theme_ridges() +  
  theme(legend.position = "none")
```

```
## Picking joint bandwidth of 458
```



```
library(plotly)
```

```
##  
## Attaching package: 'plotly'
```

```
## The following object is masked from 'package:Hmisc':  
##  
##  subplot
```

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

```
## The following object is masked from 'package:stats':  
##  
##  filter
```

```
## The following object is masked from 'package:graphics':  
##  
##  layout
```

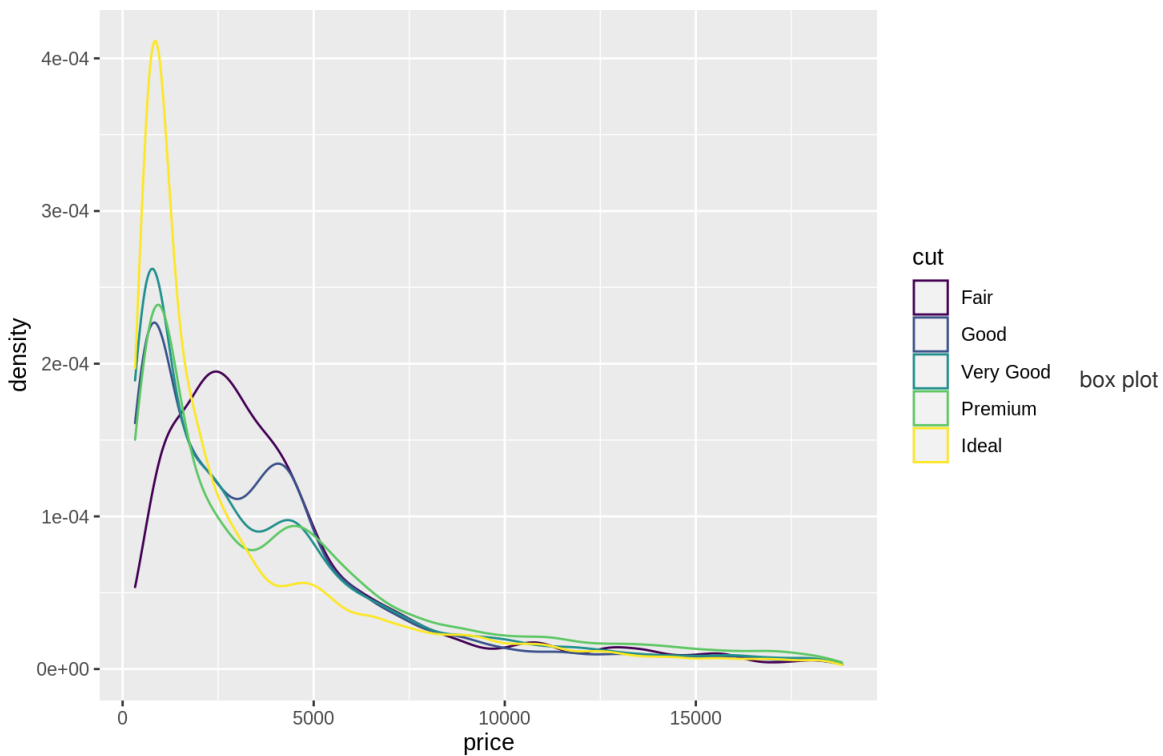
```
g <- plot_ly(x = rnorm(500), type = "histogram")  
g
```

```
plot_ly(x = rnorm(500), opacity = 0.6, type = "histogram") %>%
  add_trace(x = rnorm(500) + 1) %>%
  layout(barmode = "overlay")
```

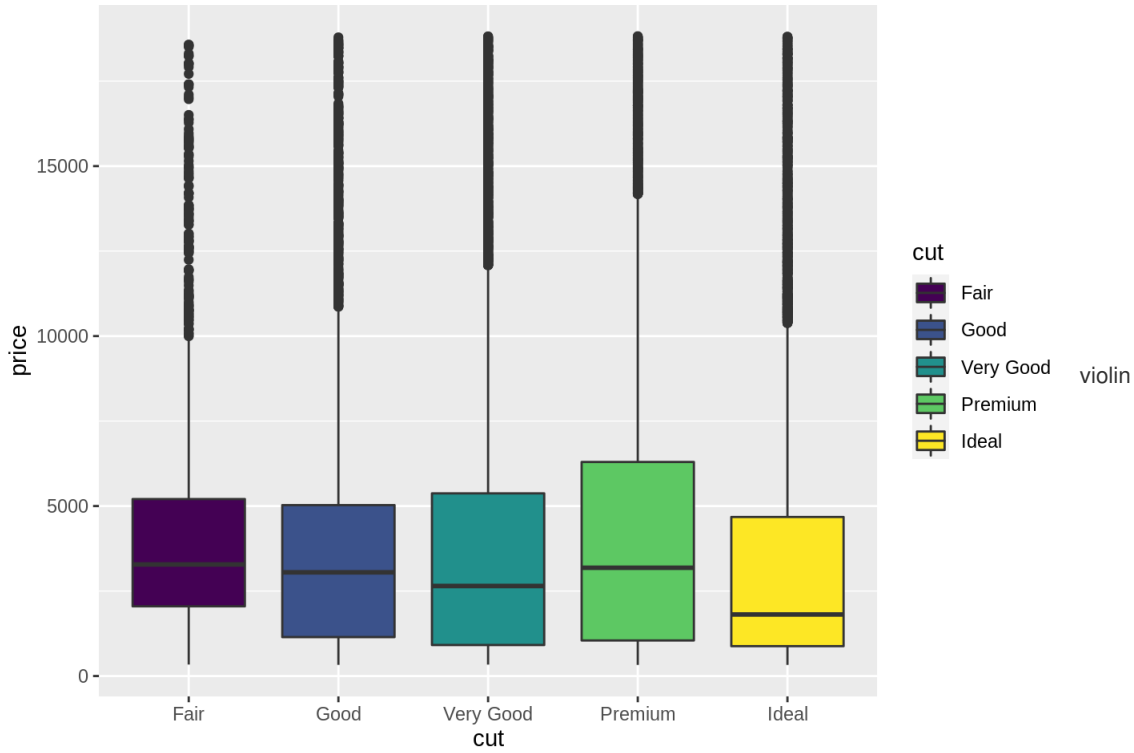
çoklu frekans

```
ggplot(df, aes(price, y = ..density..)) +
  geom_density(aes(colour = cut), binwidth = 500)
```

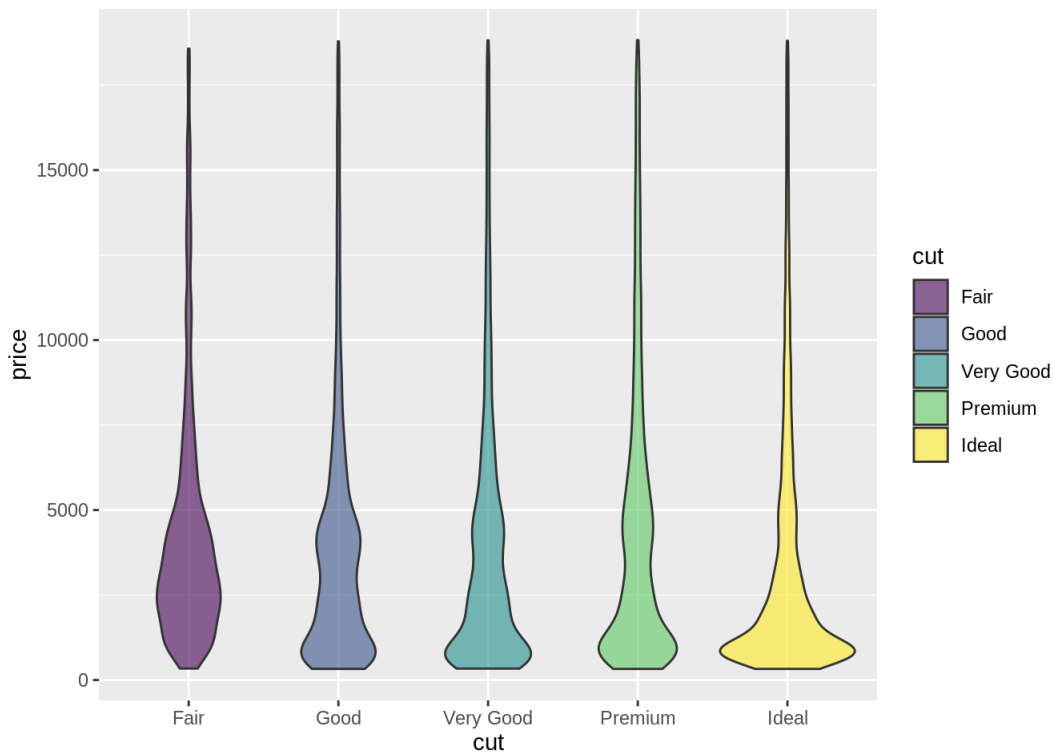
Warning: Ignoring unknown parameters: binwidth




```
ggplot(df, aes(x = cut, y = price, fill = cut)) +  
  geom_boxplot()
```



```
ggplot(df, aes(x = cut, y = price, fill = cut)) +  
  geom_violin(alpha = 0.6)
```



korelasyonların incelenmesi

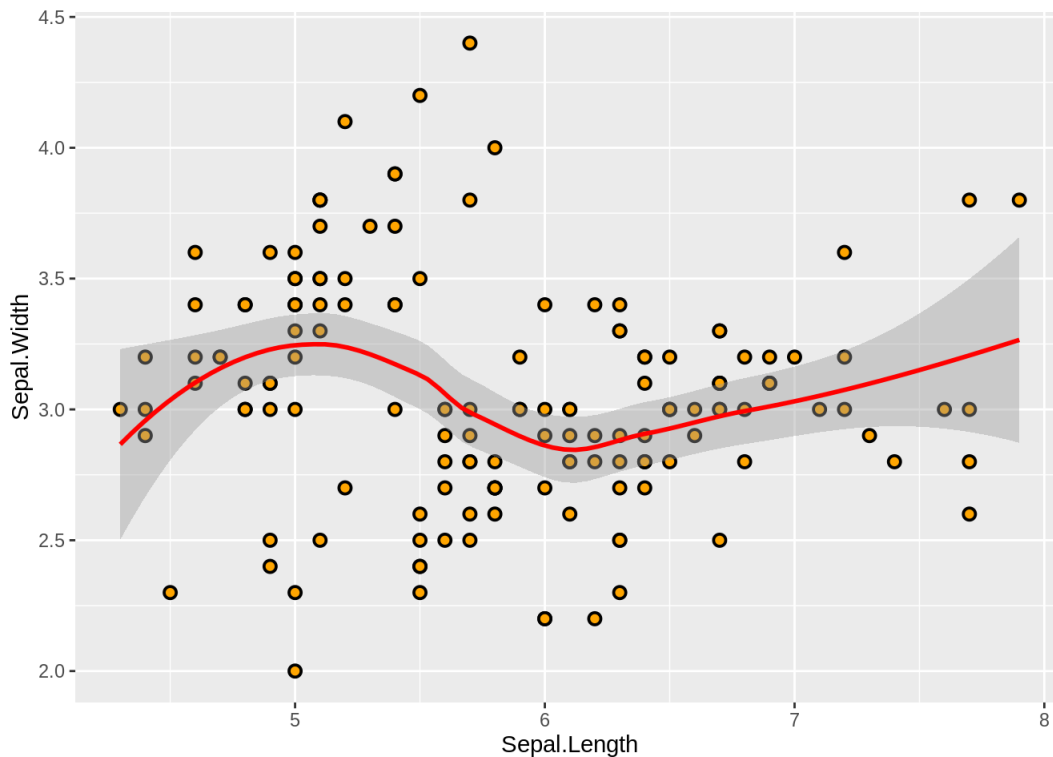
```
glimpse(iris)
```

```
## Rows: 150  
## Columns: 5  
## $ Sepal.Length <dbl> 5.1, 4.9, 4.7, 4.6, 5.0, 5.4, 4.6, 5.0, 4.4, 4.9, 5.4, 4....  
## $ Sepal.Width <dbl> 3.5, 3.0, 3.2, 3.1, 3.6, 3.9, 3.4, 3.4, 2.9, 3.1, 3.7, 3....  
## $ Petal.Length <dbl> 1.4, 1.4, 1.3, 1.5, 1.4, 1.7, 1.4, 1.5, 1.4, 1.5, 1.5, 1....  
## $ Petal.Width <dbl> 0.2, 0.2, 0.2, 0.2, 0.2, 0.4, 0.3, 0.2, 0.2, 0.1, 0.2, 0....  
## $ Species <fct> setosa, setosa, setosa, setosa, setosa, setosa, setosa, s....
```

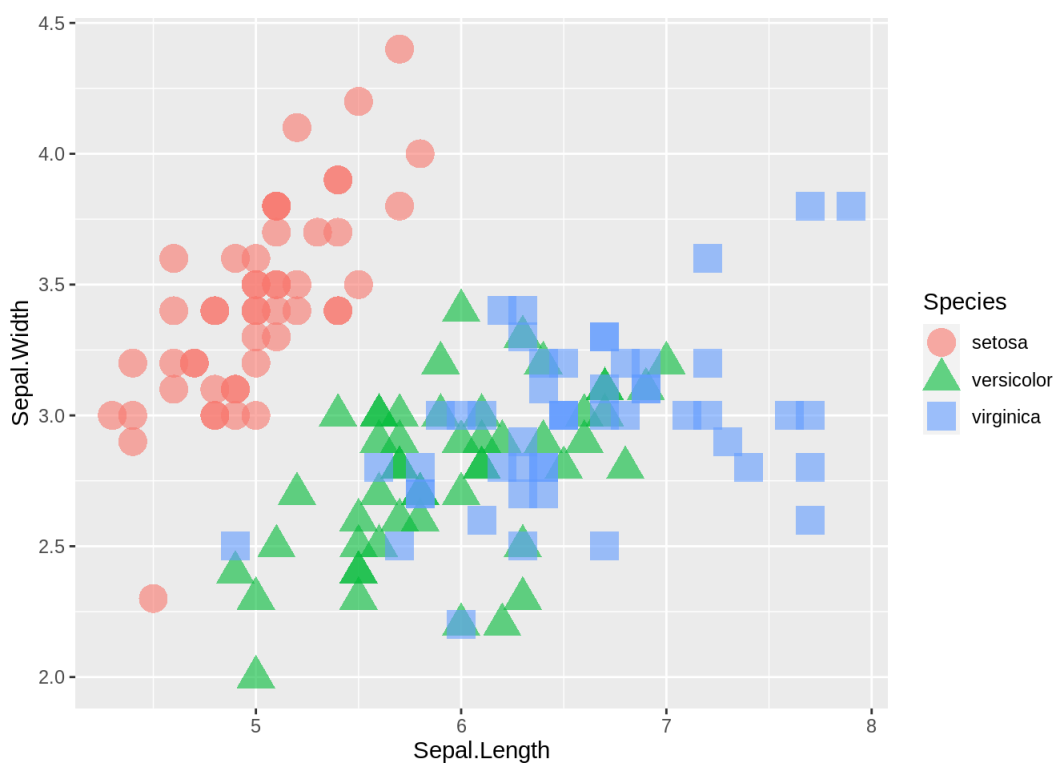
```
df1 <- iris
```

```
ggplot(df1, aes(Sepal.Length, Sepal.Width)) +
  geom_point(size = 2,
    shape = 21,
    stroke = 1,
    color = "black",
    fill = "orange") +
  geom_smooth(color = "red")
```

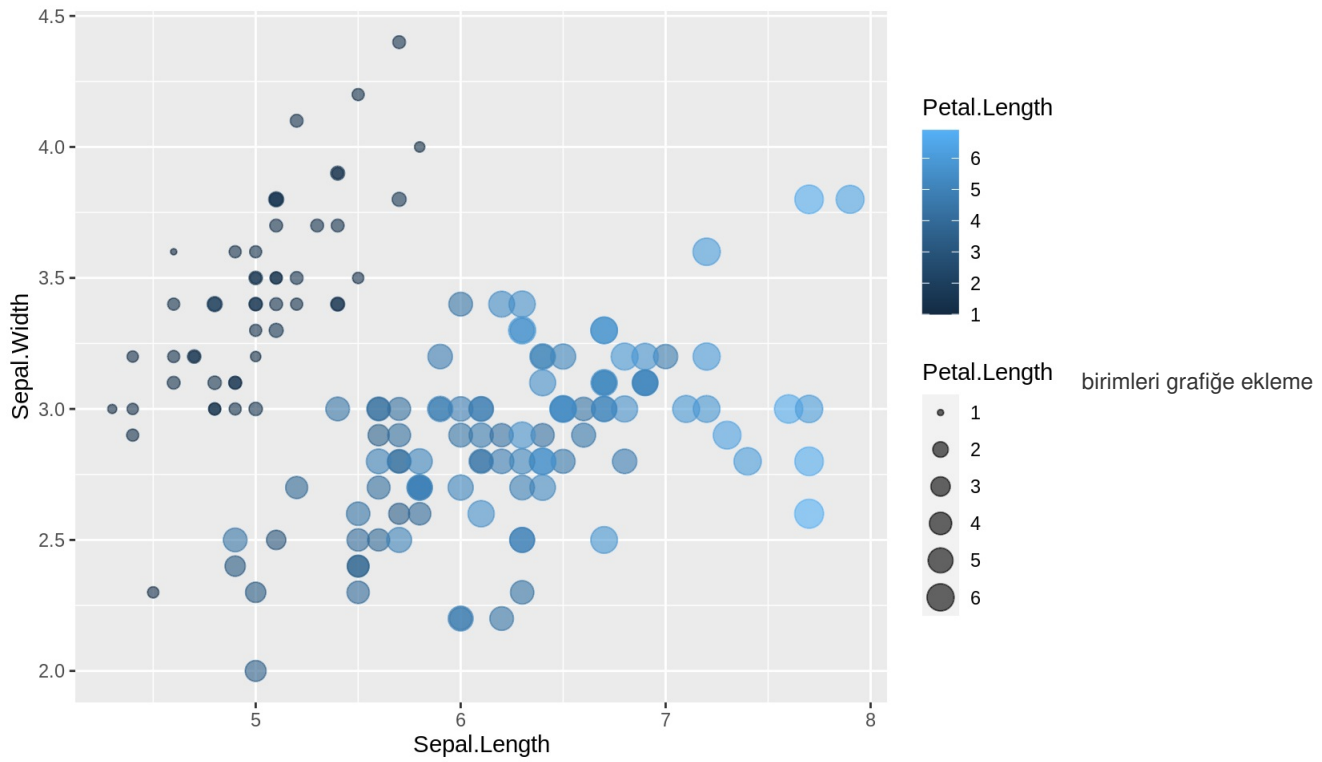
```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



```
ggplot(df1, aes(Sepal.Length, Sepal.Width, color = Species, shape = Species)) +
  geom_point(size = 6, alpha = 0.6)
```



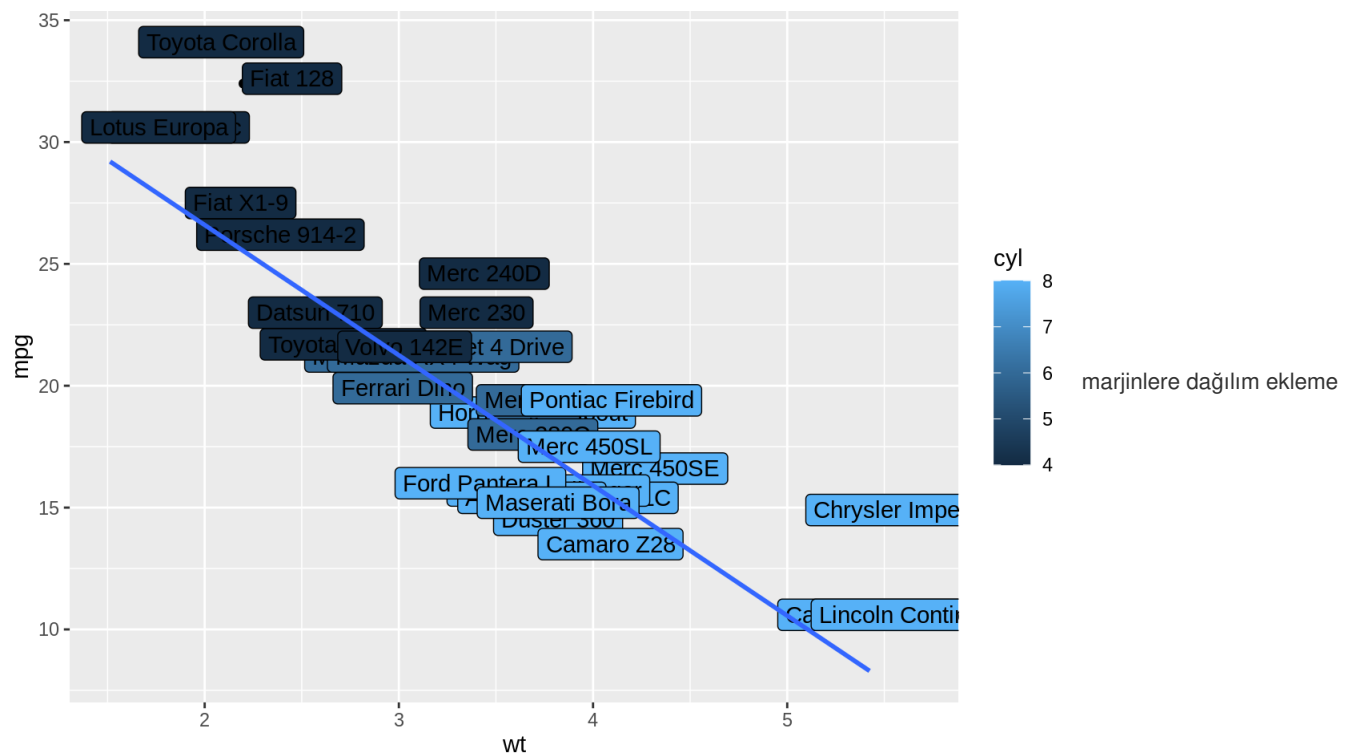
```
ggplot(df1, aes(Sepal.Length, Sepal.Width, color = Petal.Length, size = Petal.Length)) +
  geom_point(alpha = 0.6)
```



```
df <- mtcars

ggplot(df, aes(x = wt, y = mpg, fill = cyl)) +
  geom_point() +
  geom_label(label = rownames(df),
    nudge_x = 0.25,
    nudge_y = 0.2) +
  geom_smooth(method = lm, se = FALSE)
```

```
## `geom_smooth()` using formula 'y ~ x'
```



```
library(ggExtra)

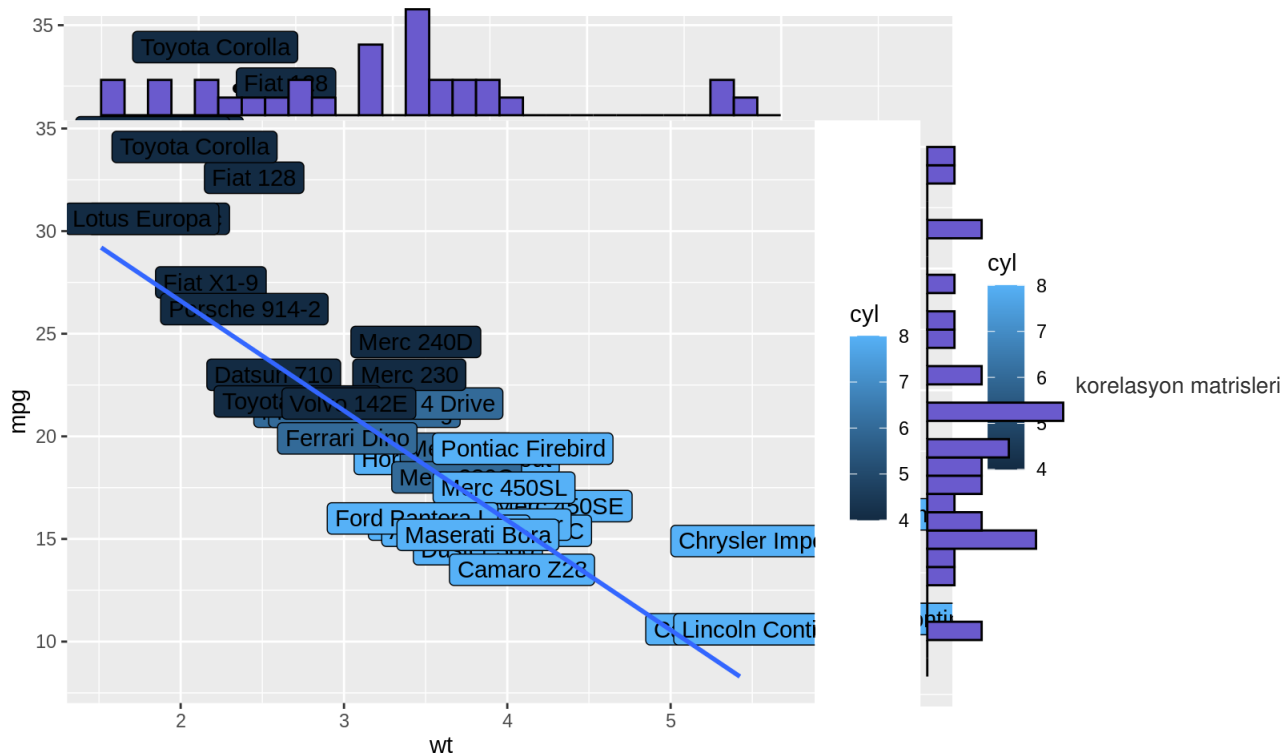
g <- ggplot(df, aes(x = wt, y = mpg, fill = cyl)) +
  geom_point() +
  geom_label(label = rownames(df),
    nudge_x = 0.25,
    nudge_y = 0.2) +
  geom_smooth(method = lm, se = FALSE)

g
```

```
## `geom_smooth()` using formula 'y ~ x'
```

```
ggMarginal(g, type = "histogram", fill = "slateblue")
```

```
## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'
```



```
library(GGally)
```

```
## Registered S3 method overwritten by 'GGally':
## method from
## +.gg ggplot2
```

```
##
## Attaching package: 'GGally'
```

```
## The following object is masked from 'package:funModeling':
##
## range01
```

```
df <- mtcars[, c(1,3:6)]
```

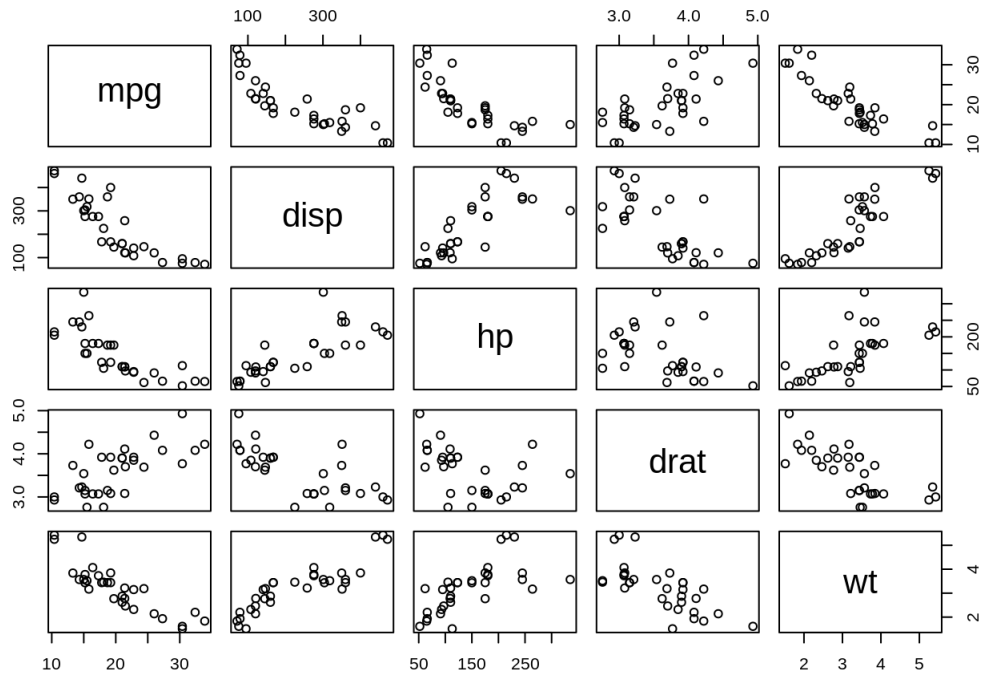
```
cor(df)
```

```
##      mpg    disp      hp    drat     wt
## mpg  1.000000 -0.847551 -0.776168  0.6811719 -0.8676594
## disp -0.847551  1.000000  0.7909486 -0.7102139  0.8879799
## hp   -0.776168  0.7909486  1.0000000 -0.4487591  0.6587479
## drat  0.6811719 -0.7102139 -0.4487591  1.0000000 -0.7124406
## wt   -0.8676594  0.8879799  0.6587479 -0.7124406  1.0000000
```

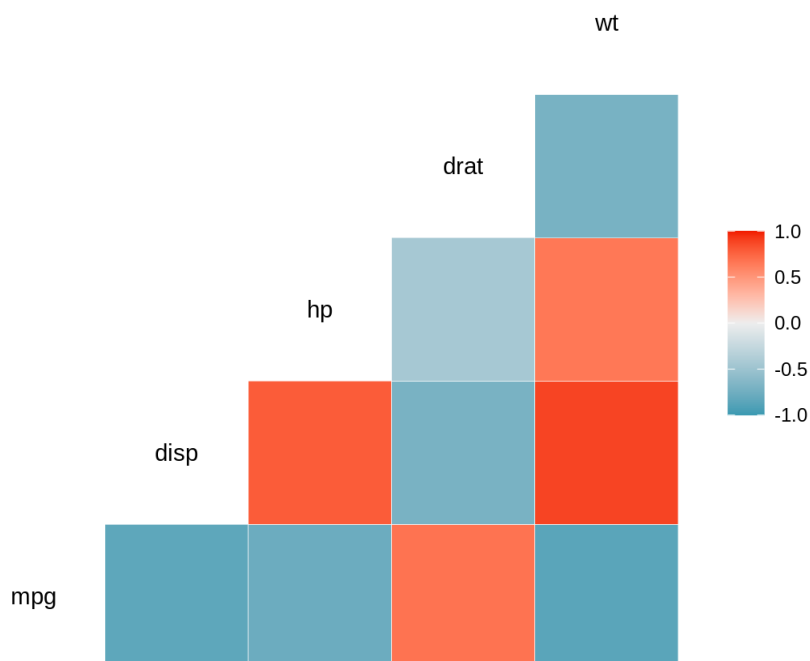
```
cor.test(df$mpg, df$wt)
```

```
##  
## Pearson's product-moment correlation  
##  
## data: df$mpg and df$wt  
## t = -9.559, df = 30, p-value = 1.294e-10  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
## -0.9338264 -0.7440872  
## sample estimates:  
##      cor  
## -0.8676594
```

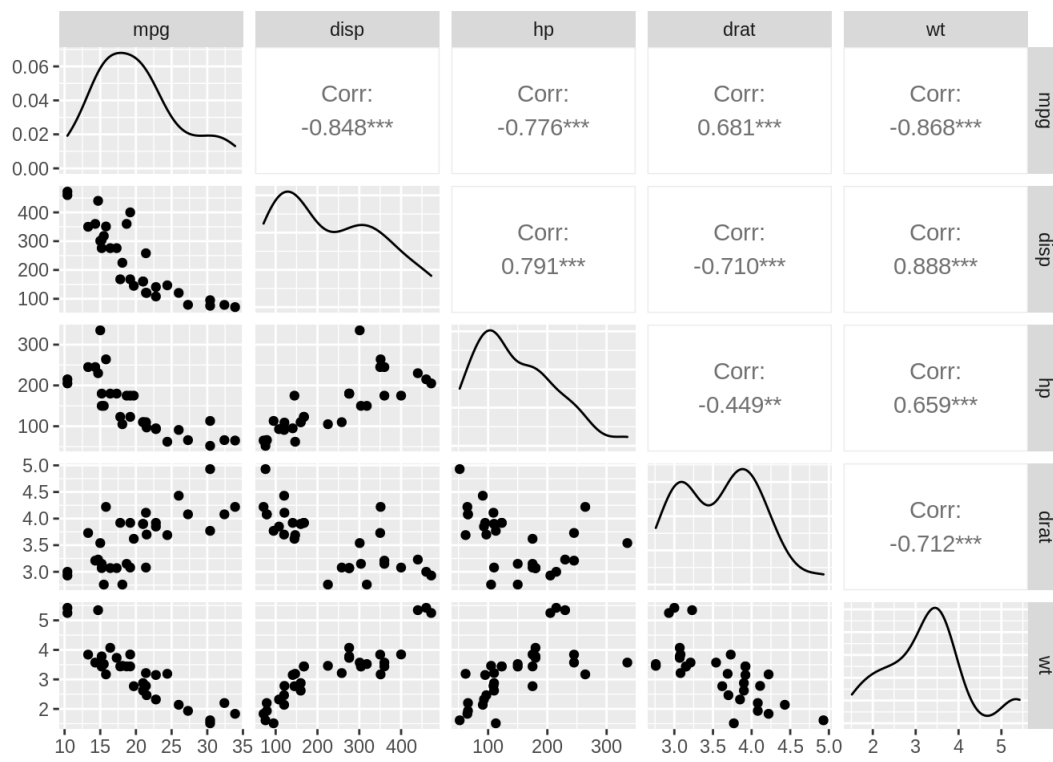
```
plot(df)      #Gli
```



```
ggcorr(df)
```



```
ggpairs(df)
```



zaman serisi görselleştirme

```
df <- economics
ggplot(df, aes(date, psavert)) +
  geom_line() +
  stat_smooth()
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

