

Simple operations

```
A = Int64[12, 3]
```

- `A=[12,3]`

- `print(A)`

```
y = 3×1 LinearAlgebra.Adjoint{Int64,Array{Int64,2}}:
      3
      4
      5
```

- `y = [3 4 5]'`

```
X = 3×2 LinearAlgebra.Adjoint{Int64,Array{Int64,2}}:
      1  1
      1  6
      1  7
```

- `X = [1 1 1; 1 6 7]'`

```
2×1 Array{Float64,2}:
 2.64516129032258
 0.29032258064516103
```

- `(X'X)^(-1)*X'y`

```
f (generic function with 1 method)
```

- `f(x) = 2x^3 + 3x^2 + 6`

```
34
```

- `f(2)`

Packages

```
• begin
•   import Pkg
•   Pkg.add("RDatasets")
•   Pkg.add("Econometrics")
•   Pkg.add("GLM")
•   Pkg.add("CSV")
•   Pkg.add("Gadfly")
•   Pkg.add("LinearAlgebra")
•   Pkg.add("Plots")
•   Pkg.add("PyPlot")
• end
```

```
• begin
•   Pkg.add("DataFrames")
• end
```

```
• begin
•   Pkg.add("CSV")
• end
```

Multiple definitions for CSV.

Combine all definitions into a single reactive cell using a `begin ... end` block.

```
• using CSV
```

Multiple definitions for CSV.

Combine all definitions into a single reactive cell using a `begin ... end` block.

```
• using DataFrames, GLM, Statistics, CSV, Econometrics
```

```
• using RDatasets
```

```
• let
•   Pkg.activate("Plots")
•   Pkg.add("Plots")
•   using Plots
• end
```

First seminar

psid =

	IntNum	PersNum	Age	Educate	Earnings	Hours	Kids	Married
1	4	4	39	12	77250	2940	2	CategoricalValue{Stri
2	4	6	35	12	12000	2040	2	CategoricalValue{Stri
3	4	7	33	12	8000	693	1	CategoricalValue{Stri
4	4	173	39	10	15000	1904	2	CategoricalValue{Stri
5	5	2	47	9	6500	1683	5	CategoricalValue{Stri
6	6	4	44	12	6500	2024	2	CategoricalValue{Stri
7	6	172	38	16	7000	1144	3	CategoricalValue{Stri
8	7	4	38	9	5000	2080	4	CategoricalValue{Stri
9	7	170	39	12	21000	2575	3	CategoricalValue{Stri
10	7	171	37	11	0	0	5	CategoricalValue{Stri
more								

```
• psid = dataset("Ecdat", "PSID")
```

```
Int32[39, 35, 33, 39, 47, 44, 38, 38, 39, 37, 48, 47, 40, 38, 41, 43, 41, 36,
```

```
• psid.Age
```

	variable	mean	min	median
1	:IntNum	4598.1	4	5464.0
2	:PersNum	59.2136	1	4.0
3	:Age	38.4629	30	38.0
4	:Educatn	16.3771	0	12.0
5	:Earnings	14244.5	0	11000.0
6	:Hours	1235.33	0	1517.0
7	:Kids	4.48126	0	2.0
8	:Married	nothing	CategoricalValue{String,UInt8} "marrie	nothing

• `describe(psid)`

	IntNum	PersNum	Age	Educatn	Earnings	Hours	Kids	Married
1	4	4	39	12	77250	2940	2	CategoricalValue{String,
2	4	6	35	12	12000	2040	2	CategoricalValue{String,
3	4	7	33	12	8000	693	1	CategoricalValue{String,
4	4	173	39	10	15000	1904	2	CategoricalValue{String,
5	5	2	47	9	6500	1683	5	CategoricalValue{String,
6	6	4	44	12	6500	2024	2	CategoricalValue{String,
7	6	172	38	16	7000	1144	3	CategoricalValue{String,

• `first(psid, 7)`

Model first seminar

```
model_a =
Continuous Response Model
Number of observations: 4855
Null Loglikelihood: -13840.79
Loglikelihood: -13838.29
R-squared: 0.0010
LR Test: 5.01 ~  $\chi^2(2) \Rightarrow \text{Pr} > \chi^2 = 0.0818$ 
Formula: log(1 + Earnings) ~ 1 + Age + Educatn
Variance Covariance Estimator: OIM
```

	PE	SE	t-value	Pr > t	2.50%	97.50%
(Intercept)	6.15726	0.421631	14.6034	<1e-46	5.33068	6.98385
Age	0.024032	0.0107385	2.23793	0.0253	0.00297964	0.0450844
Educatn	0.000165885	0.00325684	0.0509343	0.9594	-0.006219	0.00655077

```
• model_a = fit(EconometricModel, @formula(log(1+Earnings) ~ Age + Educatn), psid)
```

Second seminar

diamonds =

	Carat	Cut	Color
1	0.23	CategoricalValue{String,UInt8} "Ideal"	CategoricalValue{String,UInt8} "E
2	0.21	CategoricalValue{String,UInt8} "Premiu	CategoricalValue{String,UInt8} "E
3	0.23	CategoricalValue{String,UInt8} "Good"	CategoricalValue{String,UInt8} "E
4	0.29	CategoricalValue{String,UInt8} "Premiu	CategoricalValue{String,UInt8} "I
5	0.31	CategoricalValue{String,UInt8} "Good"	CategoricalValue{String,UInt8} "J
6	0.24	CategoricalValue{String,UInt8} "Very G	CategoricalValue{String,UInt8} "J
7	0.24	CategoricalValue{String,UInt8} "Very G	CategoricalValue{String,UInt8} "I
8	0.26	CategoricalValue{String,UInt8} "Very G	CategoricalValue{String,UInt8} "H
9	0.22	CategoricalValue{String,UInt8} "Fair"	CategoricalValue{String,UInt8} "E
10	0.23	CategoricalValue{String,UInt8} "Very G	CategoricalValue{String,UInt8} "H

more

```
• diamonds = dataset("ggplot2", "diamonds")
```

"diam.csv"

```
• CSV.write("diam.csv", diamonds)
```

	Cut	Table
1	CategoricalValue{String,UInt8} "Ideal"	55.0
2	CategoricalValue{String,UInt8} "Premiu	61.0
3	CategoricalValue{String,UInt8} "Good"	65.0
4	CategoricalValue{String,UInt8} "Premiu	58.0
5	CategoricalValue{String,UInt8} "Good"	58.0
6	CategoricalValue{String,UInt8} "Very G	57.0
7	CategoricalValue{String,UInt8} "Very G	57.0
8	CategoricalValue{String,UInt8} "Very G	55.0
9	CategoricalValue{String,UInt8} "Fair"	61.0
10	CategoricalValue{String,UInt8} "Very G	61.0

• `diamonds[!, [:Cut,:Table]]`

	Cut	Table
1	CategoricalValue{String,UInt8} "Premium"	61.0
2	CategoricalValue{String,UInt8} "Good"	65.0
3	CategoricalValue{String,UInt8} "Premium"	58.0
4	CategoricalValue{String,UInt8} "Good"	58.0

• `diamonds[2:5, [:Cut,:Table]]`

CategoricalArrays.CategoricalArray{String,1,UInt8,String,CategoricalArrays.CategoricalValue{String,UInt8}}

• `diamonds.Cut`

CategoricalArrays.CategoricalArray{String,1,UInt8,String,CategoricalArrays.CategoricalValue{String,UInt8}}

• `diamonds[!, :Cut]`

	variable	mean	min	median	
1	:Carat	0.79794	0.2	0.7	5.01
2	:Cut	nothing	CategoricalValue{String,UInt8} "Fair"	nothing	CategoricalValue{String,UInt8} "Fair"
3	:Color	nothing	CategoricalValue{String,UInt8} "D" (1/	nothing	CategoricalValue{String,UInt8} "D" (1/
4	:Clarity	nothing	CategoricalValue{String,UInt8} "I1" (1	nothing	CategoricalValue{String,UInt8} "I1" (1
5	:Depth	61.7494	43.0	61.8	79.0
6	:Table	57.4572	43.0	57.0	95.0
7	:Price	3932.8	326	2401.0	18823
8	:X	5.73116	0.0	5.7	10.74
9	:Y	5.73453	0.0	5.71	58.9
10	:Z	3.53873	0.0	3.53	31.8

• `describe(diamonds)`

Cut

```

          Cut
1      CategoricalValue{String,UInt8} "Fair"
2      CategoricalValue{String,UInt8} "Fair"
3      CategoricalValue{String,UInt8} "Ideal"
4      CategoricalValue{String,UInt8} "Premium"
5      CategoricalValue{String,UInt8} "Ideal"
6      CategoricalValue{String,UInt8} "Premium"
7      CategoricalValue{String,UInt8} "Fair"
8      CategoricalValue{String,UInt8} "Fair"
9      CategoricalValue{String,UInt8} "Very Good"
10     CategoricalValue{String,UInt8} "Very Good"

```

```
• diamonds[diamonds.X > 6, ["Cut"]] #отбор строк
```

Model

```

model = Continuous Response Model
Number of observations: 53940
Null Loglikelihood: -523775.49
Loglikelihood: -471866.68
R-squared: 0.8541
LR Test: 103817.61 ~  $\chi^2(4) \Rightarrow \text{Pr} > \chi^2 = 0.0000$ 
Formula: Price ~ 1 + Carat + X + Y + Z
Variance Covariance Estimator: OIM

```

	PE	SE	t-value	Pr > t	2.50%	97.50%
(Intercept)	1921.17	104.372	18.4069	<1e-74	1716.6	2125.74
Carat	10233.9	62.9361	162.608	<1e-99	10110.6	10357.3
X	-884.209	40.4701	-21.8485	<1e-99	-963.531	-804.887
Y	166.038	25.8582	6.42112	<1e-9	115.356	216.721
Z	-576.203	39.2819	-14.6684	<1e-47	-653.196	-499.211

```
• model = fit(EconometricModel, @formula(Price~1+Carat+X+Y+Z), diamonds)
```

```

5x5 LinearAlgebra.Hermitian{Float64,Array{Float64,2}}:
10893.6    6205.91   -2364.7    -59.1365   -540.0
 6205.91    3960.95   -1422.2    -20.4162  -310.451
-2364.7    -1422.2    1637.83   -589.677   -708.05
  -59.1365   -20.4162  -589.677   668.645   -107.215
 -540.0    -310.451   -708.05   -107.215   1543.07

```

```
• vcov(model)
```

```

5x5 LinearAlgebra.Hermitian{Float64,Array{Float64,2}}:
100962.0    67384.6   -4462.5    1093.29   -38171.3
 67384.6    45850.7   -5241.13    982.147  -22377.8
 -4462.5   -5241.13   41915.8   -10874.8  -47866.5
 1093.29    982.147  -10874.8   15301.4   -7711.91
-38171.3   -22377.8  -47866.5   -7711.91   105877.0

```

```
• vcov(model, HC1)
```

```
5x5 LinearAlgebra.Hermitian{Float64,Array{Float64,2}}:
 1.00954e5  67379.6   -4462.17   1093.21   -38168.5
 67379.6    45847.3   -5240.74    982.074  -22376.1
 -4462.17   -5240.74   41912.7   -10874.0  -47863.0
 1093.21     982.074  -10874.0   15300.3   -7711.33
 -38168.5   -22376.1  -47863.0   -7711.33   1.05869e5
```

```
• vcov(model, HC0) #Уайта
```

```
Float64[1921.17, 10233.9, -884.209, 166.038, -576.203]
```

```
• coef(model)
```

```
Float64[282.993, 388.72, 288.327, -28.2579, -58.9271, 213.943, 213.702, 129.117, 39.2819]
```

```
• residuals(model)
```

```
Float64[43.0065, -62.7202, 38.6735, 362.258, 393.927, 122.057, 122.298, 207.883, -50.8691]
```

```
• fitted(model)
```

	PE	SE	t-value	Pr > t	2.50%	97.50%
(Intercept)	1921.17	104.372	18.4069	<1e-74	1716.6	2125.74
Carat	10233.9	62.9361	162.608	<1e-99	10110.6	10357.3
X	-884.209	40.4701	-21.8485	<1e-99	-963.531	-804.887
Y	166.038	25.8582	6.42112	<1e-9	115.356	216.721
Z	-576.203	39.2819	-14.6684	<1e-47	-653.196	-499.211

```
• coeftable(model)
```

UndefVarError: diag not defined

```
1. top-level scope @ (Local: 1
```

```
• confint(model, level=0.9, se=sqrt.(diag(vcov(model, HC1))))
```

Tests

```
943745.3697994003
```

```
• aic(model)
```

```
943798.7435649805
```

```
• bic(model)
```

```
Int32[106276, 106276, 106929, 111556, 112225, 112896, 112896, 113569, 113569, 114240]
```

```
• diamonds[!, "plot2"] = diamonds[!, :Price] .^2
```

```
Float64[80085.3, 1.51103e5, 83132.2, 798.512, 3472.41, 45771.7, 45668.8, 16671.1, 1.05869e5]
```

```
• diamonds[!, "resid2"] = residuals(model) .^2
```

```

aux_model =
Continuous Response Model
Number of observations: 53940
Null Loglikelihood: -938649.49
Loglikelihood: -930453.77
R-squared: 0.2621
LR Test: 16391.44 ~  $\chi^2(5) \Rightarrow \text{Pr} > \chi^2 = 0.0000$ 
Formula: resid2 ~ 1 + Carat + X + (Carat ^ 2) + (X ^ 2) + X & Carat
Variance Covariance Estimator: OIM

```

	PE	SE	t-value	Pr > t	2.50%	97.50%
(Intercept)	-2.6743e7	1.50694e6	-17.7466	<1e-69	-2.96966e7	-2.37894e7
Carat	2.76898e7	1.22161e6	22.6667	<1e-99	2.52954e7	3.00842e7
X	3.50164e6	655055.0	5.34557	<1e-7	2.21773e6	4.78556e6
Carat ^ 2	3.18021e7	4.45418e5	71.3984	<1e-99	3.09291e7	3.26752e7
X ^ 2	9.90712e5	86860.2	11.4058	<1e-29	820465.0	1.16096e6
X & Carat	-1.45887e7	2.84973e5	-51.1933	<1e-99	-1.51472e7	-1.40301e7

```

• aux_model = fit(EconometricModel,
@formula(resid2~Carat+X+Carat^2+X^2+X*Carat),diamonds)

```

Graphics

```
iris =
```

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
1	5.1	3.5	1.4	0.2	CategoricalValue{String,UInt8} "
2	4.9	3.0	1.4	0.2	CategoricalValue{String,UInt8} "
3	4.7	3.2	1.3	0.2	CategoricalValue{String,UInt8} "
4	4.6	3.1	1.5	0.2	CategoricalValue{String,UInt8} "
5	5.0	3.6	1.4	0.2	CategoricalValue{String,UInt8} "
6	5.4	3.9	1.7	0.4	CategoricalValue{String,UInt8} "
7	4.6	3.4	1.4	0.3	CategoricalValue{String,UInt8} "
8	5.0	3.4	1.5	0.2	CategoricalValue{String,UInt8} "
9	4.4	2.9	1.4	0.2	CategoricalValue{String,UInt8} "
10	4.9	3.1	1.5	0.1	CategoricalValue{String,UInt8} "

more

```
• iris = dataset("datasets", "iris")
```

```

• let
•   Pkg.activate("Plots")
•   Pkg.add("Plots")
•   using Plots
• end

```

```
• using Plots
```


A histogram showing the frequency distribution of the variable `y1`. The x-axis represents the values of `y1` (ranging from 4 to 8), and the y-axis represents the frequency (ranging from 0 to 30). The distribution is unimodal and slightly right-skewed, with the highest frequency occurring at `y1 = 6`.

Bin Range	Frequency
4.0 - 4.5	4
4.5 - 5.0	18
5.0 - 5.5	30
5.5 - 6.0	31
6.0 - 6.5	32
6.5 - 7.0	22
7.0 - 7.5	7
7.5 - 8.0	6

A scatter plot showing the relationship between x_1 (x-axis) and y_1 (y-axis). The x-axis ranges from approximately 4.5 to 7.5, and the y-axis ranges from 1 to 7. The data points are blue circles with black outlines. The plot shows a clear positive linear trend, with a dense cluster of points between x_1 values of 5.5 and 7.0, and y_1 values of 4.0 and 6.0. There are also several points at lower y_1 values (around 1.0 to 2.0) for x_1 values between 4.5 and 5.5.

localhost:1234/edit?id=85e9cbd2-6ad1-11eb-048c-adff40c94b0a

• *Enter cell code...*