

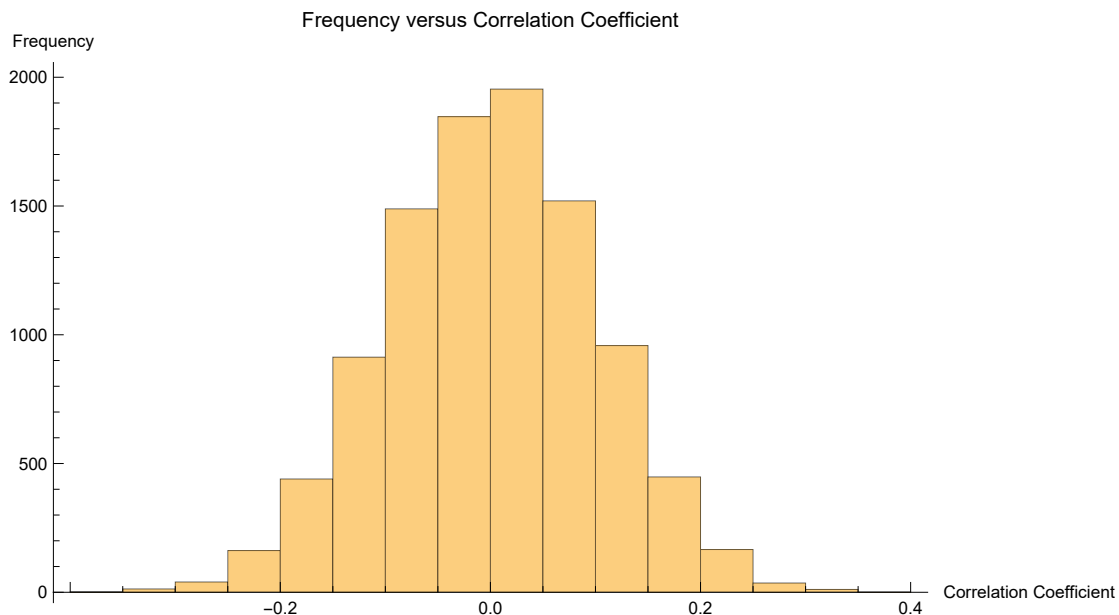
Fooled By Correlation

Correlation is a stochastic variable. The mean may be zero. Very often correlation is not even correlation.

[Image]

```
Histogram[Table[X = RandomVariate[NormalDistribution[], 100];  
  Y = RandomVariate[NormalDistribution[], 100];  
  Correlation[X, Y], {10^4}], AxesLabel → {"Correlation Coefficient", "Frequency"},  
  ImageSize → Large, PlotLabel → "Frequency versus Correlation Coefficient"]
```

Out[]=

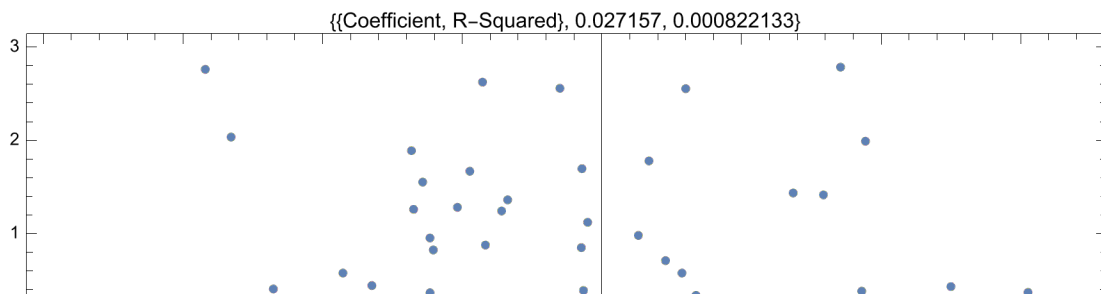


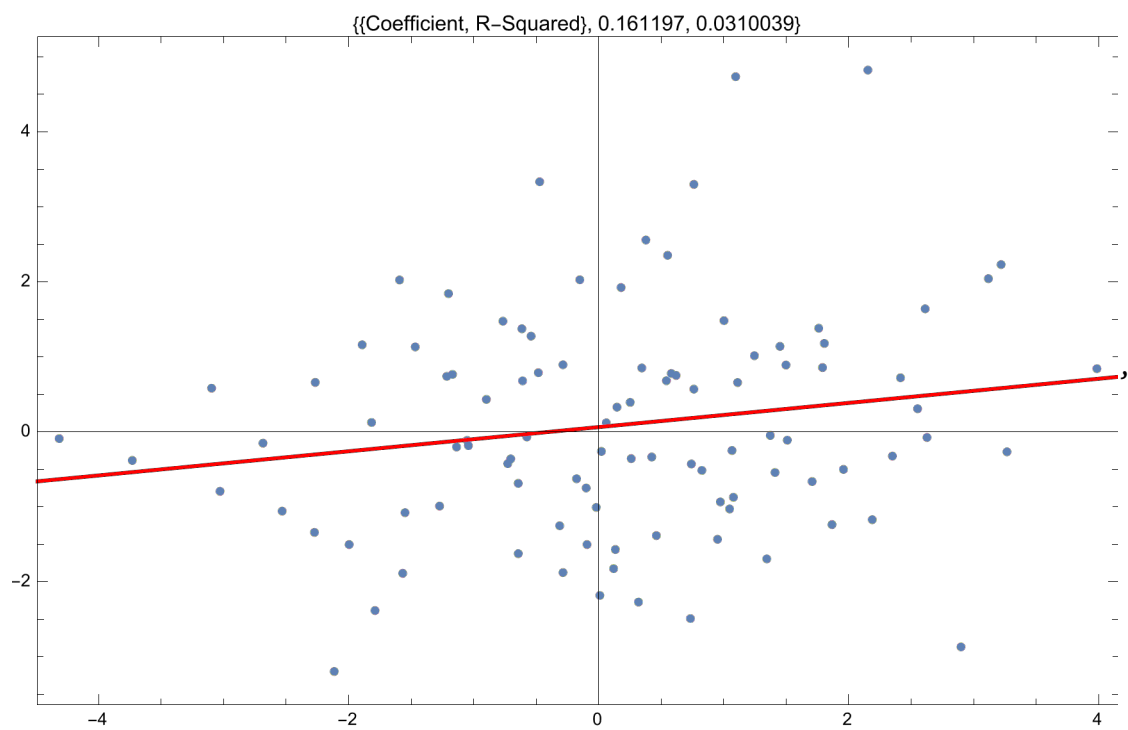
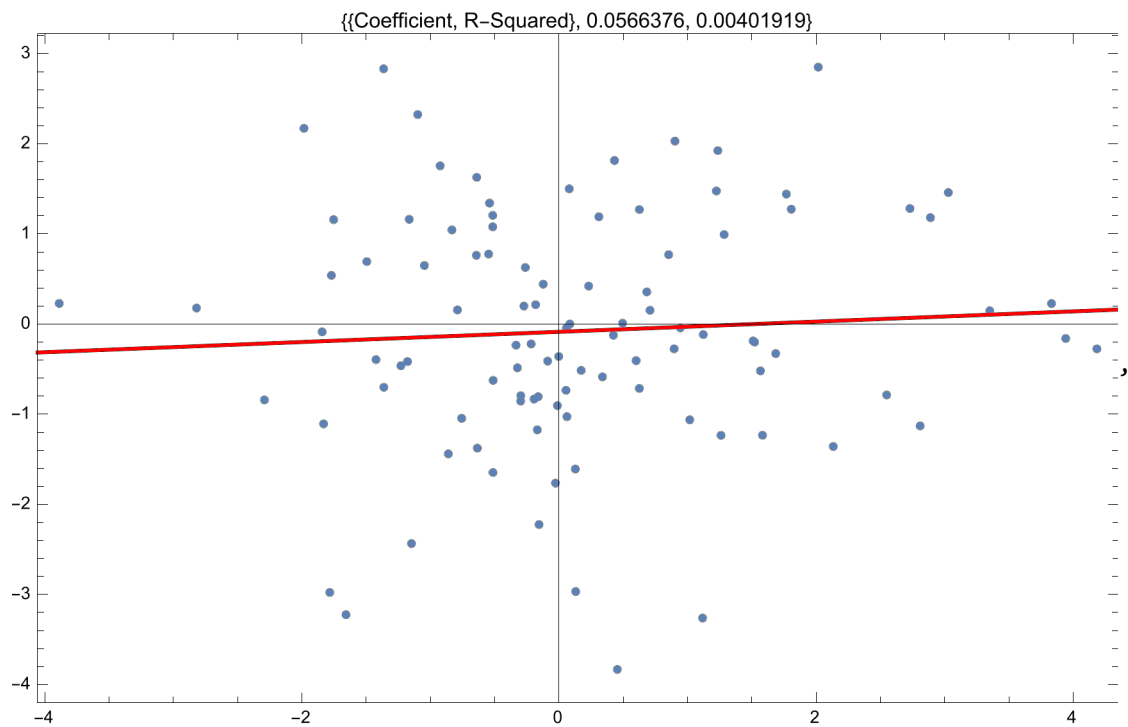
Comments

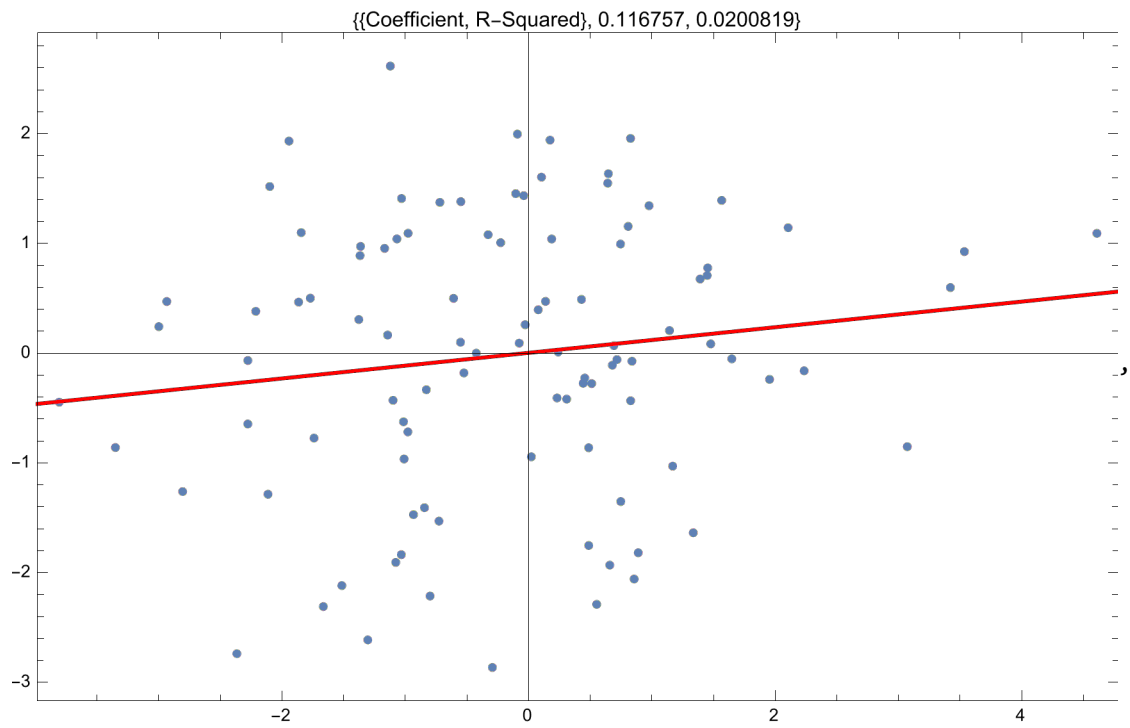
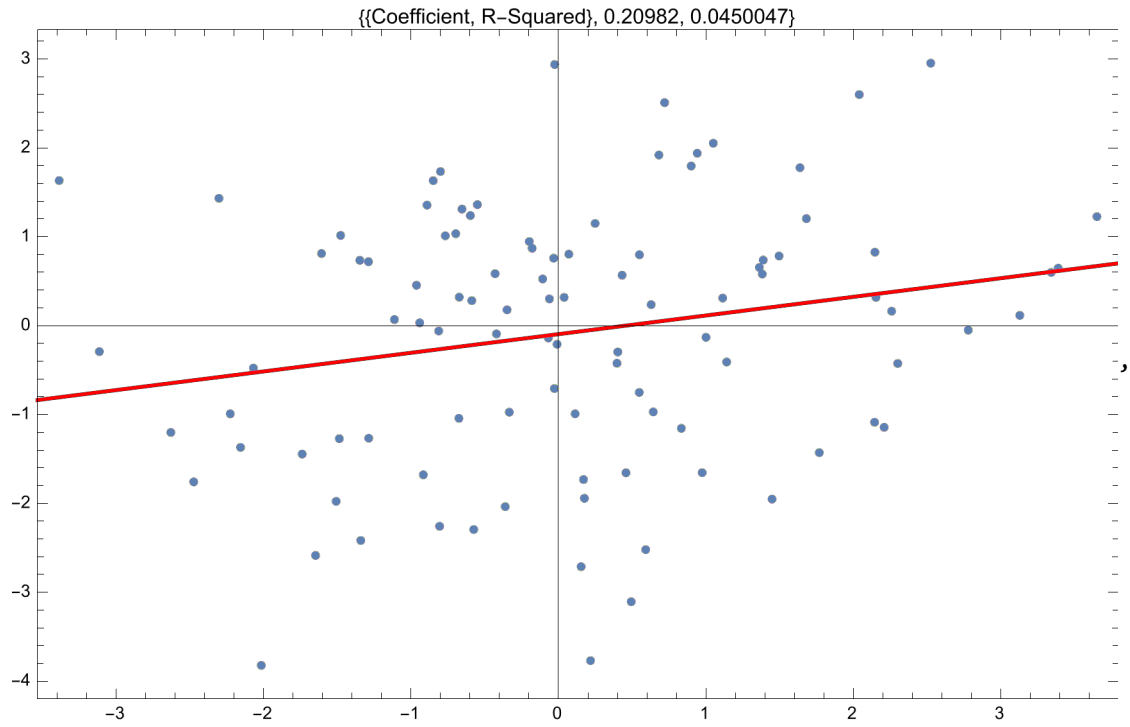
Correlations of from 0 - 0.3 is closer to zero. Entropy methods are superior.

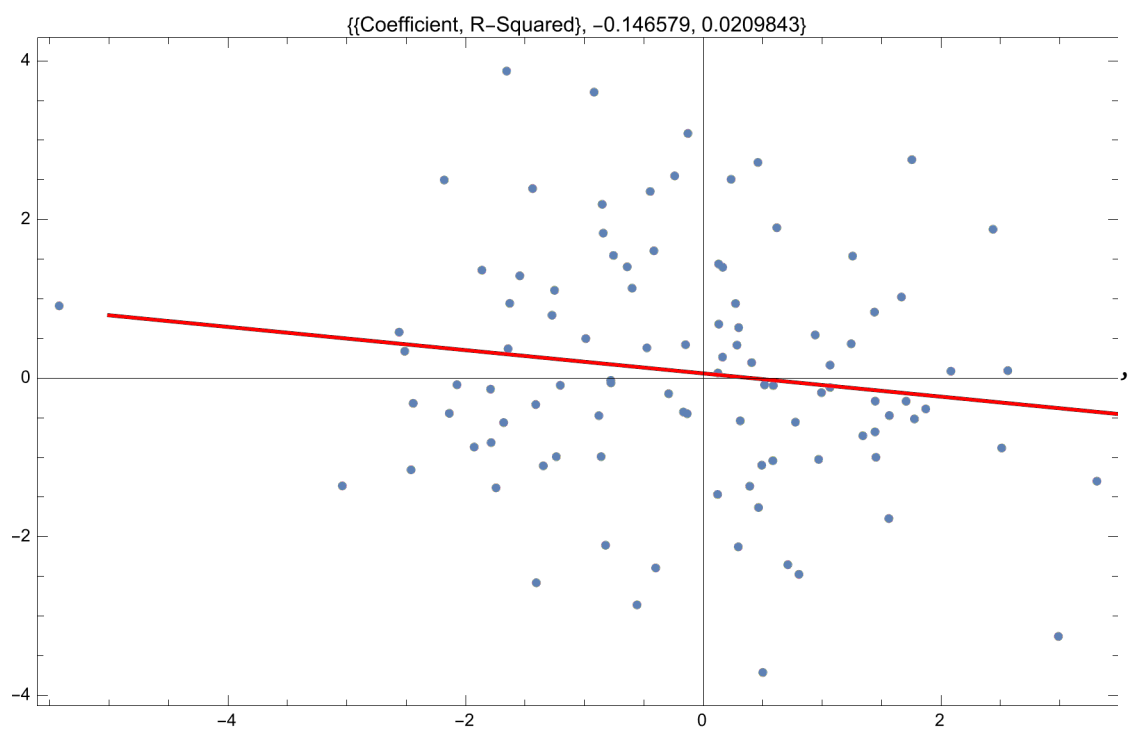
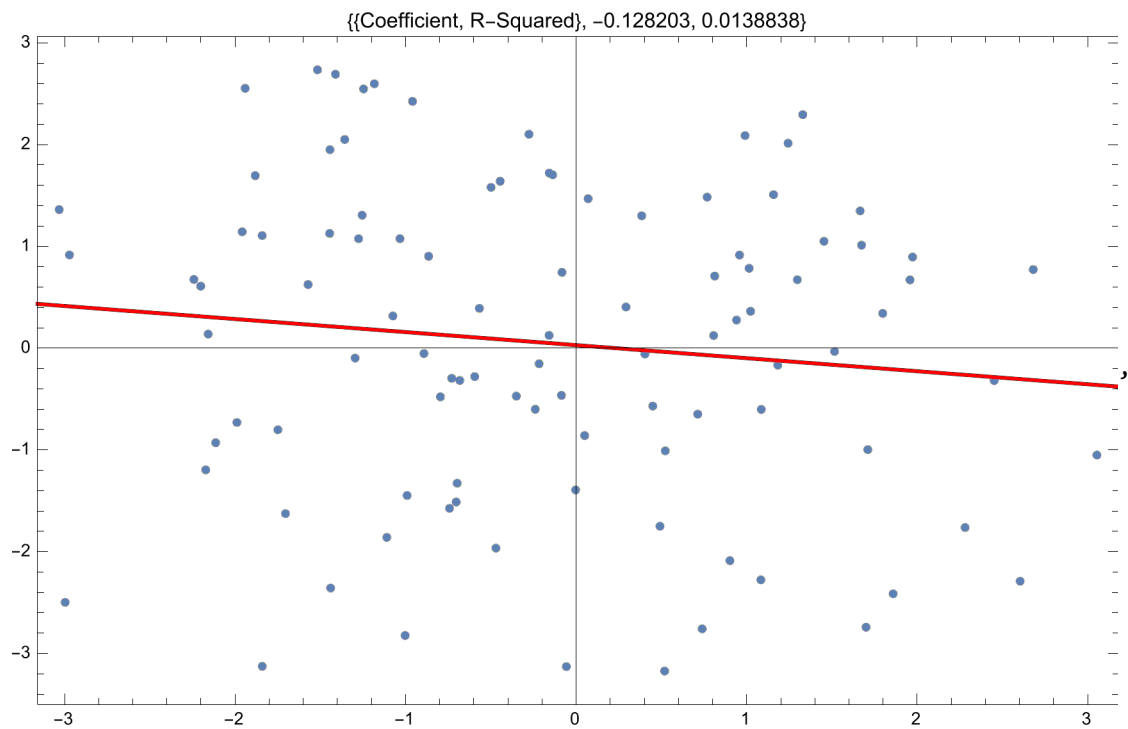
```
Table[  
  data = Table[{RandomVariate[NormalDistribution[RandomVariate[NormalDistribution[]], 1]],  
    RandomVariate[NormalDistribution[RandomVariate[NormalDistribution[]], 1]]}, {100}];  
  lm = LinearModelFit[data, x, x];  
  Show[ListPlot[data, PlotStyle → PointSize[Medium], ColorFunction → Blue,  
    ImageSize → Large], Plot[lm[x], {x, -5, 5}, PlotStyle → Red], Frame → True,  
  PlotLabel → {"{Coefficient, R-Squared}", (lm // Normal) [[2]] [[1]], lm["RSquared"]}], {12}]
```

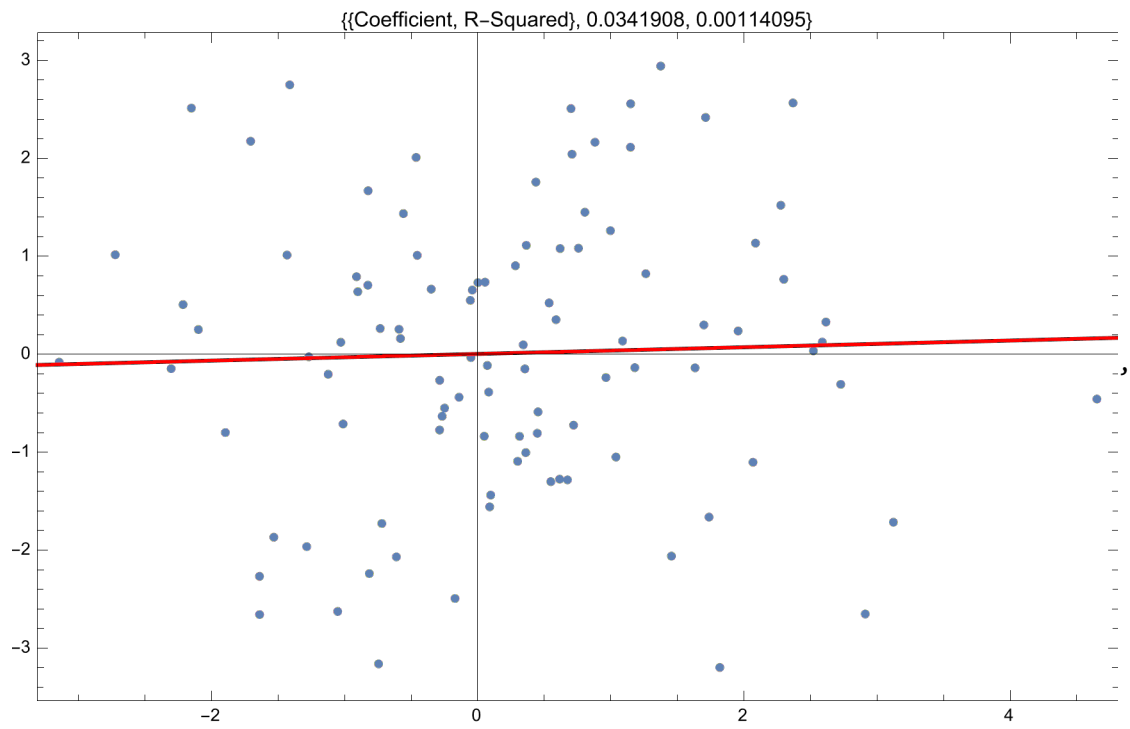
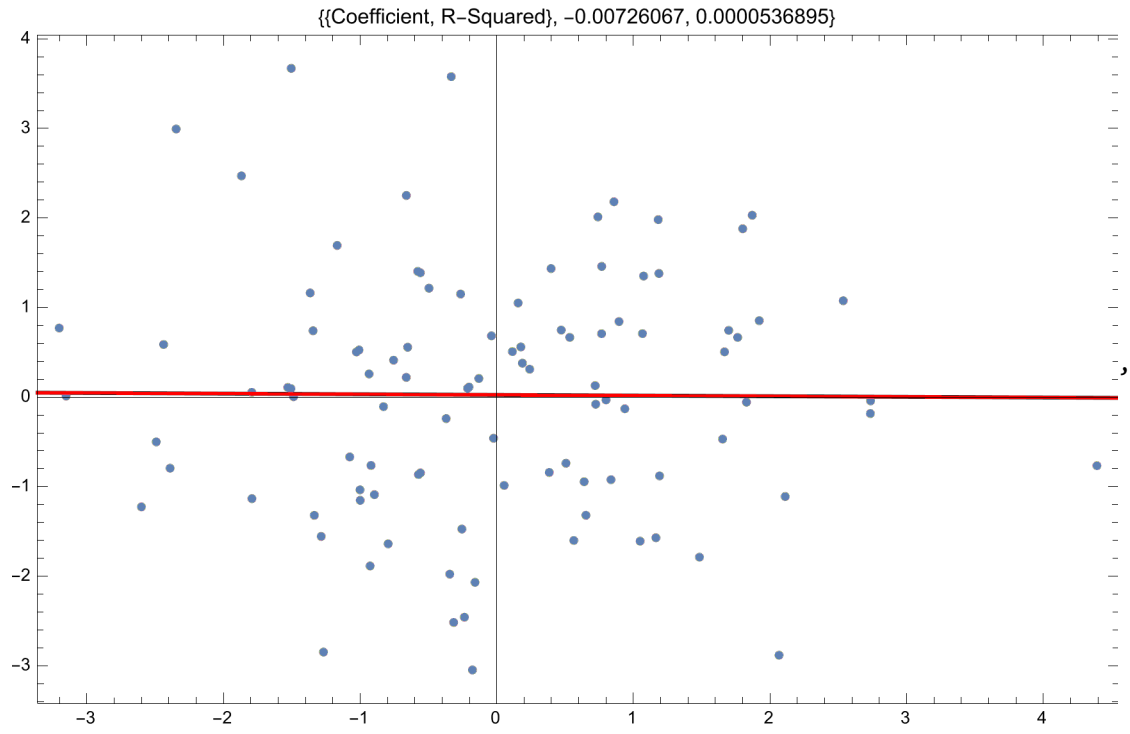
Out[]=

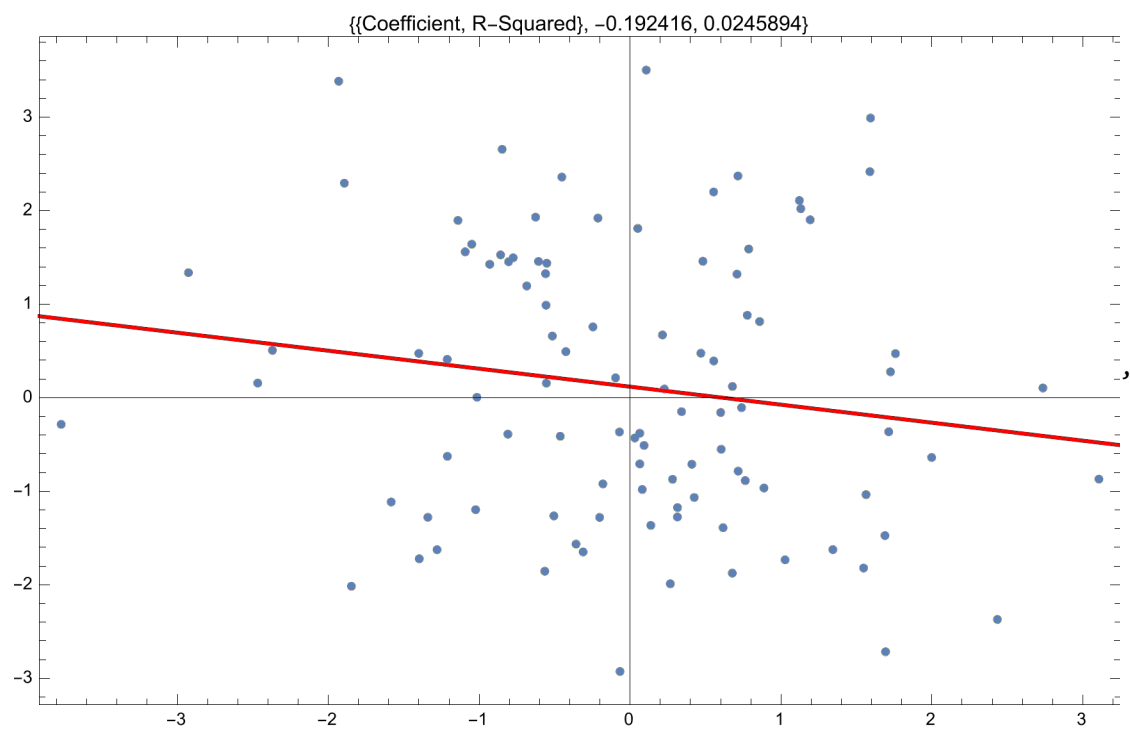
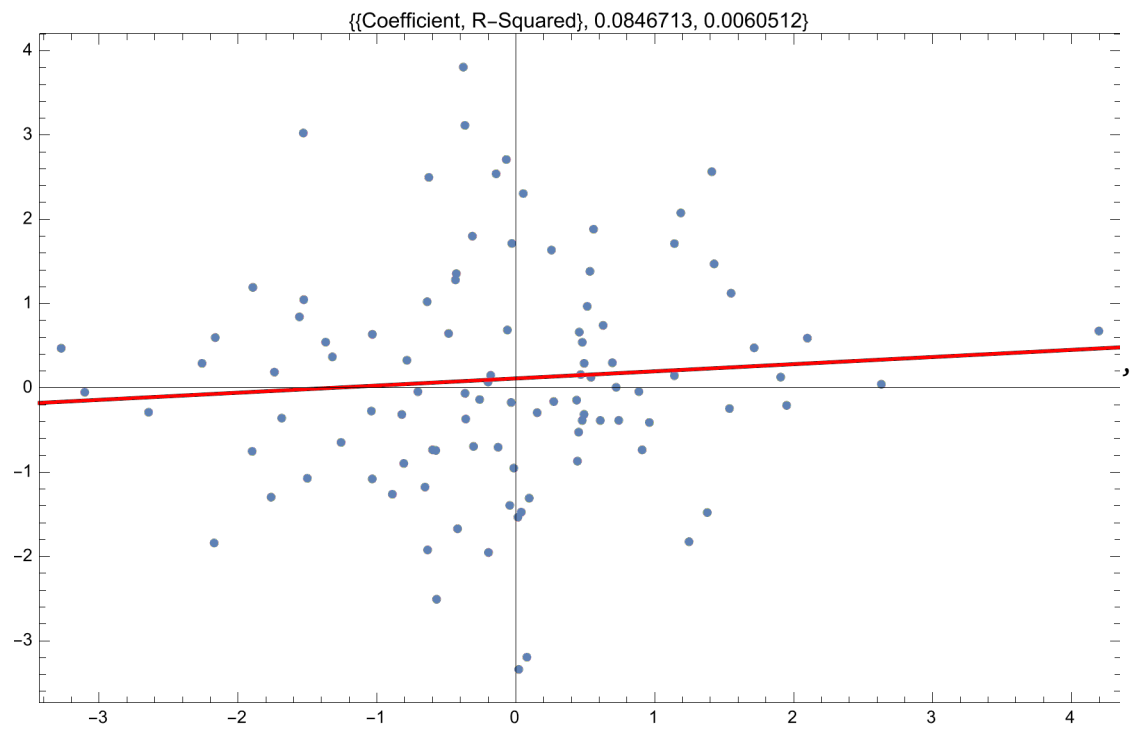


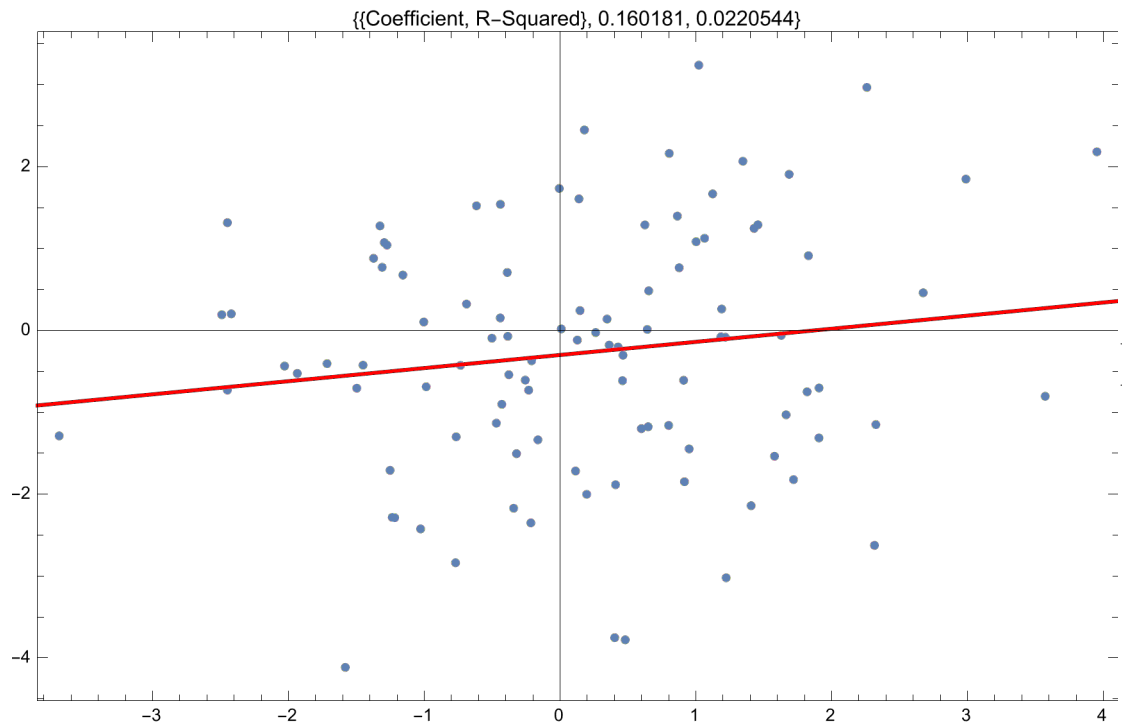












[Image]

The researcher can choose from many correlations and take the upper bound. We can find the distributions of that upper bound.

There are too many correlations in finance.

Suppose I have do correlation matrix for $0 < i \leq p$, X_i , X_i .

How many correlations we should have? It is $p(p-1)/2$. This is p^2 , remove diagonal, then take half. If you add one observation, then it will become $p(p+1)/2$.

Spuriousness decreases at $n^{1/2}$ But spuriousness increases at p^2 . This trade off is quite bad.

Clear [X]

```
f[r_] := ((1 - r^2) ^ ((n - 4) / 2)) / Beta[1 / 2, (1 / 2) (n - 2)]
```

```
dist0 = TransformedDistribution[X^2, X ≈
```

```
ProbabilityDistribution[ ((1 - r^2) ^ ((n - 4) / 2)) / Beta[1 / 2, (1 / 2) (n - 2)], {r, -1, 1}]]
```

Out[8]=

```
TransformedDistribution[X^2, X ≈ ProbabilityDistribution[ $\frac{(1 - x^2)^{\frac{1}{2}(-4+n)}}{\text{Beta}[\frac{1}{2}, \frac{1}{2}(-2+n)]}$ , {x, -1, 1}]]
```

```
PDF[TransformedDistribution[Sqrt[X], X ≈ dist0], r]
```

Out[8]=

$$\begin{cases} \frac{2 (1-r^2)^{\frac{1}{2}(-4+n)}}{\text{Beta}\left[\frac{1}{2}, \frac{1}{2}(-2+n)\right]} & 0 < r < 1 \\ 0 & r > 1 \mid r < 0 \\ \text{Indeterminate} & \text{True} \end{cases}$$

CDF[TransformedDistribution[Sqrt[X], X ≈ dist0], r]

Out[8]=

$$\begin{cases} 1 & r \geq 1 \\ \frac{2 r \text{Hypergeometric2F1}\left[\frac{1}{2}, 2-\frac{n}{2}, \frac{3}{2}, r^2\right]}{\text{Beta}\left[\frac{1}{2}, \frac{1}{2}(-2+n)\right]} & 0 < r < 1 \\ 0 & \text{True} \end{cases}$$

TransformedDistribution[Sqrt[X], X ≈ dist0] // Mean

Out[8]=

$$\frac{2}{(-2+n) \text{Beta}\left[\frac{1}{2}, \frac{1}{2}(-2+n)\right]}$$

MeanAbs = 2 / ((n - 2) Beta[1 / 2, (1 / 2) (n - 2)]);

MeanAbs /. n → 18 // N

Out[8]=

0.196381

(*TransformedDistribution[Sqrt[X], X ≈ dist0] // Variance*)

g = (2 * r * HyperGeometric2F1[1 / 2, (2 - n / 2), 3 / 2, r^2]) / (Beta[1 / 2, (1 / 2) * (n - 2)])

Out[8]=

$$\frac{2 r \text{HyperGeometric2F1}\left[\frac{1}{2}, 2 - \frac{n}{2}, \frac{3}{2}, r^2\right]}{\text{Beta}\left[\frac{1}{2}, \frac{1}{2}(-2+n)\right]}$$

fabs[r_] := (2 ((1 - r^2) ^ ((n - 4) / 2))) / (Beta[1 / 2, (1 / 2) (n - 2)])

For the special case of $\rho = 0$.

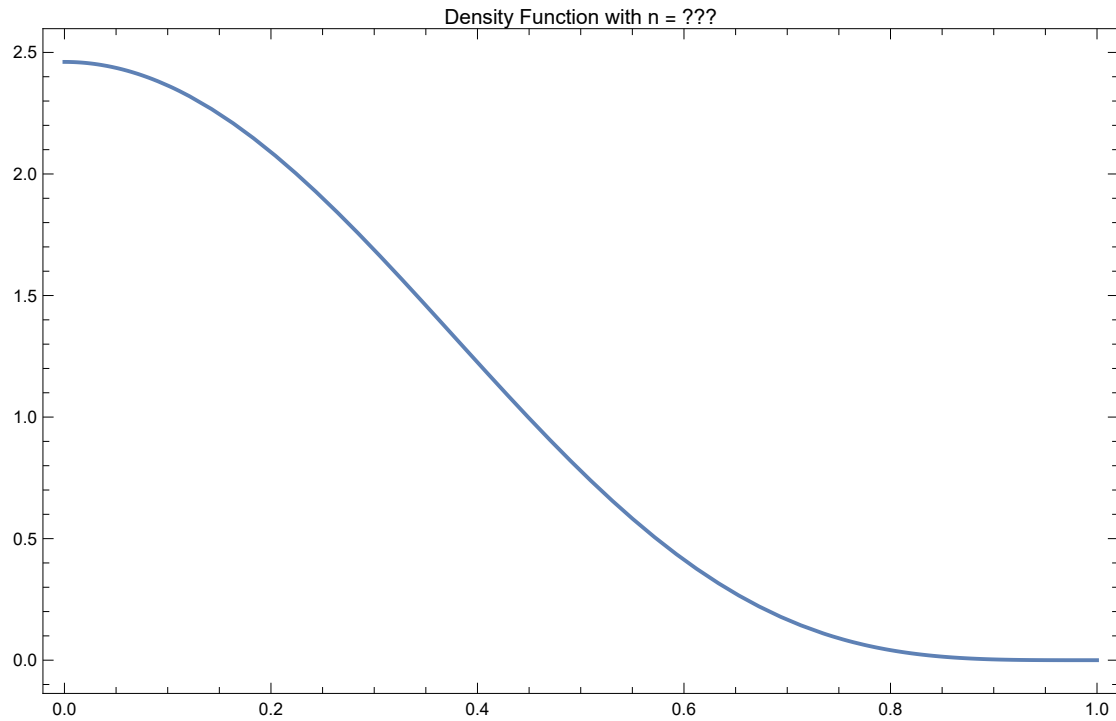
We can determine the upper bound.

Plot[fabs[r] /. n → 12, {r, 0, 1},

AxesLabel → {"Coefficient Coefficient (r)", "Density Function"},

PlotLabel → "Density Function with n = ???", Frame → True, ImageSize → Large]

Out[*]=



Example of Bad Linear Regressions

[Image]

All you need is a heavy aircraft with a low max-speed for special purposes to debunk this trend line!

[Image]

I had Gell-Mann Amnesia from narrative fallacies watching this guy.

[Image]