



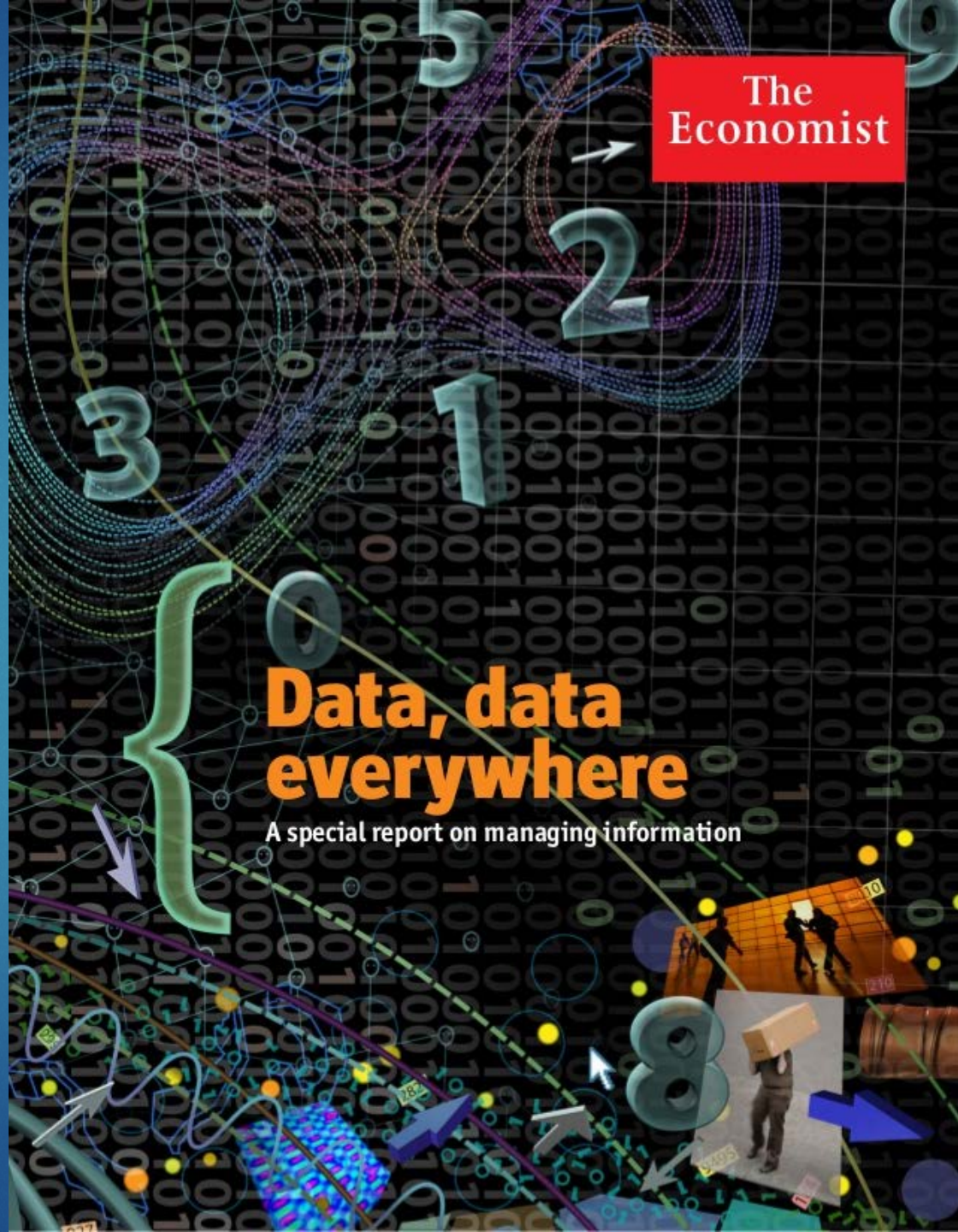
ECE3340

Numerical Fitting, Regression, Interpolation
and Approximation

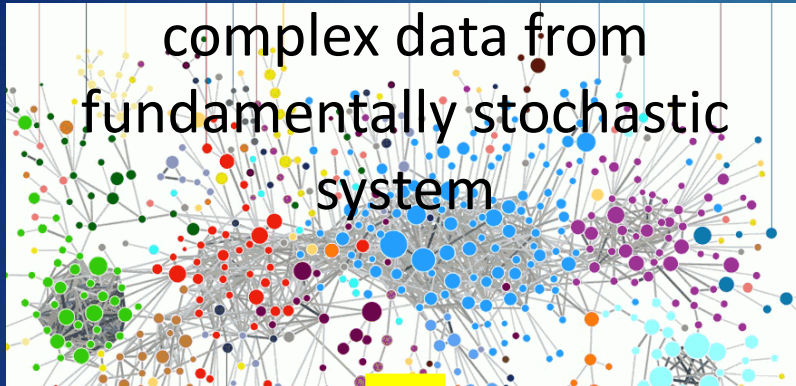
PROF. HAN Q. LE

*Note: PPT file is the main outline of the chapter topic –
associated Mathematica file(s) contain details and assignments*

Overview



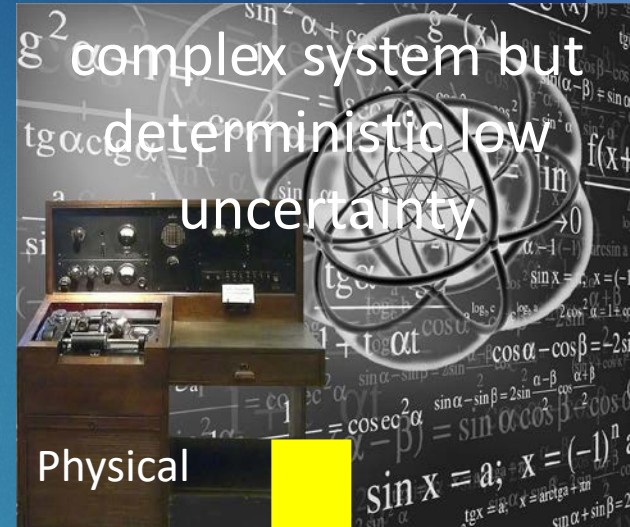
Types of problems



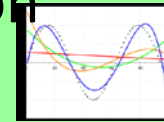
Medical, social

simplified
phenomenological
model

$$f(x_1, x_2, \dots, t)$$



simplified, fitting,
approximation

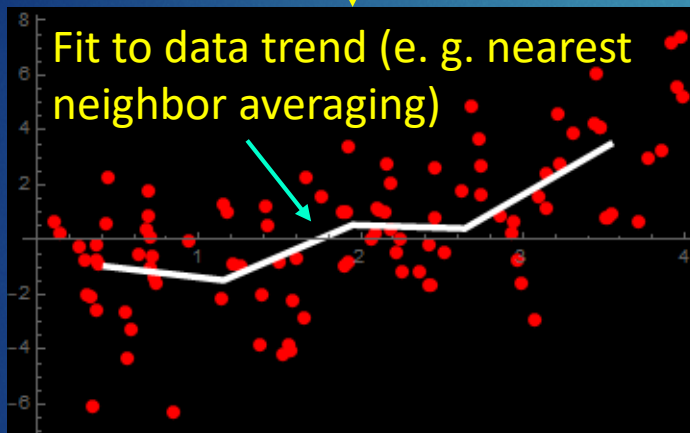


The underlying motivation is to simplify - to make things easy to understand or to implement

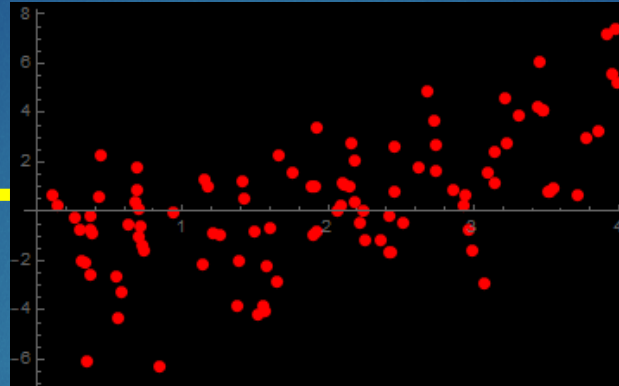
A crucial fundamental difference

Intrinsically stochastic – large fluctuation is intrinsic – not nec. errors

Data Analytics

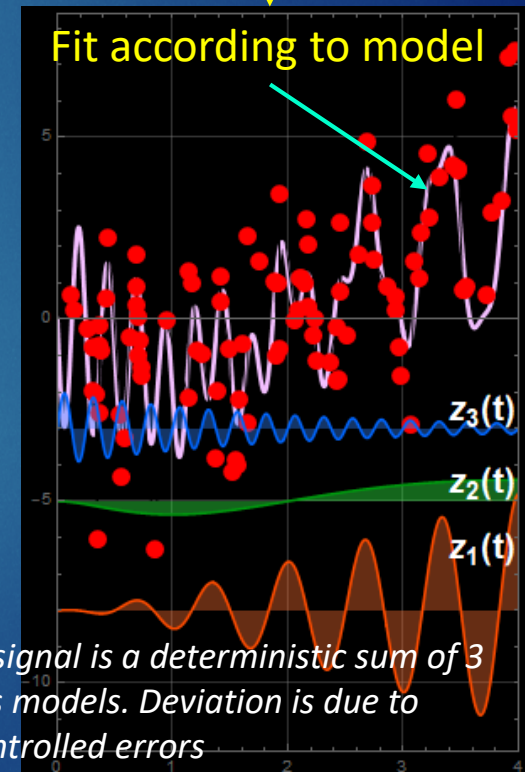


Example: a medicine efficacy: fluctuation is not due measurement errors, but genuine individual genetic/ lifestyle variation.



Deterministic basis (e. g. per physical law) with known model + random errors

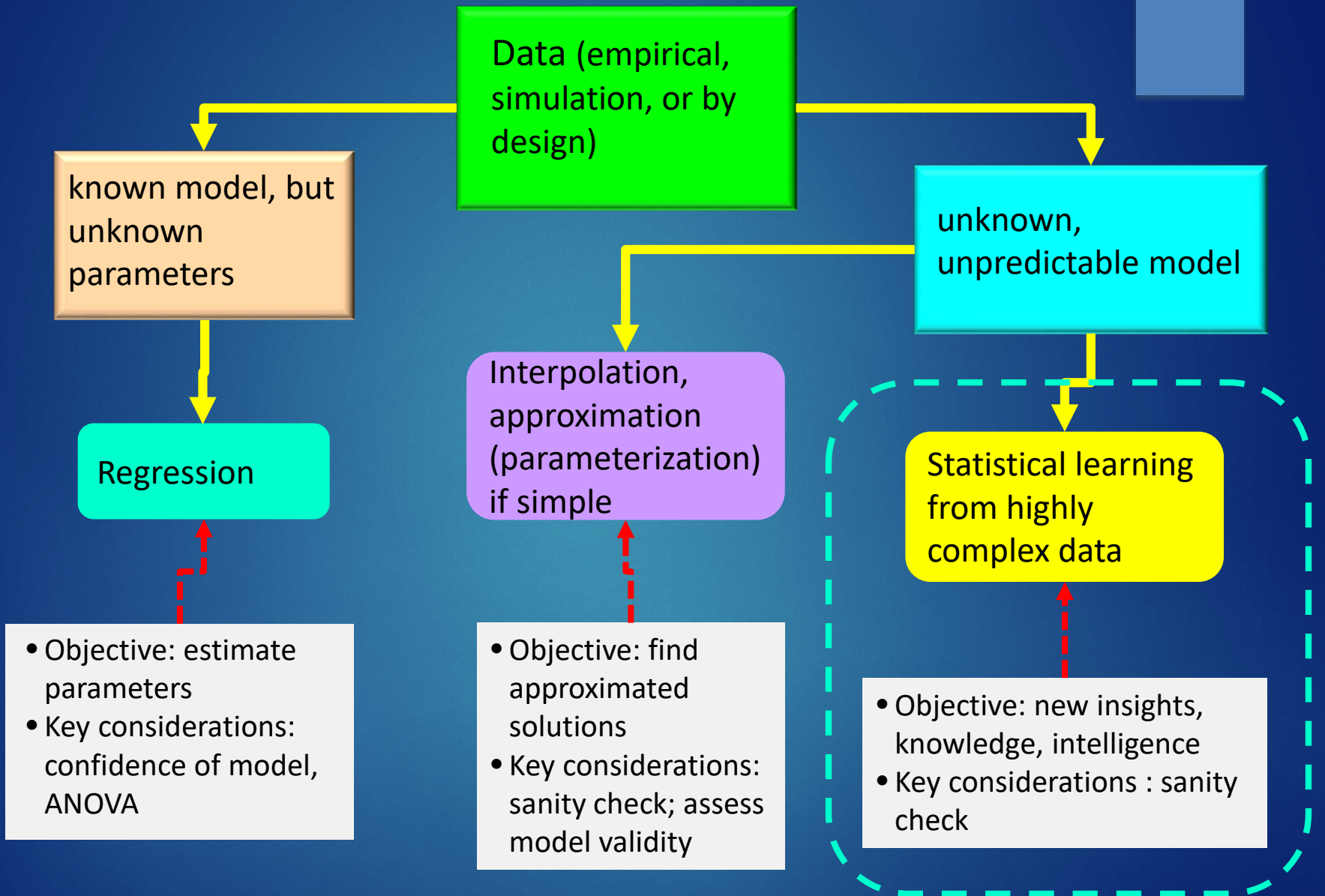
Model Analytics



Example: a signal is a deterministic sum of 3 known basis models. Deviation is due to small, uncontrolled errors

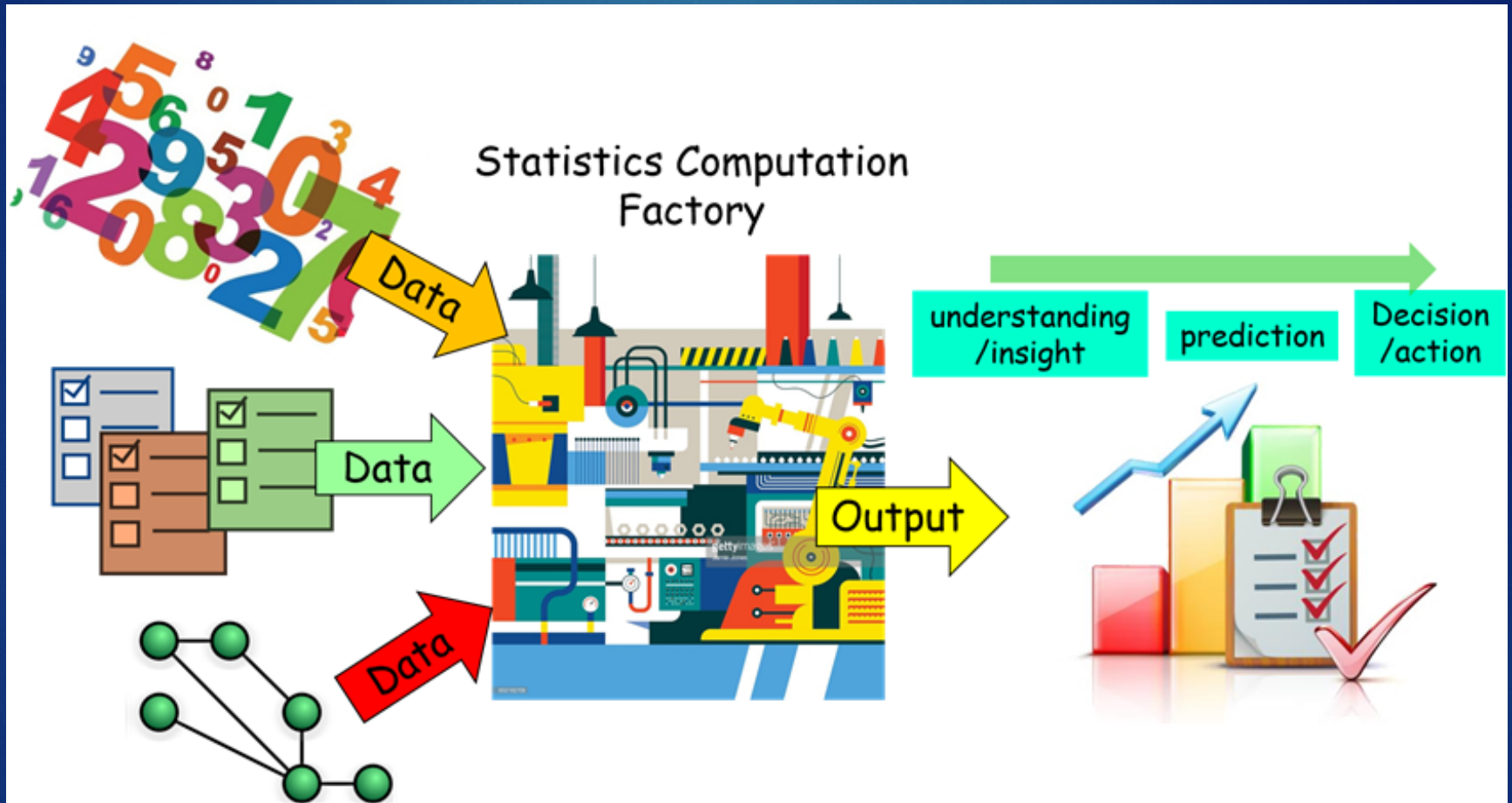
Part A – Data Analytics

DATA PRESENTATION (VISUALIZATION), STATISTICS,
REGRESSION, CLASSIFICATION.





Data Analytics – Statistical Learning



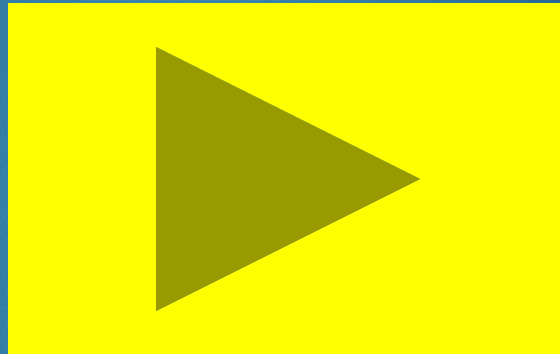
This has evolved from data analysis to data analytics, pattern classification

Outline

- ▶ Linear regression
 - ▶ single variable
 - ▶ multiple variables
- ▶ Linearization of non-linear model
 - ▶ linear-exponential or log-linear
 - ▶ power-relationship: log-log
 - ▶ general non-linear
- ▶ General model fit
 - ▶ Least squares of linear combination of basis functions
- ▶ Discrete (quantized) variables and generalized regression: logistic regression
- ▶ Introduction to data clusters and classification

Linear Regression

Introduction to regression concept
(see Mathematica lecture file)

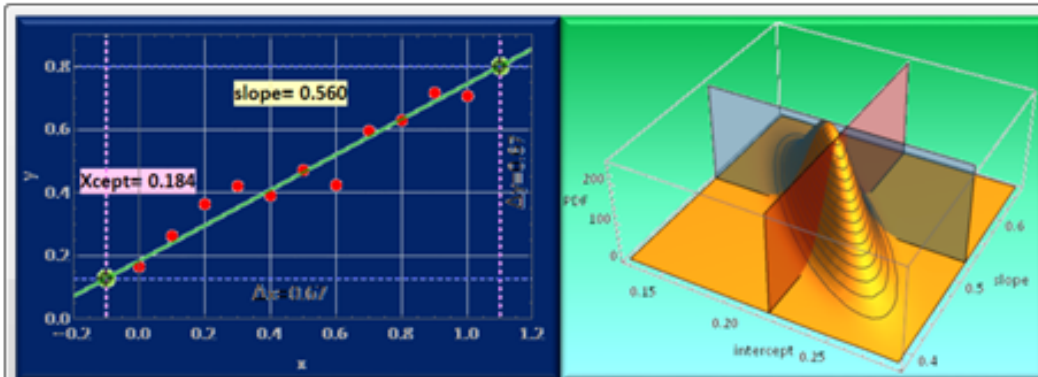


Linear regression with single variable

UNIVERSITY of HOUSTON App by Han Q. Le ©

ECE3340-APP-Linear regression

RUN STATUS/CONTROL →

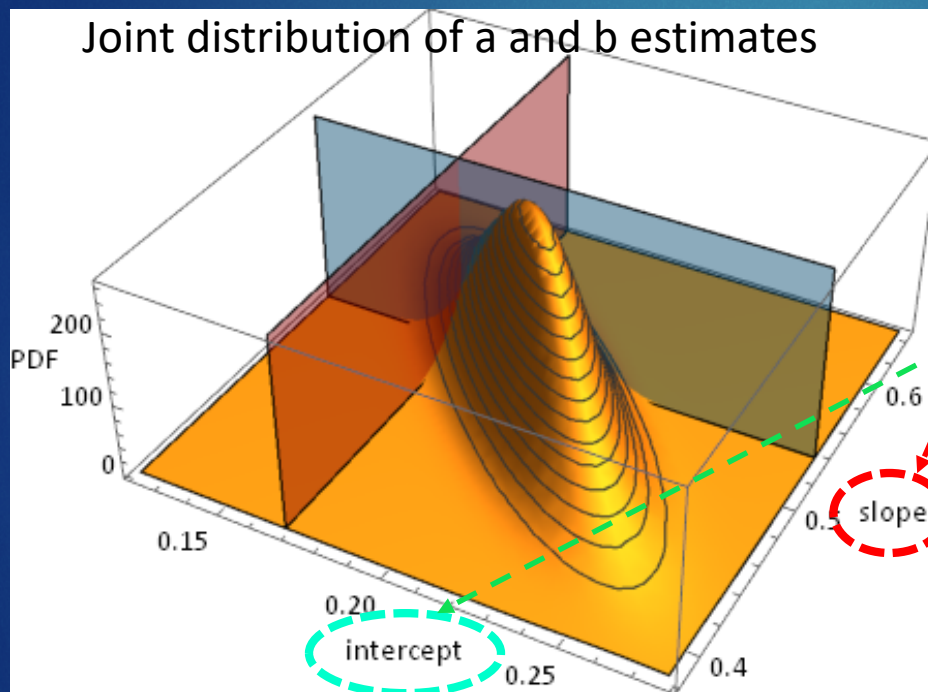


	Deg. freed.	Sum sq	Mean sq	F-Statistics	P-Value
x	1	0.296605	0.296605	133.17	1.07351×10^{-6}
Error	9	0.0200454	0.00222727		
Total	10	0.31665			

Key concepts

- ▶ Model parameters:
 - ▶ coefficients, correlation R^2
 - ▶ standard error, covariance matrix, correlation matrix
 - ▶ confidence ellipsoid
- ▶ Linear regression statistics
 - ▶ residuals
 - ▶ parameter t-statistics, P-value
 - ▶ Analysis of variance (ANOVA): dof, sum of squares, mean squares, F-statistics

Covariance matrix of parameters – Confidence ellipsoid



$$\hat{a} = \frac{\frac{\sum_{i=1}^n x_i y_i}{n} - \frac{\sum_{i=1}^n x_i}{n} \frac{\sum_{i=1}^n y_i}{n}}{\frac{\sum_{i=1}^n x_i^2}{n} - \left(\frac{\sum_{i=1}^n x_i}{n}\right)^2} = \frac{\bar{x}\bar{y} - \bar{x}\bar{y}}{\bar{x}^2 - \bar{x}^2}$$
$$\hat{b} = \bar{y} - \hat{a}\bar{x}$$

Estimates for a and b are not independent. They are related by mean x and mean y as shown, hence, the distribution of their values are not independent.

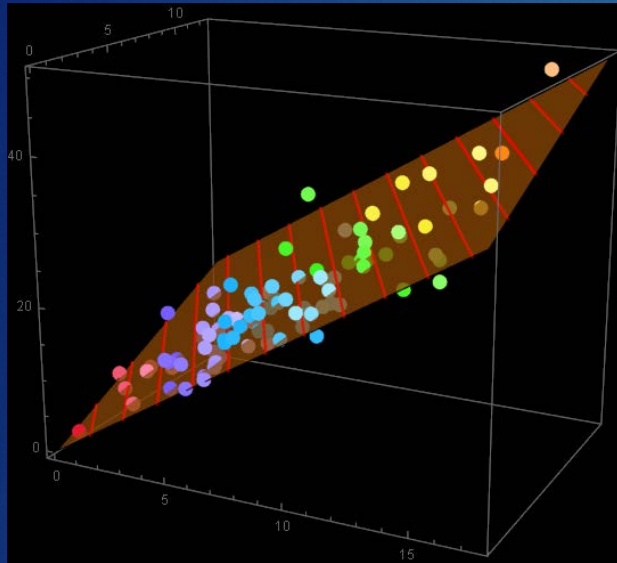
in class demo: if we know one coefficient by any other mean, this changes the estimate for the other coefficient (move the planes).

Key concepts

- ▶ Model parameters:
 - ▶ coefficients, correlation R^2
 - ▶ standard error, covariance matrix, correlation matrix
 - ▶ confidence ellipsoid
- ▶ Linear regression statistics
 - ▶ residuals
 - ▶ parameter t-statistics, P-value
 - ▶ Analysis of variance (ANOVA): dof, sum of squares, mean squares, F-statistics

From the example discussed

Null hypothesis

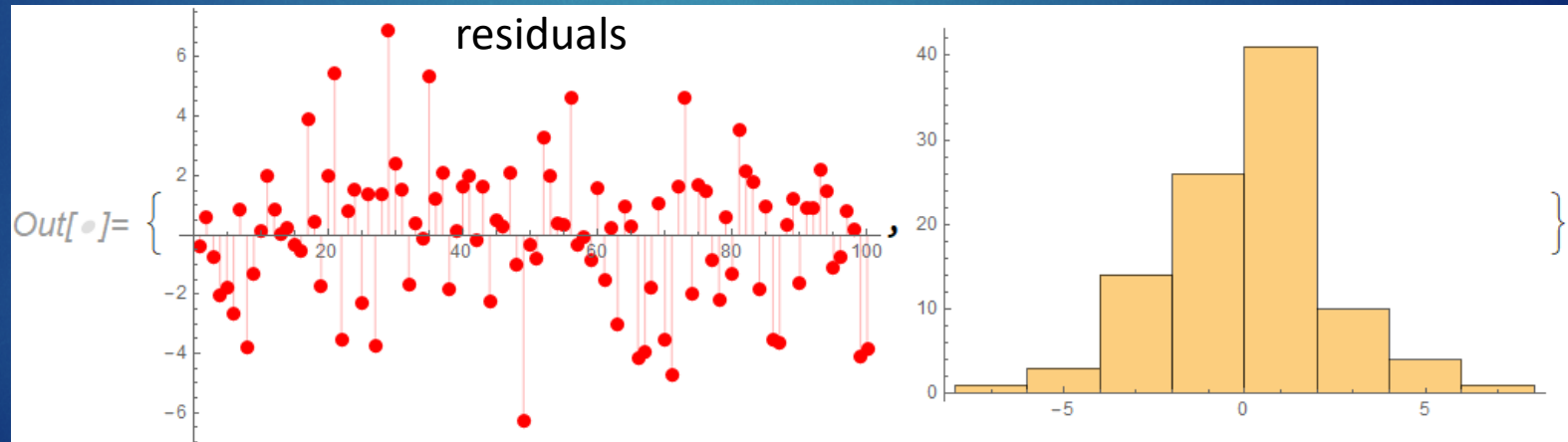


Out[•]=

	Estimate	Standard Error	t-Statistic	P-Value
u1	2.03334	0.0463809	43.8399	3.34417×10^{-66}
u2	1.1749	0.0853734	13.7619	1.25441×10^{-24}

Out[•]=

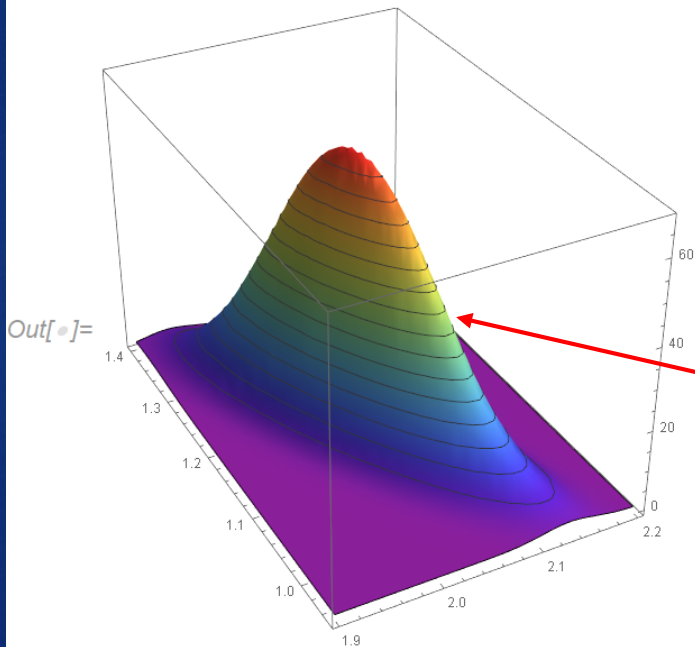
	DF	SS	MS	F-Statistic	P-Value
u1	1	48363.1	48363.1	8665.47	1.93448×10^{-97}
u2	1	1057.02	1057.02	189.391	1.25441×10^{-24}
Error	98	546.951	5.58113		
Total	100	49967.1			




```
In[•]:= covMat // MatrixForm
```

```
Out[•]//MatrixForm=  $\begin{pmatrix} 0.00215119 & -0.00319625 \\ -0.00319625 & 0.00728862 \end{pmatrix}$ 
```

```
In[•]:= Plot3D[PDF[MultinormalDistribution[ahat, covMat], {u, v}],  
{u, 1.9, 2.2}, {v, 0.95, 1.4}, PlotRange → All,  
, BoxRatios → {3, 4.5, 3}, ColorFunction → "Rainbow",  
, MeshFunctions → {#3 &}]
```



```
In[•]:= eigen
```

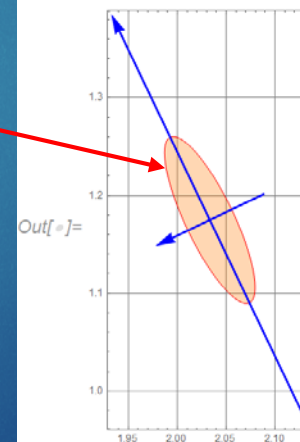
	Eigenvalue	Index	u1	u2
Out[•]=	1.10248	1.	0.448759	0.448759
	0.897518	1.10832	0.551241	0.551241

```
In[•]:= ahatconf
```

	Estimate	Standard Error	Confidence Interval	
Out[•]=	u1	2.03334	0.0463809	{1.9413, 2.12538}
	u2	1.1749	0.0853734	{1.00548, 1.34432}

```
Out[•]= FittedModels`ParameterEllipsoid[  
{2.03334, 1.1749}, {0.233444, 0.0618608},  
{{-0.432183, 0.901786}, {-0.901786, -0.432183}}]
```

```
In[•]:= Graphics[{{EdgeForm[{Thickness[0.005], Red}],  
Hue[0.08, 1, 1, 0.3], Ellipsoid[ahat, covMat]  
, Blue, Opacity[1], Thickness[0.01], Arrowheads[0.1]  
, Arrow[#] & /@  
{({ahat, ahat} + ahatellipse[[2, 1]] *  
{-ahatellipse[[3, 1]], ahatellipse[[3, 1]])}  
, ({ahat, ahat} + ahatellipse[[2, 2]] *  
{-ahatellipse[[3, 2]], ahatellipse[[3, 2]])}}},  
Frame → True, GridLines → Automatic]
```

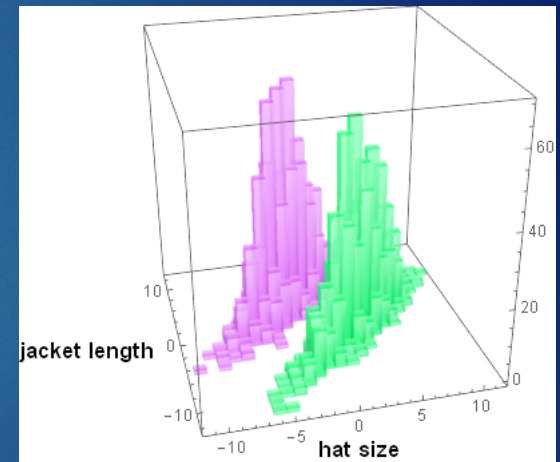
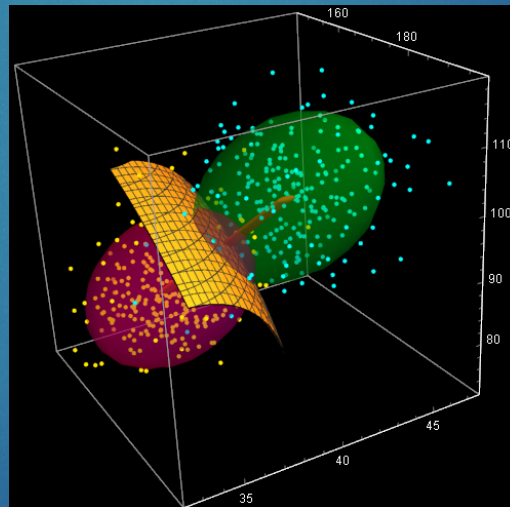
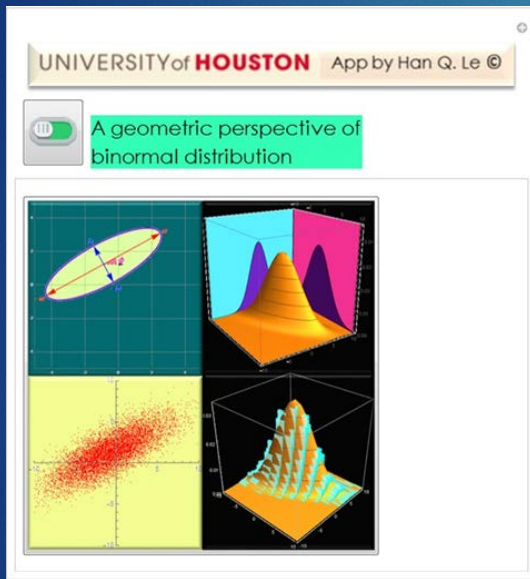


A very useful exercise to understand variable correlation

Review of binormal distribution

Example: human body measures- gender correlation

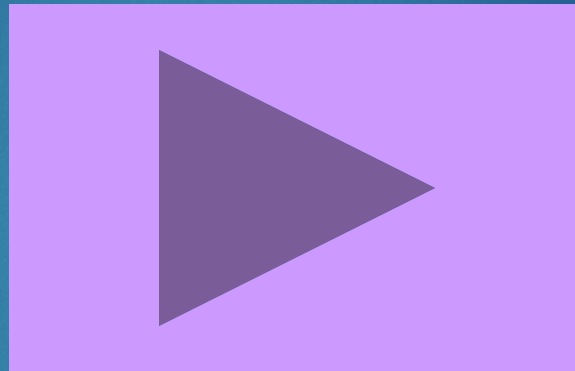
Example in HW



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Mathematica file on generalized least-square regression



Introduction to data survey and visualization methods
(see Mathematica lecture file)

