

# *Mathematical Foundations for ML*

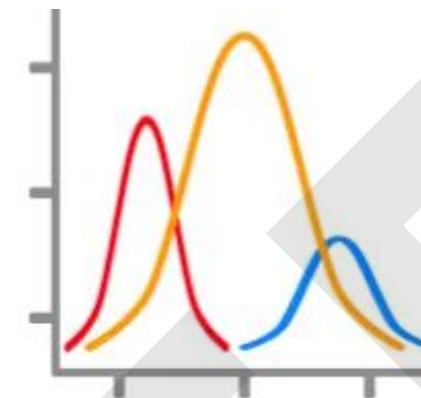
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*Dr. Arulalan Rajan*

# Machine Learning!

Linear Algebra

$$d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

[matrix]

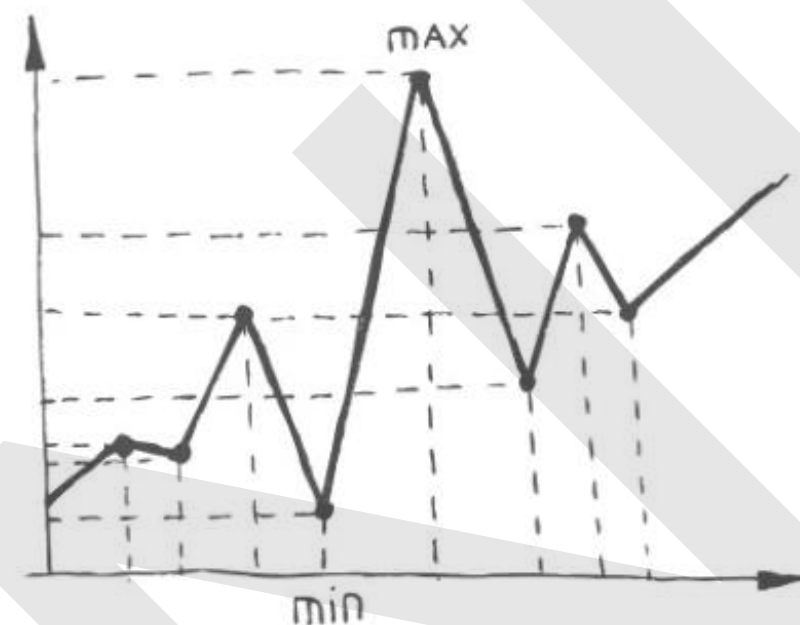


$P(A | B)$

Statistics



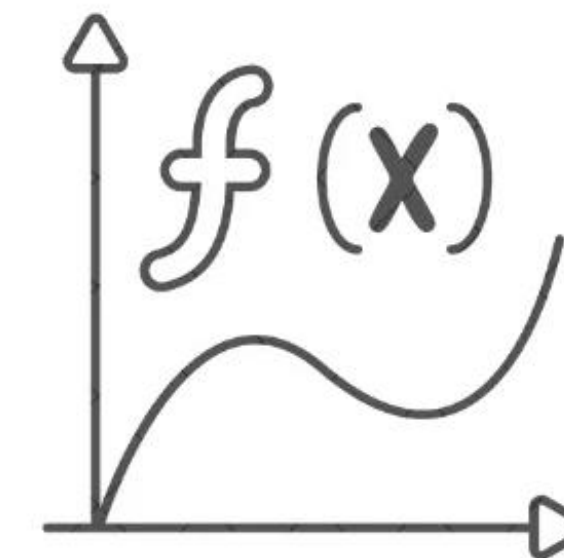
Graphs



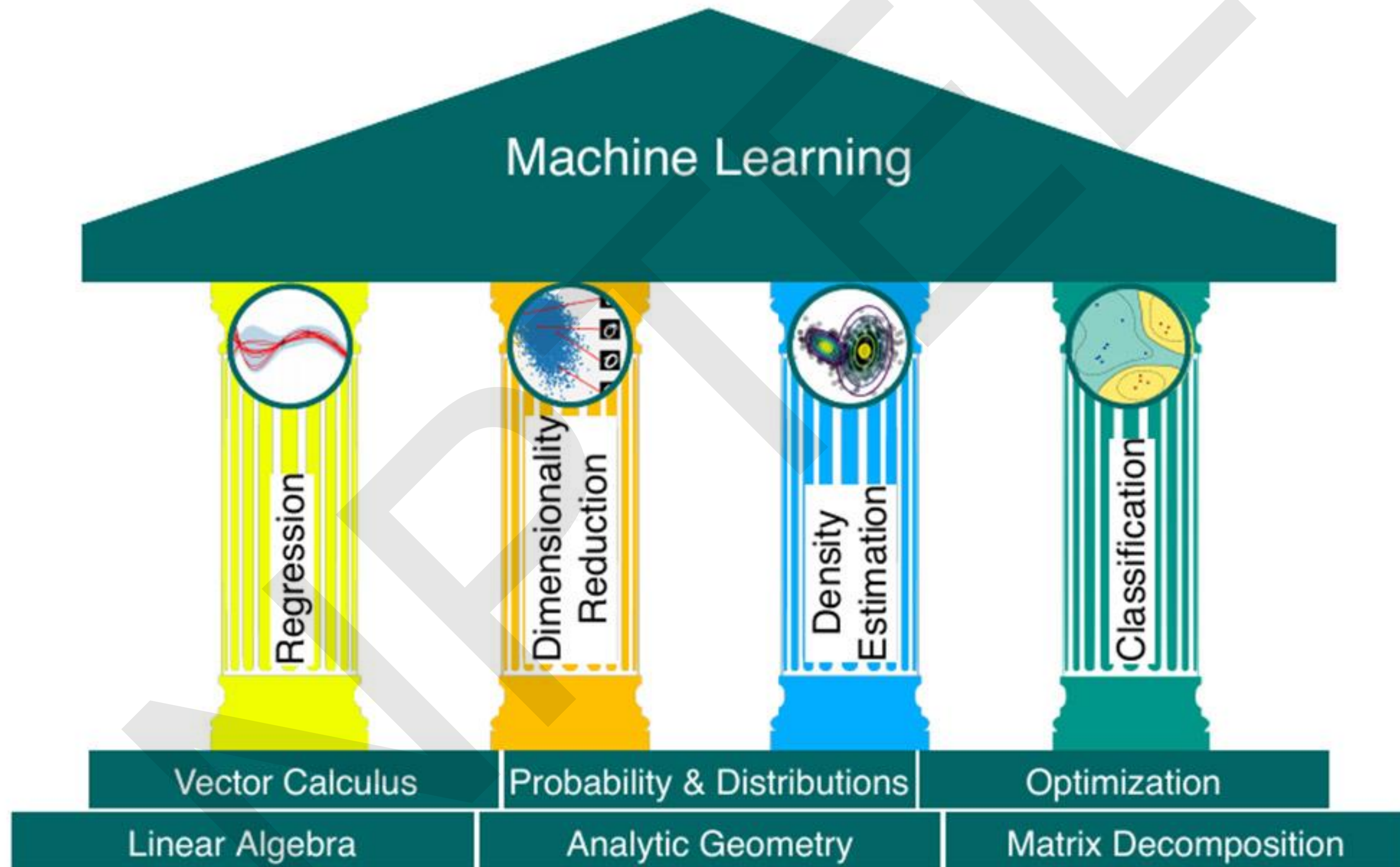
**Machine Learning**

enjoyalgorithms.com

Calculus

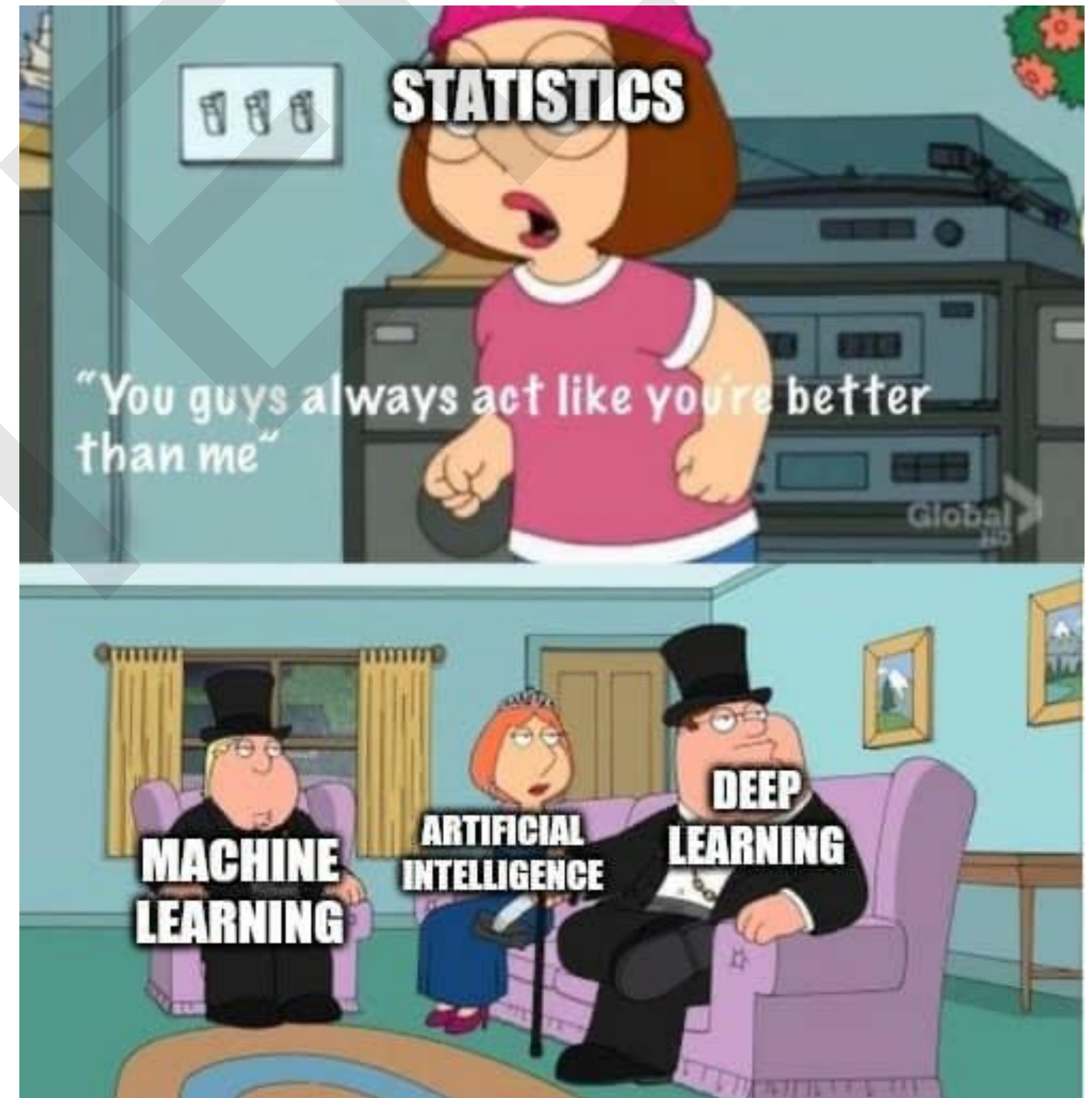
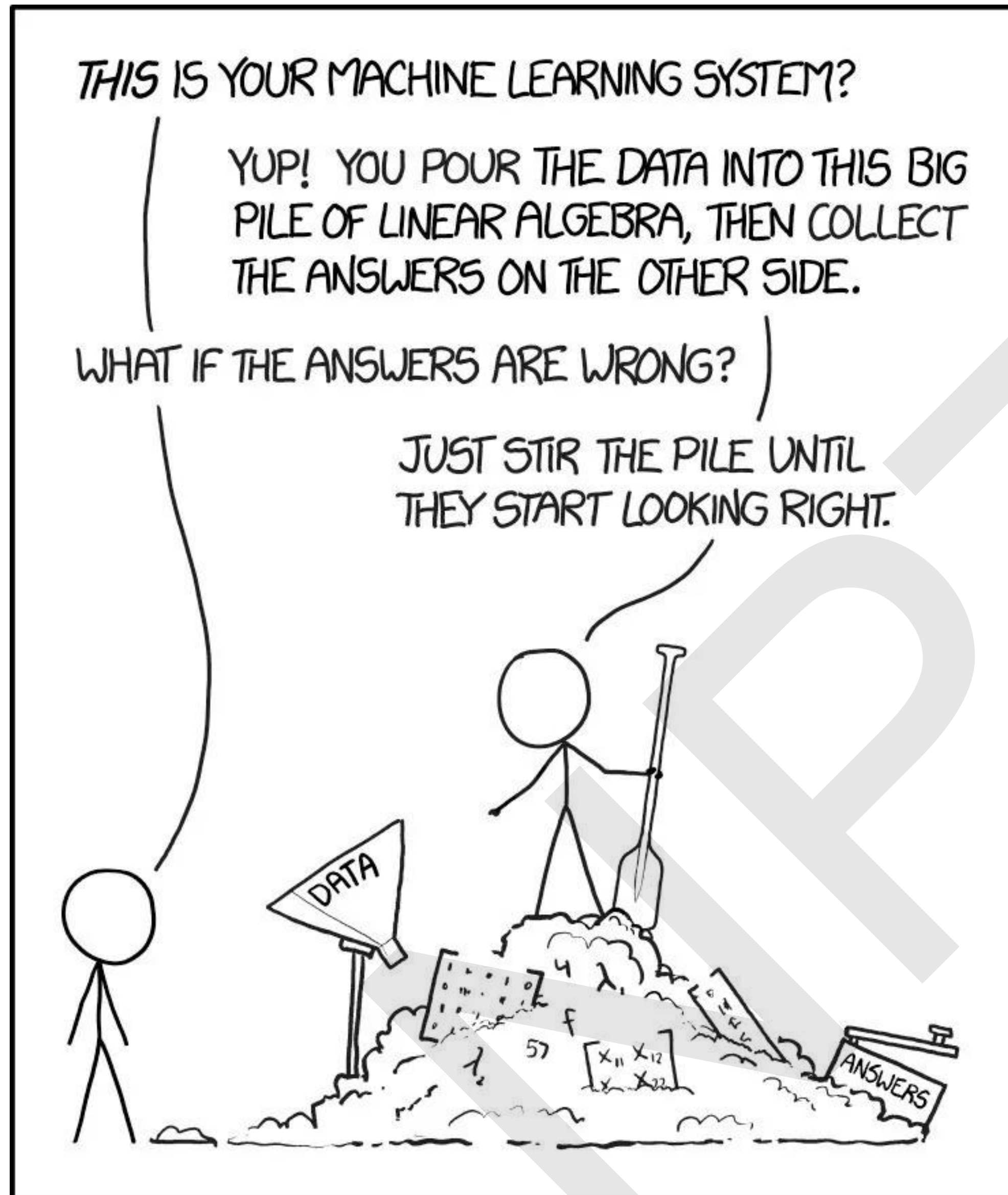


# *Machine Learning!*





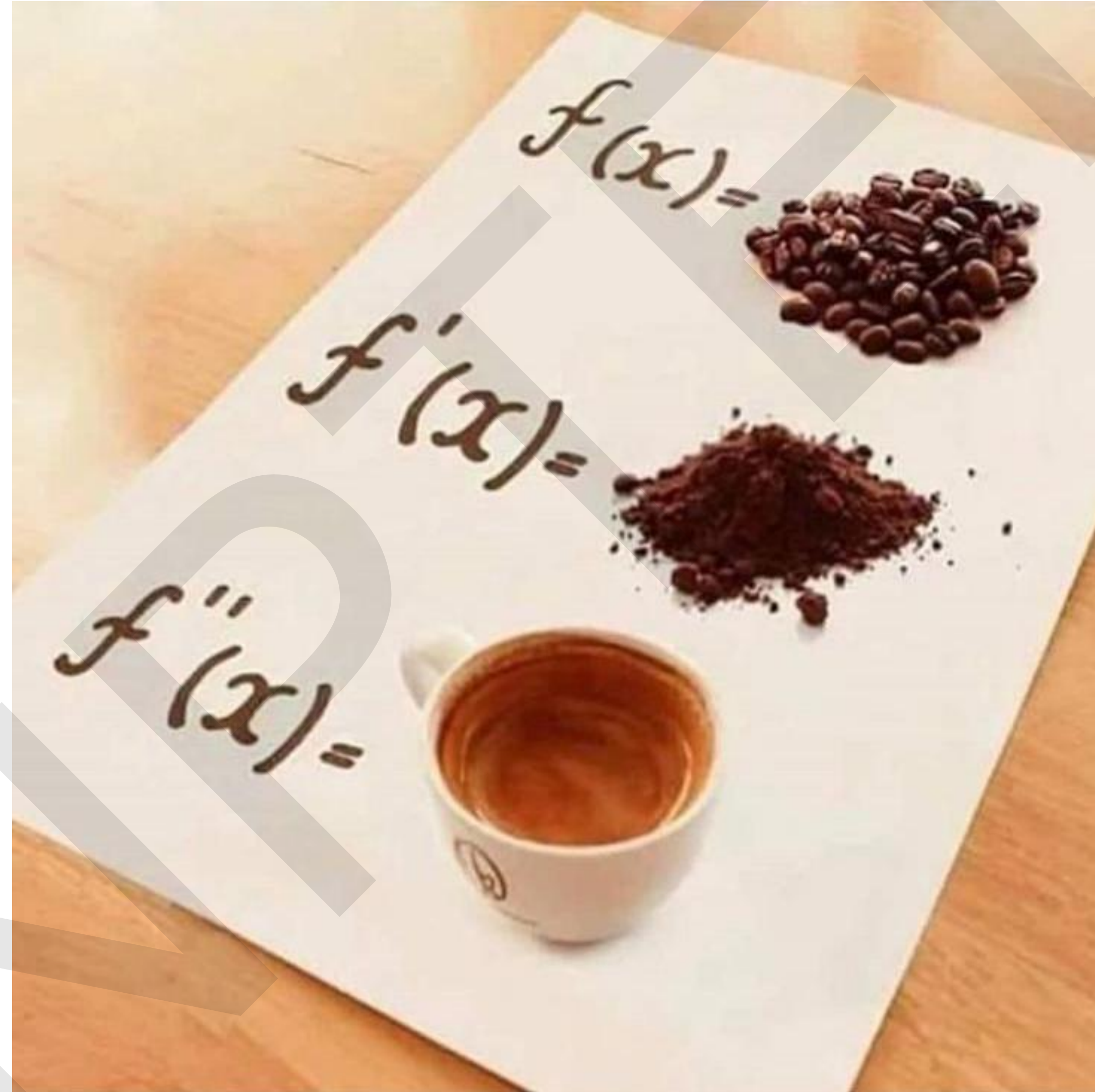
# From the www



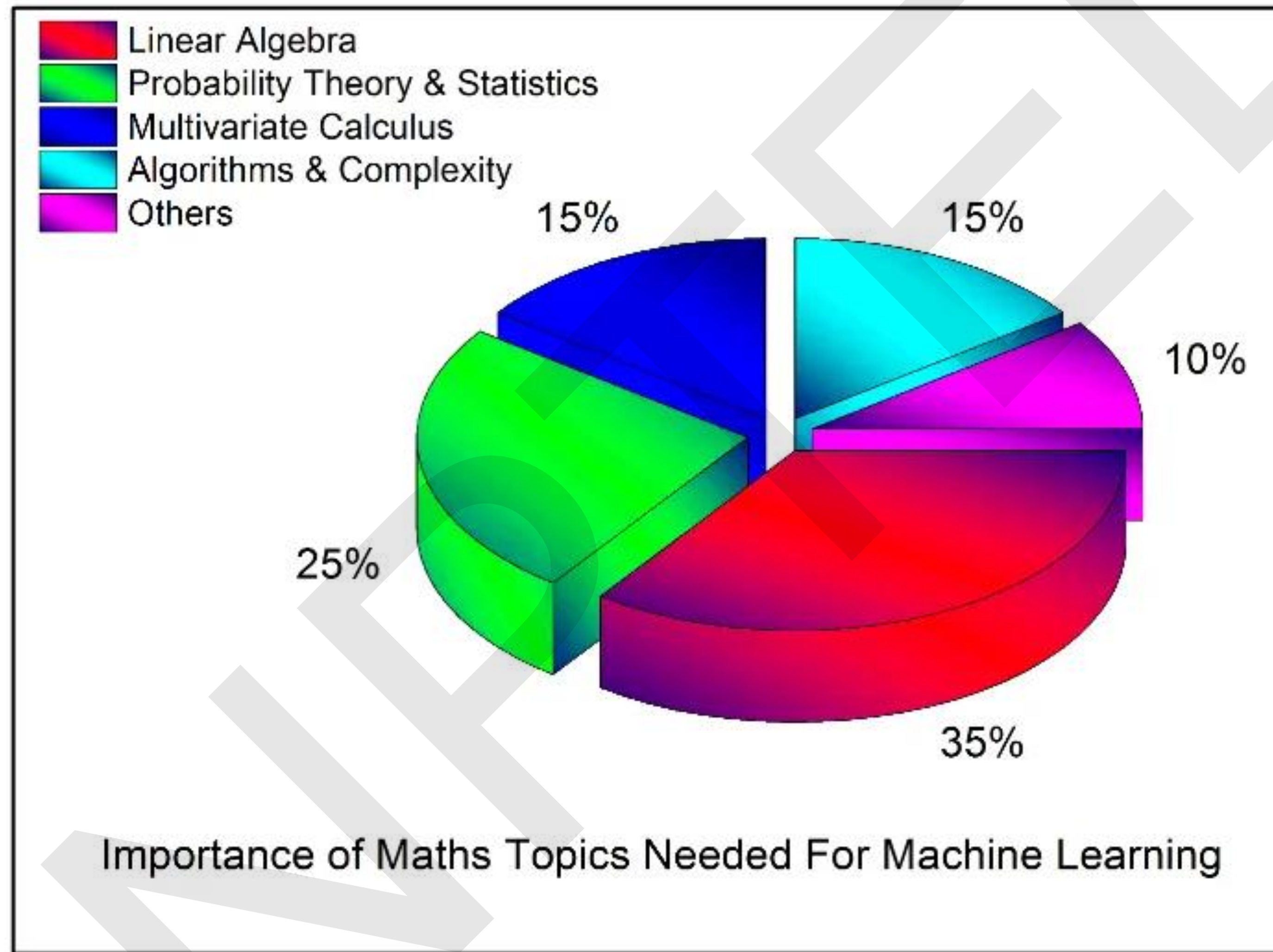
<https://thunderdungeon.com/2024/05/17/ai-memes-3/>



# *How I wish Calculus was!*



# *Machine Learning!*



# *Math in Machine Learning*

- Data for each variable - vector
- Table of such vectors - Matrix
- Playing around with matrices - Linear Algebra and Matrix Theory
- Dot product of two vectors - Similarity
- Combination of vectors - Feature in ML
- Idea of Projection - Dimensionality reduction

$$\begin{aligned}a_{11} x_1 + a_{12} x_2 &= b_1 \\ a_{21} x_1 + a_{22} x_2 &= b_2 \\ x_1 \begin{bmatrix} a_{11} \\ a_{21} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \end{bmatrix} &= \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}\end{aligned}$$

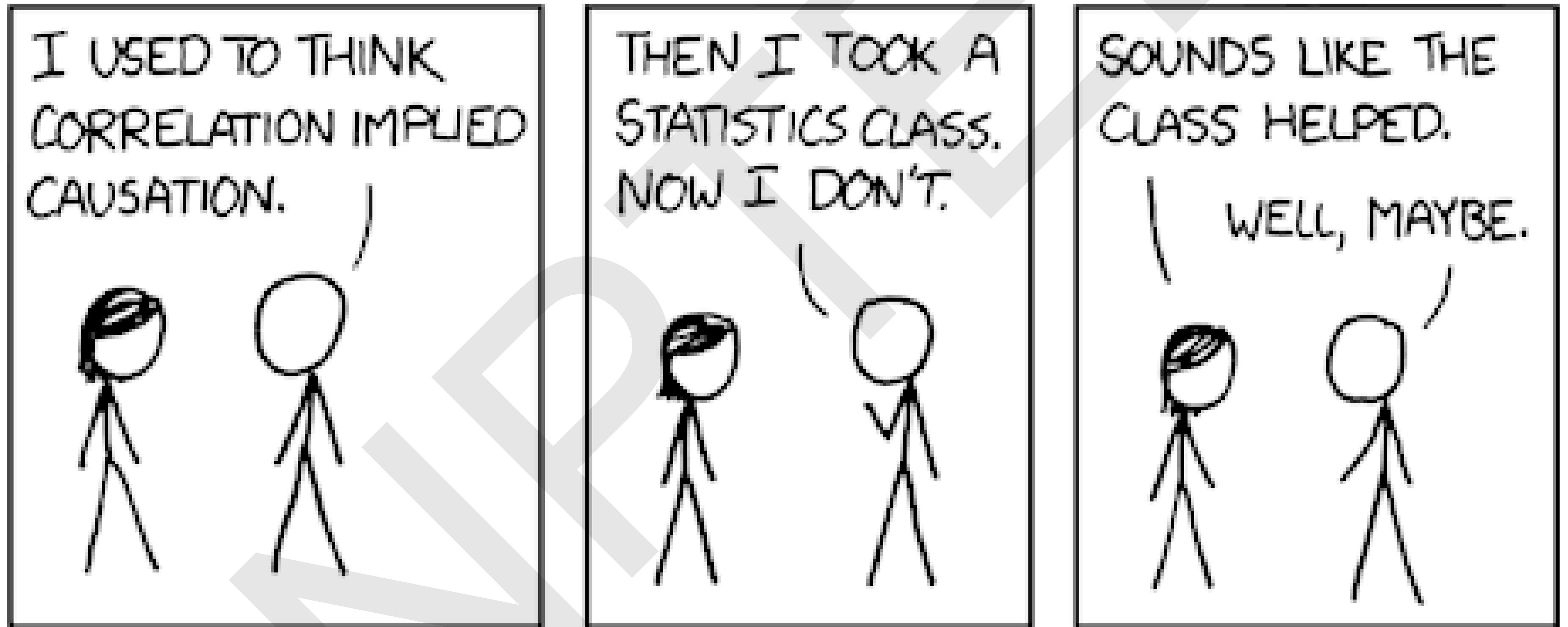


# *Math in Machine Learning*

- ML - Prediction - Classification and Regression
- Prediction to make informed decisions and effective risk management
- Confidence quantification - Requires Probability
- Ideas of Conditional Probability, Bayes Theorem, random variables, Conditioning random variables, Correlation, Covariance etc...



# *Math in Machine Learning*



# *Math in Machine Learning*

- ML - Training on historical data
- Training ML models - Finding out and adjusting the model parameters to optimise (maximise or minimise) some performance measure
- Maximize or minimise - Multivariable calculus and optimisation
- Gradient descent algorithm to find parameters that optimise - multivariable calculus
- Optimisation - Constrained or unconstrained!



# *Course Flow*

# *Week 1 to Week 4 - Linear Algebra*

- Week 1: Vectors, Vector Spaces and Subspaces
- Week 2: Linear Transformations, eigenvalues and eigenvectors
- Week 3: Orthogonality, Projection and Real symmetric matrices
- Week 4: Singular value decomposition, Principal Component Analysis, Support Vector Machines and Applications



## *Week 5 to Week 8 - Probability and Random Variables*

- Week 5: Probability Foundations - From Events to Bayes' Theorem
- Week 6: Random Variables, Moments of Random Variables
- Week 7: Jointly Distributed Random Variables, Conditioning of Random variables
- Week 8: Limit Theorems, Sample Geometry, Covariance Matrices and Properties

## *Week 9 to Week 12 -Multivariate Calculus, Optimization and Applications*

- Week 9: Taylor's series, Partial Derivatives, Chain rule, Gradient, Jacobian, Hessian
- Week 10: Matrix Derivatives, Gradient Descent and Stochastic Gradient Descent, Constrained and Unconstrained optimization, Lagrangian, Least Squares and PCA
- Week 11: Neural Nets, Perceptron, Back Propagation Algorithm
- Week 12: Algorithms for ML - Classification, Clustering and Regression

$$b^{m \times 1} = A^{m \times n} x^{n \times 1}$$

↓  
m functions → n variables



# *From the www*

**When you finally study Linear Algebra,  
Probability Theory and Statistics after  
completing a Machine Learning course**



*Namaste!*