

PSY9511: Seminar 5

Unsupervised learning

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1. Overview of unsupervised learning
2. Clustering
 - K-means
 - Hierarchical
3. Dimensionality reduction
 - Principal component analysis (PCA)
 - Independent component analysis (ICA)
 - Partial least squares (PLS)

Unsupervised learning



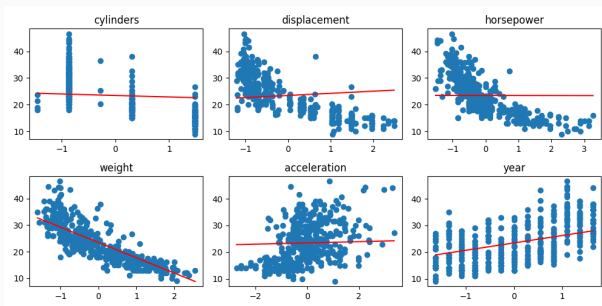
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Supervised learning: Find $\hat{y} = f(X)$

Unsupervised learning: Motivation

Supervised learning: Find $\hat{y} = f(X)$

- Descriptive: Understand the relationship between X and y



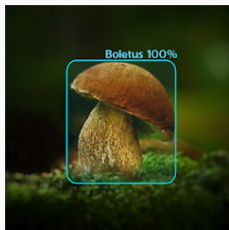
Supervised learning: Find $\hat{y} = f(X)$

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- Predictive: Predict y given new X .

Unsupervised learning: Motivation

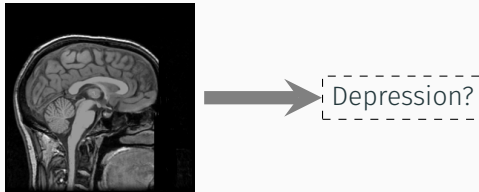
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- Can we find subgroups or interesting axes of variability?



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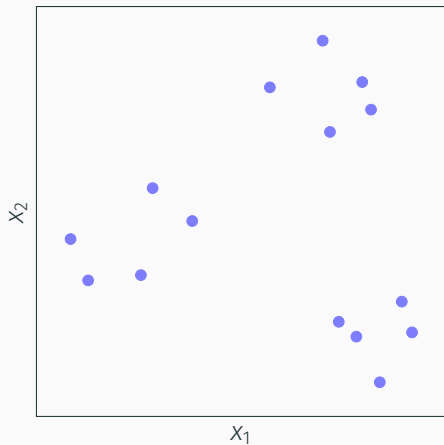
- Can we find subgroups or interesting axes of variability?
- Visualization

Dimensionality reduction

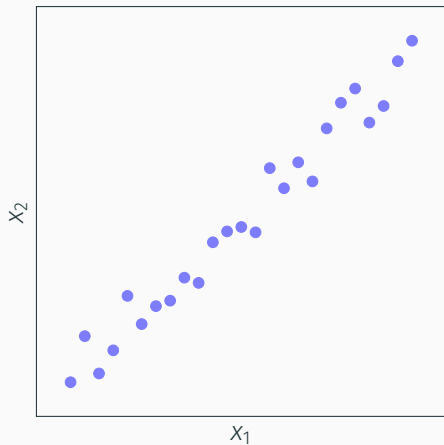


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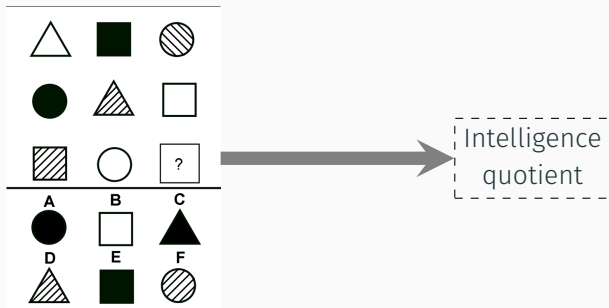
Dimensionality reduction: Motivation



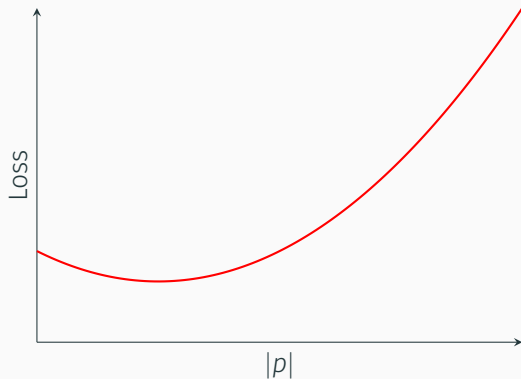
Dimensionality reduction: Motivation



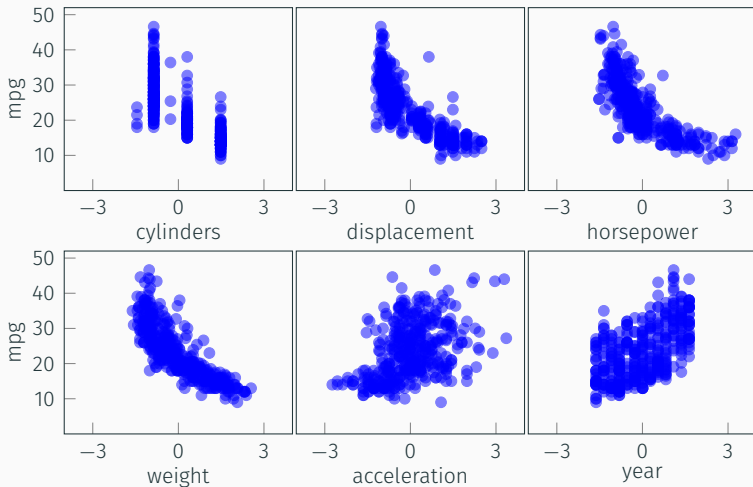
Dimensionality reduction: Motivation



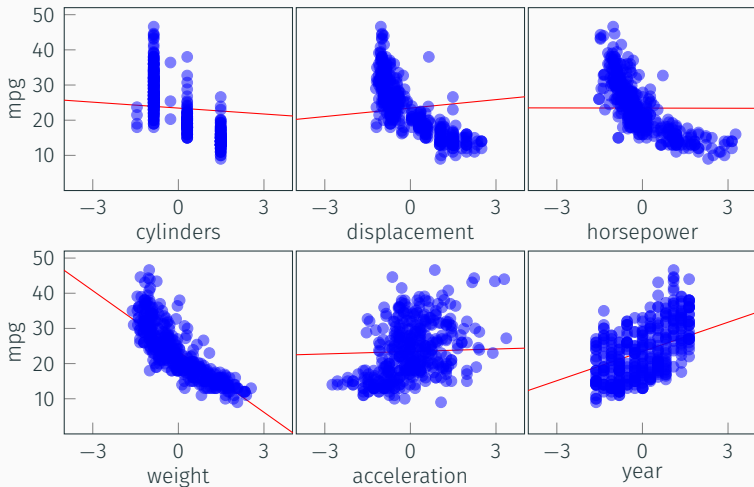
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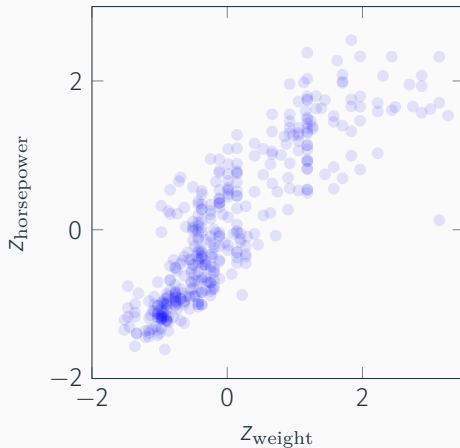
Dimensionality reduction: Motivation

1	0.30	0.86	0.89	0.41	0.93
0.30	1	0.41	0.34	0.29	0.36
0.86	0.41	1	0.84	0.68	0.89
0.89	0.34	0.84	1	0.50	0.95
0.41	0.29	0.68	0.50	1	0.54
0.93	0.36	0.89	0.95	0.54	1

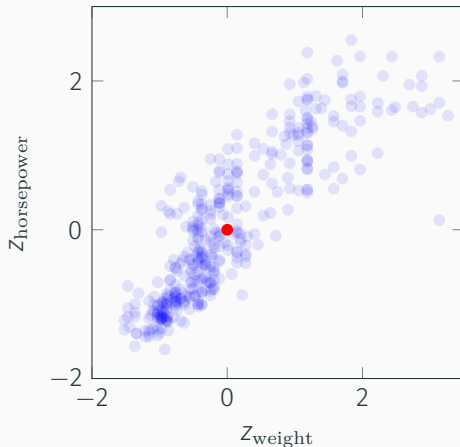
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Dimensionality reduction: Principal component analysis

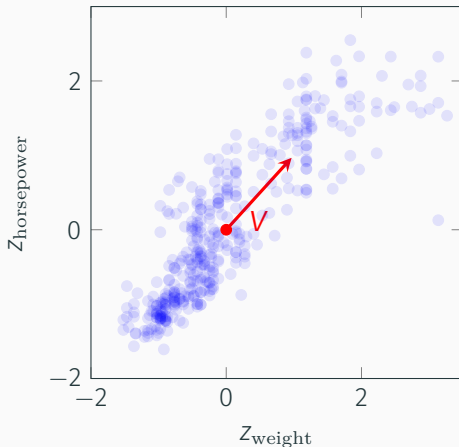


Dimensionality reduction: Principal component analysis



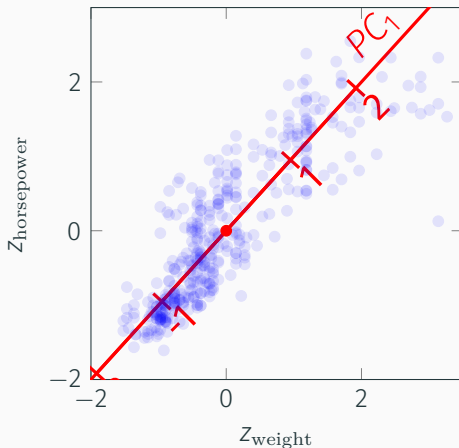
$c \rightarrow$ center of the data

Dimensionality reduction: Principal component analysis



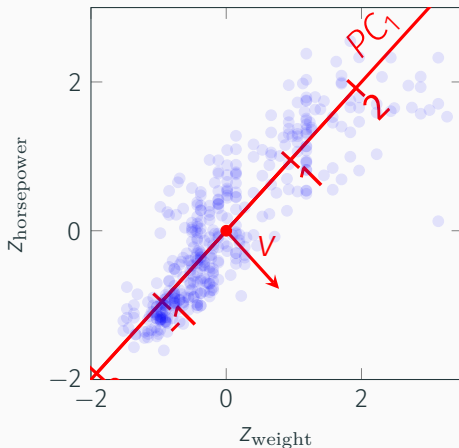
$v \rightarrow$ direction of maximum variance

Dimensionality reduction: Principal component analysis



$$PC_1 \rightarrow 0.69 * Z_{\text{horsepower}} + 0.71 * Z_{\text{weight}}$$

Dimensionality reduction: Principal component analysis



$v \rightarrow$ direction of maximum variance **orthogonal** to PC_1