

[illegible]

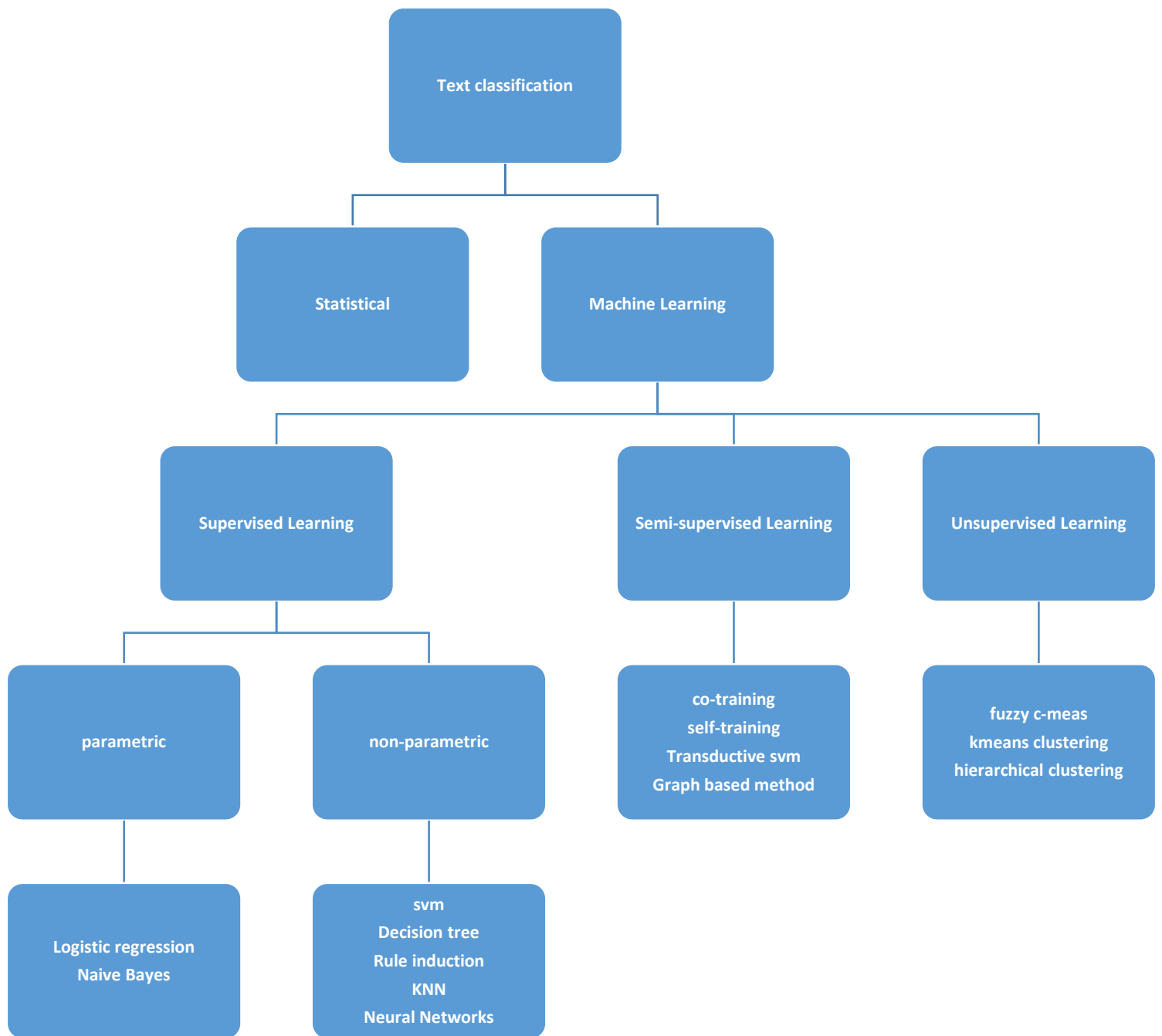


Figure 1 Representation of Text Classification methods

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