

Supplemental material to: Analysis of the Term ‘Big Data’: Usage in Biomedical Publications

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I. APPENDIX A

The six identified biomedical papers that do not use the term “Big Data” in their title, abstract, or keywords are: [1, 2, 3, 4, 5, 6].

II. APPENDIX B

A. Visual inspection of classification results

The following documents were selected for inspection, presented here in decreasing order of classification strength:

- TP: [7, 8, 9, 10, 11, 12, 13, 14, 15, 16];
- FN: [17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33];
- FP: [34, 35, 36, 37, 38, 39, 40, 41, 42, 43].

Analysis of TP.

Upon reading of the original title and abstract of the TP documents, the following Big Data topics were identified by consensus:

- [7]: Volume, Velocity, Value, Methods, Technology, Impact.
- [8]: Volume, Variety, Veracity, Value, Methods, Technology, Impact;
- [9]: Volume, Technology, Beyond conventional;
- [10]: Volume, Variety, Value, Method, Impact;
- [11]: Volume, Variety, Methods, Technology;
- [12]: Volume, Value;
- [13]: Volume, Variety, Value, Impact;
- [14]: Volume, Velocity, Variety, Value, Methods, Impact;
- [15]: Veracity, Value; and
- [16]: Veracity, Value, Impact.

Analysis of FP.

- [34]: Methods, Variety;
- [35]: Methods, Technology;
- [38]: Variety, Value, Methods;
- [39]: Volume, Value, Methods;
- [40]: Volume, Value, Impact;
- [41]: Variety, Value, Methods, Impact; and
- [42]: Volume, Technology.
- [43]: Value;

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