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**PAJAIS Submission Style Guide**

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***Abstract***

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| *All PAJAIS manuscripts should be submitted in the style of a structured abstract with no more than 300 words. Write an informative abstract that addresses key points. The abstract should not include references. The structured abstract has four labeled sections3.*  **Background:**  **Method:**  **Results:**  **Conclusions:**  **Keywords:** Keyword1, Keyword2, Keyword3, Keyword4, Keyword5. |

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3 Jiang, J., & Tsai, J. C. A. (2019). Constructing an Effective Abstract: Guidelines and New Standards in PAJAIS. *Pacific Asia Journal of the Association for Information Systems*, *11*(3), 1-4.

# Introduction

The rapid development of software technologies, the proliferation of digital devices and networking infrastructure of today, have by and large, augmented user’s capability to generate data (Rada, Ataeib, Khakbizc, & Akbarzadehd, 2017). In the age of information, users are unceasing generators of structured, semi-structured, and unstructured data that if collected and crunched correctly, may reveal game-changing patterns (Ataei & Litchfield, 2020).

The unprecedented proliferation of data have emerged a new ecosystem of technologies; one of these technologies is big data (Mannering, Bhat, Shankar, & Abdel-Aty, 2020; Rad & Ataei, 2017). Big data is a term emerged to describe large amount of data that comes in various forms from different channels. Within the years, big data has attained a lot of attention from academia and industry, and many strive to benefit from this new material (Erevelles, Fukawa, & Swayne, 2016). Howbeit, adopting big data requires the absorption of great deal of complexity and many traditional systems cannot cope with characteristics of this domain.

Based on various reports and surveys published within the last decade, approximately 75% of big data projects have failed (AI, 2019; Gartner, 2014; Manyika et al., 2011; Nash, 2015; Partners, 2019; White, 2019). Among the challenges of adopting big data, the most frequently mentioned are 1) Architectural and system development challenges 2) Rapid technology change challenges and 3) Organizational challenges (Bashari Rad, Akbarzadeh, Ataei, & Khakbiz, 2016; Chen, Kazman, Garbajosa, & Gonzalez, 2017; Singh, Lai, Vejvar, & Cheng, 2019).

Today, majority of big data systems are designed underlying ad-hoc and complicated architectural solutions that do not favour many principles of software engineering (Gorton & Klein, 2015; Hummel, Eichelberger, Giloj, Werle, & Schmid, 2018; Nadal et al., 2017). As the systems grow bigger and new technologies are introduced, software architectures will have harder time to design the suitable solution for any given context. This is a foundation for an immature architecture that is hard to scale and maintain.

Since the approach of ad-hoc design to big data systems is undesirable and leaves many engineers in the dark, there is a need for more software engineering research for big data systems. To this end, this study presents a systematic literature review (SLR) on big data (BD) reference architectures (RAs). Conceptualisation of the system as a reference architecture, helps with understanding of the system’s key components, behaviour, composition and evolution, which in turn affect quality attributes such as maintainability, scalability and performance (Hilliard). Therefore RAs can be a good standardisation artefact and a communication medium that not only results in a concrete architecture for big data systems, but also provide stakeholders with unified elements and symbols to discuss and progress big data projects (Galster & Avgeriou, 2011) (Angelov, Grefen, & Greefhorst, 2009).

# Review Methodology

This research has been designed following the guidelines demonstrated by B. Kitchenham et al. (2009) and Shamseer et al. (2015). B. A. Kitchenham, Dyba, and Jorgensen (2004) framework is used because of its clear instructions on critically appraising evidence for impact, validity and applicability. In addition, to further increase systemacity , transparency and to prevent bias, we used the guidelines provided by Shamseer et al. (2015) on Preferred Reporting Items for Systematic Reviews and Meta-Analysis ( PRISMA ).

Of high importance is the quality of evidence collected in data gathering phase. Here, evidence is defined as the composition of quality literature. SLR has been chosen because it is qualitative research methodology that is aimed at driving knowledge and understanding about the subject matter and the elements around it. Besides, SLR provides with a transparent and reproducible procedure that is in-ling with research question and elicits patterns, relationships, trends, and delineates the overall picture of the subject (Borrego, Foster, & Froyd, 2014).   
  
The main objective of this study is to assess the current state of BD RAs, identify their major architectural components, point out fundamental concepts and discuss their limitations. This objective is achieved in four phases (figure 1). In first phase, research questions are stated, literature are identified and pooled, and exclusion and inclusion criteria are defined. In second phase, literatures are assessed for their quality based on inclusion/exclusion criteria and relevance to research questions. Thirdly, selected pool of literature is coded based on research questions. Lastly, findings are synthesized, trends and patterns understood and delineated.

This SLR is based on the following research questions:

1. What are the fundamental concepts of RAs?
2. How can RAs help BD system development?
3. What are current BD RAs?
4. What are the common architectural components of these BD RAs?
5. What are the limitations of current BD systems?

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| **Figure 1 – PRISMA flowchart of the SLR (Moher, Liberati, Tetzlaff, Altman, & The, 2009)** |

## **Identification**

In identification phase, we selected literature from the years 2010-2021. Most literature selected were within the years 2016-2021 as they provided with most recent and relevant information. Howbeit, some studies dating back to 2010 have been included to provide fundamental knowledge regarding big data systems.

Databases searched were ScienceDirect, IEEE Explore, SpringerLink, AISeL, Elsevier, MIS Quarterly and ACM library. To pursue to goal of finding all literature available on the topic, and to avoid overlooking valuable research, abstract and citation databases and search engines such as Scopus, Google Scholar, Web of Science, and Research Gate have been utilized.

In this phase, it becomes apparent that AISeL and Elsevier are good sources for good quality big data literature, whereas MIS Quarterly provided with the highest quality of Information Systems (IS) research.   
  
A combination of short-tail keywords and long-tail keywords based on research question has been used. These keywords are as followings:

* *Big Data Reference Architectures*
* *Reference Architectures in the domain of Big Data*
* *Reference Architectures and Big Data*
* *Reference Architectures concepts*
* *The concept of Reference Architectures*
* *Reference Architectures in the domain of Big Data*
* *Big Data specific Reference architectures*

A total of 84 studies have been pooled at this stage.

## Screening and Eligibility

In this phase, first we removed duplicates and then screened records for relevancy to our research questions. As a result, 15 studies excluded.

From there on, a full-text assessment took place based on our inclusion and exclusion criteria. These criteria are as following:

**Inclusion Criteria:**

* The study Includes detailed analysis, preferably in practice
* The study showed considerable case studies that aimed to explore data-business context
* Quantitative or qualitative research that illuminates on industry gaps
* Illuminates on RA concepts
* Indicates the current state of RAs in the field of BD and demonstrates possible outcomes
* Explore BD RAs extensively, and discusses the ecosystem, drivers and challenges
* Studies that are within the specified time range
* Studies that are scholarly publications, book, book chapter, thesis, white paper, dissertation, or conference proceedings

**Exclusion Criteria:**

* Studies that are not English
* Very short studies (less than 8 pages)
* Studies that do not aim to explore practice, or discuss practice related concepts
* Studies that provided with low quality information
* Studies that do not directly address the research questions
* Duplicated studies

After excluding papers based on inclusion and exclusion criteria, we assessed studies based on their quality. The quality assessment took place based on the following three factors.

* Is the study rich in terms of its relevant to practice and case studies?
* Does the study provide with ample information/data?
* Is the study discussing contemporary trends in BD RAs?

Regarding the first quality factor, richness is defined as quality and volume of information provided. Primary studies that have an international focused was one of the important facets of the quality assessment. For example, studies that followed research methodologies with good pedigrees and aimed to solve a complex problem in an actual industrial setup with a prototype that is extensively evaluated has been considered rich in terms of relevance to practice and cast studies.   
  
In regard to the second quality factor, studies that revolved around either creation of novel or exploration and examination of current RAs have been perceive as quality researched and have been included in the pool.

Lastly, any study that discussed the recent trends in BD RAs have been added to the pool. A lot of attention has been paid to research methodology and evaluation.

As result of these 57 studies have been included.

## Data Collection and Synthesis

**Document Layout**

***Page Layout***

* A4 size
* 1” (2.54cm) from top and bottom margins
* 1” (2.54cm) from left and right margins

\* All items such as paragraph, table, and figure should fit in this width. Otherwise, use horizontal orientation

***Page Alignment***

* Title and authorship: Centered
* Main text: justified

***Type Font, Spacing, and Indent***

* **Title:** Arial 16, Bold
* **Abstract:** heading: Arial 14, Bold, Italic; text: Arial 11, Italic
* **Keywords:** text: Arial 11, three to five keywords
* **Heading 1** (e.g., introduction): Arial 14, Bold (\* Headings do not have roman numbers)
* **Heading 2:** Arial 12, Bold, Italic (\* Headings do not have roman numbers)
* **Heading 3:** Arial 11, Bold (\* Headings do not have roman numbers)
* Line spacing: single
* All other text: Arial 11
* No indent: additional one-line spacing separates paragraphs

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| **Heading 1**  ***Heading 2***  **Heading 3** |

***Page Number***

* Do not insert the page number.

***Document Format***

Submitted file should be in Microsoft Word format.

**Manuscript Material Ordering**

1. Title
2. Abstract
3. Keywords
4. Main Body

Typically beginning with introduction and ending with conclusions; Section numbering not allowed. Suggested reading: Getting Published in PAJAIS: A Practical Guide from the Perspectives of Editors (Jiang & Tsai, 2019).

1. Acknowledgments\*
2. References
3. Appendices
4. About the Authors\*

\*The author information, acknowledgment, and other information about manuscript should be filled out in the title page of online manuscript systems during review process. PAJAIS logo, header, footer is not allowed to use when the manuscript is under review.

***Elements Style***

1. Footnotes

Footnotes should be avoided if possible. If they are absolutely necessary, footnotes are referred to by superscript number, and displayed on the same page as referred in text; font is Arial 10, single-spaced.

1. In-Text Reference

In-text reference should be referred to in text within parentheses, as follows.

* + - Single author - Fichman (2004)… ; …(Fichman, 2004).
    - Two authors - Lyytinen and King (2004)… ; …(Lyytinen & King, 2004).
    - More than two authors, including first citation - Rossi et al. (2004)… ; (Rossi et al., 2004).

1. Figures

Figures should be inserted at the end of the paragraph in which they are first referred. If there is not sufficient space for full display, it can appear on the next page.

* Figure should be boxed.
* Figure should be aligned center (both vertically and horizontally).
* The font of entries is Arial 10 point
* Figure titles should be at the bottom of the figure box.
* Figure titles should be horizontally left aligned, vertically center aligned; white font, Arial 11 point, bold on a black box; the height of the black box for figure title is 0.7cm” for single line title, and 1.2cm” for double line title.
* Notes should be added at bottom of figure box in Arial 9 point.

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| **Figure 1 - Example of Figure Appearance** |

1. Tables

Tables should be inserted at the end of the paragraph in which they are first referred. If there is not sufficient space for full display, it can appear on the next page.

* Table should be boxed.
* Table should be aligned center (both vertically and horizontally), but the entry alignments are at authors’ disposal.
* The font of entries is Arial 10 point
* Table titles should be at the head of tables.
* Table titles should be horizontally left aligned, vertically center aligned; white font, Arial 11 point, bold on a black box. The height of the black box for table title is 0.7cm for single line title, and 1.2cm for double line title.
* Notes should be added (1) at bottom of table in Arial 9 point or (2) in the last raw of table without indent.

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| **Table 1 - Example of Table Appearance** | | | |
|  |  | Mean | St.dev |
| TEL | 1985 | -- | -- |
| 1991 | 15.9 | 14.0 |
| 2001 | 17.7 | 11.5 |
| CALL | 1985 | -- | -- |
| 1991 | 0.1 | 0.06 |
| 2001 | 0.1 | 0.09 |
| URBAN | 1985 | 64.8 | 20.3 |
| 1991 | 67.1 | 19.6 |
| 2001 | 69.9 | 18.7 |
| TEL = Average monthly telephone subscription cost;  CALL = Average cost of local call;  URBAN = Size of urban population, as a percentage of total population; | | | |

Notes: Year dummies are included in all regressions. Standard errors are in parentheses. \*p<0.1. \*\*p<0.05.\*\*\*p<0.01

**References**

All references should follow the American Psychological Association (2010) guide. The only major difference between APA and PAJAIS style does not use Digital Object Identifiers (DOIs).

* **Journal Papers:**

Hevner, A., March, S., Park, J., & Ram, S. (2004). Design science in information systems research. *MIS Quarterly*, *28*(1), 75-105.

* **Magazine Articles:**

Ackerman (2007). The perfect blend. *Consumer Goods Technology Magazine*. Retrieved from http://consumergoods.edgl.com/media/publicationsarticle/art-mar07-6.pdf

* **Books:**

Alvesson, M., & Deetz, S. (2000). *Doing critical management research*. Sage Publications: Thousand Oaks, California.

* **Book Chapters:**

Keren, G. (1990). Cognitive aids and debiasing methods: Can cognitive pills cure cognitive ills? In J. Caverni, J. Fabre, & M. Gonzalez (Eds.), *Cognitive biases* (pp. 523-555). Amsterdam, North Holland: Elsevier.

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