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| **Link** | **Description** | **Useful for** | **Parts used** |
| [**https://doi.org/10.3390/s20205947**](https://doi.org/10.3390/s20205947) | Does something like our study. Compares 8 methods to find best for each BMS sensor | Previous literature,  FFIL, BFIL, Interpolation | Previous literature in the introduction |
| <https://doi.org/10.1109/icoac.2014.7229721> | Hot Deck comparison Neural Network performance | Comparison in introduction | ‘he results concluded that the machine learning methods outperformed statistical methods with significant improvement in prediction accuracy.’ |
| <https://doi.org/10.1109/sege.2019.8859963> | Imputation of BMS lighting and occupancy data | Use of KNN and precedent K selection | Comparison study in introduction and KNN k selection part |
| <https://doi.org/10.1038/s41598-018-24271-9> | Explanation and use of RNN GRU and LSTM in BMS imputation | Precedent and why GRU is selected | Why RNN over other NN. Performance indicator introduction |
| <https://doi.org/10.1007/978-3-319-07995-019> | Use of KNN and LOCF in time-series data | Precedent KNN and LOCF in time-series imputation | KNN performance indicator in conclusion |
| <https://doi.org/10.2307/2532847> | LOCF is biased and is precedented in this study. **NOT BMS is compared on medical data**. But it is said to have a bias in high-velocity data. | LOCF downsides  Explanation and precedents | LOCF explanation |
| <https://doi.org/10.1016/j.enbuild.2020.109941> | LSTM beats other methods in imputation performance based on RMSE **BMS data** | Introduction comparison start. Explanation RNN and precedent | RNN precedent study |
| <https://www.knmi.nl/nederland-nu/klimatologie/uurgegevens> | KNMI weather API | Access KNMI weather | Data source |
| <https://doi.org/10.1136/bmj.310.6975.298> | When is a normal distribution applicable with large data sets?  Central theorem thingy | Kurtosis research part in eval criteria | Eval criteria Kurtosis part. |
| <https://doi.org/10.4103/aca.aca_157_18> | When Kurtosis and Skewness are applicable in big data sets. | To verify the validity of Kurtosis and Skewness? | Kurtosis part in evaluation criteria |
| <https://doi.org/10.1016/j.dibe.2020.100037> | Multiple methods get used in this paper that we use. Neural Network training model advice with Kurtosis and Skewness. For some reason BMS is called smart building. | Precedent,  Kurtosis,  Neural Network,  Skewness,  Sources,  Traing NN | RNN training kurtosis and skewness part. |
| <https://gmd.copernicus.org/articles/7/1247/2014/> | Why RMSE is actually pretty good compared to MAE | Eval criteria | Not applicable |
| [10.1109/ISGT.2016.7781213](https://doi.org/10.1109/ISGT.2016.7781213) | Use of MAPE and precedent in power data impuatation | Eval criteraia | Not applicable |
| <https://doi.org/10.1080/00949655.2018.1530773> | Very technical paper on kurtosis and skewness pattern in data. Compares three non parametic imp methods kNN? | KNN, missforerest?  Kurtosis and skewness in eval crietia | Was too technical didn’t feel confident enough to quote it. |
| [https://doi.org/10.1037/1082-989X.2.3.292](https://psycnet.apa.org/doi/10.1037/1082-989X.2.3.292) | What is kurtosis? | Could be handy as a source on Kurtosis in a easy way? | Felt a bit too outside scope |
| [10.1109/IC3.2018.853060 8](https://doi.org/10.1109/IC3.2018.8530608) | Arima and RNN are used in time series prediction. Time series in this case network traffic might some what similar to BMS? | RNN, Arima? | Felt out of place in paper |
| https://kdd-milets.github.io/milets2019/papers/milets19\_poster\_3.pdf | Bi-directional RNN  Sensor data  Error grows with sequential missing data | Bi-directional  References for RNN | None |
| https://stefvanbuuren.name/fimd/sec-evaluation.html | Guide to imputation in general |  | Was intended to learn from |
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