

# PERFORMANCE OF THE XGBOOST METHOD IN MEASURING UNCERTAINTY OF SMALL AREA ESTIMATION

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STATISTICS AND DATA SCIENCE STUDY PROGRAM  
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## RINGKASAN

FIDA FARIHA AMATULLAH. Evaluasi Kinerja Metode XGBoost untuk Mengukur Ketidakpastian dalam Pendugaan Area Kecil. Dibimbing oleh ANWAR FITRIANTO dan KHAIRIL ANWAR NOTODIPUTRO.

Survei sampel merupakan salah satu sumber utama dalam menghasilkan indikator sosial ekonomi, termasuk rata-rata pengeluaran per kapita. Namun, pada tingkat area kecil seperti kecamatan, ukuran sampel yang terbatas sering menyebabkan pendugaan langsung memiliki ragam yang besar dan tingkat ketelitian yang rendah. *Small Area Estimation* (SAE) menjadi salah satu solusi untuk meningkatkan presisi pendugaan dengan memanfaatkan informasi tambahan dari peubah penyerta. Metode Fay–Herriot (FH) merupakan pendekatan SAE yang paling banyak digunakan, tetapi memiliki keterbatasan karena mengandalkan asumsi linearitas, normalitas, dan independensi komponen model. Di sisi lain, perkembangan *machine learning* menghadirkan metode *Extreme Gradient Boosting* (XGBoost) yang mampu memodelkan hubungan *non-linear* dan interaksi kompleks tanpa memerlukan asumsi parametrik yang ketat. Penelitian ini bertujuan mengevaluasi kinerja metode XGBoost dalam menduga rata-rata pengeluaran per kapita tingkat kecamatan di Provinsi Jambi, serta mengevaluasi ketidakpastian pendugaan yang dihasilkan.

Data yang digunakan berasal dari Survei Sosial Ekonomi Nasional (SUSENAS) dan Potensi Desa (PODES) Provinsi Jambi tahun 2021. Sebelum pemodelan, dilakukan pembersihan data pengeluaran rumah tangga dengan menghilangkan komponen pengeluaran yang bersifat tidak rutin, kemudian dilakukan penanganan pencilan menggunakan metode *winsorization* pada persentil ke-5 dan ke-95. Hasil eksplorasi menunjukkan bahwa pendugaan langsung memiliki distribusi yang menjulur dan ukuran sampel antar kecamatan yang tidak merata, sehingga mendukung penggunaan pendekatan SAE. Pada model EBLUP-FH dilakukan transformasi logaritma untuk memperbaiki distribusi peubah respons dan memenuhi asumsi model. Hasil pengujian menunjukkan bahwa setelah transformasi, komponen efek acak area dan galat *sampling* menjadi lebih mendekati normal dengan nilai *p-value* masing-masing sebesar 0,5263 dan 0,05009. Namun demikian, asumsi independensi antar komponen belum terpenuhi karena ditemukan korelasi yang cukup tinggi sebesar 0,8457.

Hasil evaluasi menunjukkan bahwa model EBLUP-FH menghasilkan *Mean Square Error* (MSE) sebesar  $1,81 \times 10^9$  dan meningkat menjadi  $2,14 \times 10^9$  setelah *bootstrap*, dengan nilai *Mean Absolute Percentage Error* (MAPE) masing-masing sebesar 1,89% and 2,93%. Nilai MAPE yang sangat rendah menunjukkan bahwa hasil pendugaan EBLUP-FH memiliki tingkat kedekatan yang tinggi terhadap pendugaan langsung. Selain itu, nilai *shrinkage factor* yang sebagian besar berada pada kisaran 0,80–0,95 menunjukkan bahwa EBLUP-FH secara aktif memanfaatkan mekanisme *borrowing strength* untuk meningkatkan stabilitas pendugaan pada kecamatan dengan ukuran sampel kecil. Sementara itu, model XGBoost menunjukkan kemampuan yang baik dalam menangkap pola hubungan yang kompleks antar peubah, namun menghasilkan tingkat ketidakpastian yang lebih besar dibandingkan EBLUP-FH berdasarkan hasil *bootstrap*.

Perbandingan *standard error* menunjukkan bahwa metode EBLUP-FH secara umum menghasilkan pendugaan yang lebih stabil dibandingkan pendugaan langsung. XGBoost mampu menghasilkan estimasi yang kompetitif dan lebih fleksibel terhadap struktur data yang kompleks, namun variasi hasil *bootstrap* menunjukkan bahwa ketidakpastian model masih relatif lebih tinggi. Temuan ini menunjukkan bahwa pada data pengeluaran per kapita tingkat kecamatan di Provinsi Jambi, EBLUP-FH masih menjadi pendekatan yang sangat efektif karena mampu memanfaatkan informasi tambahan melalui *borrowing strength* sehingga menghasilkan estimasi yang presisi dan stabil. Namun demikian, pelanggaran asumsi independensi komponen model mengindikasikan bahwa pendekatan *machine learning* seperti XGBoost tetap memiliki potensi sebagai alternatif atau pelengkap model SAE parametrik pada data sosial ekonomi yang kompleks.

Implikasi dari penelitian ini adalah bahwa penggunaan SAE dapat meningkatkan kualitas indikator pengeluaran per kapita pada wilayah dengan ukuran sampel kecil sehingga mendukung penyusunan kebijakan yang lebih tepat sasaran. Hasil penelitian juga menunjukkan bahwa integrasi metode statistik klasik dan *machine learning* membuka peluang pengembangan SAE yang lebih fleksibel untuk menghadapi karakteristik data sosial ekonomi *modern* yang semakin kompleks dan heterogen.

**Kata kunci:** *Bootstrap*, EBLUP-FH, SAE, *Winsorization*, XGBoost





## SUMMARY

FIDA FARIHA AMATULLAH. Performance of The XGBoost Method in Measuring Uncertainty of Small Area Estimation. Advised by ANWAR FITRIANTO and KHAIRIL ANWAR NOTODIPUTRO.

Sample surveys are one of the primary sources for producing socioeconomic indicators, including average per capita expenditure. However, at small-area levels such as sub-districts, limited sample sizes often lead to direct estimates with large variances and low precision. Small Area Estimation (SAE) provides an effective solution by incorporating auxiliary information to improve estimation accuracy. Among SAE approaches, the Fay–Herriot (FH) model is the most widely used; nevertheless, it relies on assumptions of linearity, normality, and independence of model components. In contrast, recent advances in machine learning have introduced methods such as Extreme Gradient Boosting (XGBoost), which can capture non-linear relationships and complex interactions without requiring strict parametric assumptions. This study aims to evaluate the performance of XGBoost in estimating average per capita expenditure at the sub-district level in Jambi Province and to assess the associated uncertainty.

The study used data from the 2021 National Socioeconomic Survey (SUSENAS) and the 2021 Village Potential Statistics (PODES) of Jambi Province. Prior to modeling, household expenditure data were cleaned by removing non-recurring expenditure components, followed by outlier treatment using Winsorization at the 5th and 95th percentiles. Exploratory analysis revealed a skewed distribution of direct estimates and substantial variation in sample sizes across sub-districts, supporting the application of SAE methods. For the EBLUP-FH model, a logarithmic transformation was applied to the response variable to improve its distributional characteristics and satisfy model assumptions. The results showed that, after transformation, both the area random effect and the sampling error became closer to normality, with p-values of 0.5263 and 0.05009, respectively. However, the independence assumption was not fully satisfied, as a relatively strong correlation of 0.8457 was observed between the two model components.

The evaluation results indicate that the EBLUP-FH model achieved an MSE of  $1.81 \times 10^9$ , which increased to  $2.14 \times 10^9$  after bootstrap resampling, while the corresponding MAPE values were 1.89% and 2.93%, respectively. These low MAPE values indicate a high degree of agreement between the EBLUP-FH estimates and the direct estimates. Furthermore, most shrinkage factor values ranged between 0.80 and 0.95, indicating that EBLUP-FH effectively employed the borrowing-strength mechanism to improve estimate stability in sub-districts with small sample sizes. Meanwhile, XGBoost demonstrated strong capability in capturing complex relationships among variables, although its bootstrap-based uncertainty measures were generally larger than those of EBLUP-FH.

A comparison of standard errors showed that EBLUP-FH generally produced more stable estimates than direct estimation. XGBoost provided competitive results and greater flexibility in handling complex data structures; however, its bootstrap results indicated relatively higher uncertainty. These findings suggest that, for estimating per capita expenditure at the sub-district level in Jambi Province, EBLUP-FH remains a highly effective approach, as it effectively utilizes



auxiliary information through the borrowing-strength mechanism, resulting in precise and stable estimates. Nevertheless, the violation of the independence assumption indicates that machine learning approaches such as XGBoost remain promising alternatives or complementary methods for SAE applications involving complex socioeconomic data.

The findings of this study highlight the importance of SAE methods in improving the quality of per capita expenditure indicators for areas with limited sample sizes, thereby supporting more targeted, evidence-based policymaking. Moreover, the results demonstrate that integrating classical statistical methods with machine learning approaches offers significant potential for developing more flexible SAE frameworks capable of addressing the increasing complexity and heterogeneity of modern socioeconomic data.

**Keywords:** Bootstrap, EBLUP-FH, SAE, Winsorization, XGBoost



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# PERFORMANCE OF THE XGBOOST METHOD IN MEASURING UNCERTAINTY OF SMALL AREA ESTIMATION

**FIDA FARIHA AMATULLAH**

Thesis  
as a partial fulfilment for the requirements of  
Masters degree  
in Statistics and Data Science

**STATISTICS AND DATA SCIENCE STUDY PROGRAM  
SCHOOL OF DATA SCIENCE, MATHEMATICS, AND INFORMATICS  
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Bogor, August 2026

Fida Fariha Amatullah



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