



## Machine Learning Can Mitigate Racial Bias in Orthopaedics: Predicting Overnight Stay After Knee Arthroscopy

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# DISCLOSURES

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- I (and/or my co-authors) have nothing to disclose directly related to this talk.
- I have no conflicts.





## Background

- Past studies have employed machine learning to predict hospital readmission, rates of achieving clinical thresholds, and functional improvements.
- Due to the utility of these models, some orthopaedic researchers have gone as far as publishing these algorithms through online calculators, allowing users to input pertinent patient data and derive a predicted clinical outcome.
- Despite machine learning's immense potential for clinical application, before researchers can offer these tools to a broad and diverse patient population, they must first ensure that their machine learning model performs with high degrees of accuracy irrespective of patient race.

## Purpose

- The purpose of the present study was to develop a racially equitable machine learning model that accurately predicts overnight stay following arthroscopic knee surgery.

## Methods

### Source Data:

- This retrospective study queried the NSQIP database for patients  $\geq 18$  years old who underwent knee arthroscopy between 2011 and 2020.
- Patients who self-identified as White, Black or African American, Asian, Native American or Alaskan Native, or Native Hawaiian or Pacific Islander and had complete demographic, intraoperative, and post-operative outcome data were included in our analysis.

### Model Development:

- All model development and statistical analysis was carried out in Python using packages from Scikit-learn. Predictive analysis was performed using a Random Forest (RF) and performance was optimized through feature selection, hyperparameter tuning with cross validation, and over- and under-sampling techniques with SMOTE.
- The following patient variables to train the RF model included but were not limited to sex, age, BMI, wound classification, ASA class, and type of anesthesia.
- Patients with a hospital length of stay (LOS)  $\geq 1$  day were placed into the overnight stay cohort, while patients with a LOS  $< 1$  represented the same-day discharge cohort. Model performance was assessed using AUC, accuracy, and F1 score

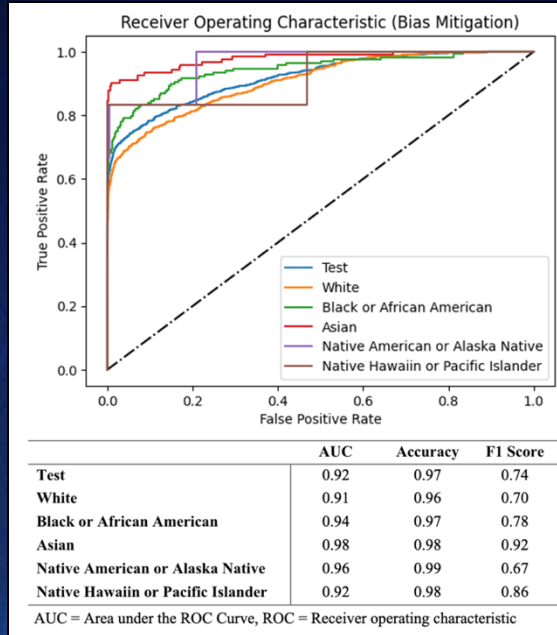


Table 1. Patient Demographics

Patients (n=75,775)	
Same-Day Discharge	5,714 (7.5%)
Overnight Stay	70,061 (92.5%)
Age, years	37.2 $\pm$ 14.4
BMI, kg/m <sup>2</sup>	29.5 $\pm$ 7.3
Sex	
Male	32,101 (42.4%)
Female	43,674 (57.6%)
Racial Self-Identification	
White	59,313 (78.3%)
Black or African American	10,359 (13.7%)
Asian	4,099 (5.4%)
Native American or Alaska Native	1,152 (1.5%)
Native Hawaiian or Pacific Islander	852 (1.1%)
Current Procedural Terminology	
29875	5,291 (7.0%)
29876	6,819 (9.0%)
29877	15,361 (20.3%)
29882	9,025 (11.9%)
29888	39,279 (51.8%)

29875 (arthroscopy knee, surgical; synovectomy, limited)  
29876 (arthroscopy, knee, surgical; synovectomy, major, 2 or more compartments)  
29877 (arthroscopy knee, surgical; debridement/shaving of articular cartilage)  
29882 (arthroscopy, knee surgical; with meniscal repair medial or lateral)  
29888 (arthroscopically aided anterior cruciate ligament repair/augmentation or reconstruction)

## Results

- Of the 75,775 patients who met inclusion criteria, 7.5% (n=5,714) required overnight stay.
- When stratified by racial self-identification, most patients identified as White (78.3%), followed by Black or African American (13.7%), Asian (5.4%), Native American or Native Alaskan (1.5%), and Native Hawaiian or Pacific Islander (1.1%) (**Table 1**).
- Despite a lack of representation among minority patient groups, the racial-equitably trained RF model showed high levels of performance for AUC (mean =  $0.94 \pm 0.03$ ) and accuracy (mean =  $0.97 \pm 0.01$ ) irrespective of patient race. F1 score, however, varied with the model performing best for Asian (0.92), Native Hawaiian or Pacific Islander (0.86), and Black or African American patients (0.78) (**Figure 1**).
- Of the variables included to train the RF model, 5 features accounted for 73% of model development (**Table 2**).

Table 2. Top variables contributing to model development

Variables	Contribution
Arthroscopic knee procedure	36%
Admission quartile	16%
Anesthesia type	13%
ASA class	5%
Days from hospital admission to operation	3%

## Conclusion

- We developed a racially equitable machine learning model that effectively predicted overnight stay in patients undergoing knee arthroscopy.
- As the utilization of machine learning in orthopaedics continues to rise, it is crucial that models intended for clinical application are generalizable and exhibit high levels of performance for minority patients.
- The onus is on orthopaedic researchers to assess the internal validity of their machine learning models before offering them to a broader patient population. Failing to do so may result in inequitable clinical models that exacerbate poor health outcomes among minority patients.





# REFERENCES

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