

# Predicting Non-Home Discharge following Primary Total Hip Arthroplasty Using an Artificial Neural Network to Identify the Most Important Variables

Caillin Marquardt<sup>1</sup>, Theodore Quan<sup>2</sup>, Alex Gu<sup>3</sup>, Jordan Cohen<sup>4</sup>, Amil Raj Agarwal, Tara G Harmon, Gregory Golladay<sup>5</sup>, Joshua Campbell<sup>6</sup>, Savyasachi C Thakkar

<sup>1</sup>GWUMC - Ortho Surgery, <sup>2</sup>George Washington University, <sup>3</sup>George Washington University School of Medicine An, <sup>4</sup>University of Pennsylvania Hospital, <sup>5</sup>VCU Health, <sup>6</sup>The GW Medical Faculty Associates

**INTRODUCTION:** Total hip arthroplasty (THA) has continued to increase in incidence. With expanding focus on value-based care and adoption of bundled payments at some facilities, cost containment measures including discharging patients to home as soon as possible postoperatively have gained attention. Furthermore, discharge to home is often desired by patients and may improve their overall satisfaction and experience with recovery. Therefore, it is important to determine patient characteristics associated with discharge disposition. Artificial neural networks (ANNs) are computational models that change based on the input and output of data. The purpose of this study was to develop and utilize an ANN model to determine the most important non-modifiable and modifiable factors which can predispose patients to be discharged to a non-home destination following primary THA.

**METHODS:** From 2016-2019, the National Surgical Quality Improvement Program database was used to identify patients who underwent primary THA. Overall, 22 demographic, comorbidity, preoperative, and intraoperative variables were analyzed in this study. Chi-squared tests and analysis of variance were performed to compare the various variables between the cohort of patients who were discharged to a home destination and those who were not discharged to a home location. Variables that were statistically significant, with a p-value < 0.05, were inputted into the ANN model. The ANN module of Statistical Package for Social Sciences was used for analysis.

**RESULTS:** In total, 124,691 patients were analyzed in the study, of which 19,275 (15.5%) were discharged to a non-home destination (Table 1). The ANN reached a Receiver Operating Characteristic (ROC) Area-Under-the-Curve (AUC) of 0.793, which is considered to be acceptable/excellent accuracy (Figure 1). It was found that 21 out of the 22 variables analyzed were important to predict non-home discharge. The five most important variables which helped to predict non-home discharge following primary THA were age (older; importance = 0.204), operative time (increased; importance = 0.159), preoperative hematocrit (decreased; importance = 0.095), functional status (dependent; importance = 0.080), and preoperative international normalized ratio (INR) (increased; importance = 0.078) (Figure 2).

**DISCUSSION AND CONCLUSION:** The present study used an ANN model and identified several significant factors which can help predict patients being discharged to a non-home location following primary THA, including age, operative time, preoperative hematocrit, functional status, and preoperative INR. Clinicians should be aware of these important variables and explore reductions to non-home discharges through preoperative patient optimization, where possible. Particularly, several of these factors, such as hematocrit and INR, can be evaluated preoperatively through laboratory tests and medically managed prior to the surgery. This model also provides a simple representation of the importance of each factor in predicting discharge to home, which could be easier for surgeons to better counsel their patients preoperatively. Ultimately, evaluating factors that impact patient discharge can help optimize costs and direct social work resources more effectively to patients more likely to need support. Our study demonstrated that ANN modeling represents a novel and unique approach to determine which modifiable and non-modifiable factors can assist in predicting discharge status following primary THA.

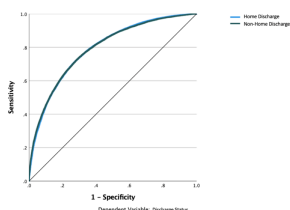


Figure 1. Receiver Operating Characteristic curve of the Artificial Neural Network model in predicting Non-Home Discharge

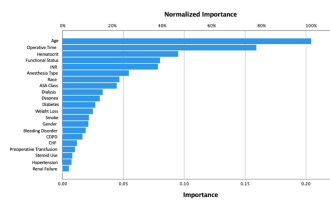


Figure 2. Importance of Variables used by the Neural Network model to predict Non-Home Discharge

Patient Characteristics	Total		Home Discharge		Non-Home Discharge	
	N	%	N	%	N	%
<b>Sex</b>	124,691		19,275		14,275	
Male	68,395	54.9%	10,503	54.5%	12,059	84.5%
Female	56,296	45.1%	8,772	45.5%	5,216	35.5%
<b>Race</b>						
White	108,805	87.3%	18,216	94.3%	13,527	94.7%
Black or African American	12,837	10.3%	2,004	10.4%	2,111	14.8%
Hispanic	4,768	3.8%	3,473	17.9%	826	5.8%
Asian	2,101	1.7%	1,828	9.5%	313	2.2%
Native American or Alaska	588	0.5%	302	1.6%	197	1.4%
<b>Age (years)</b>	62,511.2		63,010.0		62,511.0	
<b>BMI (kg/m<sup>2</sup>)</b>	30.543.2		31.346.1		30.343.1	
<b>Comorbidity</b>						
Diabetes	42,867	34.4%	11,807	61.3%	11,068	77.6%
Neurological	40,889	32.8%	16,348	84.9%	4,706	33.7%
Respiratory	2,838	2.3%	1,248	6.5%	282	2.0%
MAC	12,298	9.8%	19,908	103%	1,699	12%
<b>Medication</b>						
Non-Insulin-Dependent Diabetes Mellitus	12,872	10.3%	3,366	17.5%	2,566	18.0%
Insulin-Dependent Diabetes Mellitus	3,064	2.5%	2,753	14.3%	1,151	8.1%
<b>Smoker</b>	19,719	15.8%	19,274	100%	2,347	17.4%
<b>Alcohol</b>						
Drinks on occasion	5,203	4.2%	3,629	18.8%	1,654	11.6%
Drinks a lot	238	0.2%	123	6.4%	108	0.8%
<b>Functional Status</b>						
Partially dependent	1,095	0.9%	813	4.2%	1,008	7.1%
Dependent	28	0.0%	28	0.1%	0	0%
<b>Chronic Obstructive Pulmonary Disease</b>	3,101	2.5%	3,204	16.7%	1,257	9.0%
<b>Cardiovascular Disease</b>						
Coronary Heart Failure	52	0.0%	20	0.1%	24	0.2%
Myocardial Infarction	76,648	61.5%	17,267	89.6%	11,085	77.8%
Stroke	74	0.0%	33	0.2%	33	0.2%
<b>Diabetes</b>	58	0.0%	102	0.5%	165	1.2%
<b>Chronic Kidney Disease</b>	4,031	3.2%	3,587	18.6%	1,066	7.6%
<b>Bone Weight Loss</b>	275	0.2%	108	0.5%	108	0.8%
<b>Bleeding Disorder</b>	2,767	2.2%	1,806	9.3%	861	6.1%
<b>Preoperative comorbidity</b>						
ASA Class I/II	71	0.0%	76	0.4%	97	0.7%
<b>Preoperative Assessment (h)</b>	74,294	59.7%	43,888	228%	13,206	93.9%
<b>Preoperative Assessment (m)</b>	4,112,024.0		41,002,017.0		2,024,210.0	
<b>Preoperative ASA</b>	1,614,24		1,014,24		1,024,27	
<b>Operative Time (min)</b>	62,5363.1		9,133,04		97,544.2	

ASA, American Society of Anesthesiologists; BMI, Body Mass Index; MAC, Modified Alcohol Consumption; N, Number of Patients; % , Percentage of Patients