Matthew Churpek

List of Publications by Year in descending order

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88 5,307 37 69
papers citations h-index g-index

91 91 91 6899 all docs docs citations times ranked citing authors

#	Article	IF	CITATIONS
1	Quick Sepsis-related Organ Failure Assessment, Systemic Inflammatory Response Syndrome, and Early Warning Scores for Detecting Clinical Deterioration in Infected Patients outside the Intensive Care Unit. American Journal of Respiratory and Critical Care Medicine, 2017, 195, 906-911.	5.6	496
2	Multicenter Comparison of Machine Learning Methods and Conventional Regression for Predicting Clinical Deterioration on the Wards. Critical Care Medicine, 2016, 44, 368-374.	0.9	423
3	Incidence and Prognostic Value of the Systemic Inflammatory Response Syndrome and Organ Dysfunctions in Ward Patients. American Journal of Respiratory and Critical Care Medicine, 2015, 192, 958-964.	5.6	267
4	The Development of a Machine Learning Inpatient Acute Kidney Injury Prediction Model*. Critical Care Medicine, 2018, 46, 1070-1077.	0.9	214
5	Multicenter Development and Validation of a Risk Stratification Tool for Ward Patients. American Journal of Respiratory and Critical Care Medicine, 2014, 190, 649-655.	5.6	203
6	Big Data and Data Science in Critical Care. Chest, 2018, 154, 1239-1248.	0.8	184
7	Understanding Why Patients With COPD Get Readmitted. Chest, 2015, 147, 1219-1226.	0.8	179
8	Comparison of variable selection methods for clinical predictive modeling. International Journal of Medical Informatics, 2018, 116, 10-17.	3.3	160
9	Derivation of a cardiac arrest prediction model using ward vital signs*. Critical Care Medicine, 2012, 40, 2102-2108.	0.9	154
10	Predicting Cardiac Arrest on the Wards. Chest, 2012, 141, 1170-1176.	0.8	137
11	Inherited mutations in cancer susceptibility genes are common among survivors of breast cancer who develop therapyâ€related leukemia. Cancer, 2016, 122, 304-311.	4.1	129
12	Inherited mutations in cancer susceptibility genes are common among survivors of breast cancer who develop therapyâ€related leukemia. Cancer, 2016, 122, 304-311. The value of vital sign trends for detecting clinical deterioration on the wards. Resuscitation, 2016, 102, 1-5.	3.0	129
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12	develop therapyâ€related leukemia. Cancer, 2016, 122, 304-311. The value of vital sign trends for detecting clinical deterioration on the wards. Resuscitation, 2016, 102, 1-5.	3.0	126
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12 13 14	develop therapyâ€related leukemia. Cancer, 2016, 122, 304-311. The value of vital sign trends for detecting clinical deterioration on the wards. Resuscitation, 2016, 102, 1-5. Rapid response systems. Resuscitation, 2018, 128, 191-197. Using Electronic Health Record Data to Develop and Validate a Prediction Model for Adverse Outcomes in the Wards*. Critical Care Medicine, 2014, 42, 841-848. Inherited predisposition to breast cancer among African American women. Breast Cancer Research and	3.0 3.0 0.9	126 125 117
12 13 14	develop therapyâ€related leukemia. Cancer, 2016, 122, 304-311. The value of vital sign trends for detecting clinical deterioration on the wards. Resuscitation, 2016, 102, 1-5. Rapid response systems. Resuscitation, 2018, 128, 191-197. Using Electronic Health Record Data to Develop and Validate a Prediction Model for Adverse Outcomes in the Wards*. Critical Care Medicine, 2014, 42, 841-848. Inherited predisposition to breast cancer among African American women. Breast Cancer Research and Treatment, 2015, 149, 31-39.	3.0 3.0 0.9	126 125 117

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19	Identifying Patients With Sepsis on the Hospital Wards. Chest, 2017, 151, 898-907.	0.8	94
20	Association between intensive care unit transfer delay and hospital mortality: A multicenter investigation. Journal of Hospital Medicine, 2016, 11, 757-762.	1.4	90
21	Trends in Survival After In-Hospital CardiacÂArrest During Nights and Weekends. Journal of the American College of Cardiology, 2018, 71, 402-411.	2.8	90
22	Development of a Multicenter Ward–Based AKI Prediction Model. Clinical Journal of the American Society of Nephrology: CJASN, 2016, 11, 1935-1943.	4.5	88
23	Sharing ICU Patient Data Responsibly Under the Society of Critical Care Medicine/European Society of Intensive Care Medicine Joint Data Science Collaboration: The Amsterdam University Medical Centers Database (AmsterdamUMCdb) Example*. Critical Care Medicine, 2021, 49, e563-e577.	0.9	87
24	Comparison of Early Warning Scoring Systems for Hospitalized Patients With and Without Infection at Risk for In-Hospital Mortality and Transfer to the Intensive Care Unit. JAMA Network Open, 2020, 3, e205191.	5.9	81
25	Practice Changes at U.S.ÂTransplantÂCenters After the NewÂAdult Heart Allocation Policy. Journal of the American College of Cardiology, 2020, 75, 2906-2916.	2.8	7 5
26	Differences in Vital Signs Between Elderly and Nonelderly Patients Prior to Ward Cardiac Arrest. Critical Care Medicine, 2015, 43, 816-822.	0.9	71
27	Predicting clinical deterioration in the hospital: The impact of outcome selection. Resuscitation, 2013, 84, 564-568.	3.0	66
28	The impact of vaccination to control COVID-19 burden in the United States: A simulation modeling approach. PLoS ONE, 2021, 16, e0254456.	2.5	62
29	Investigating the Impact of Different Suspicion of Infection Criteria on the Accuracy of Quick Sepsis-Related Organ Failure Assessment, Systemic Inflammatory Response Syndrome, and Early Warning Scores*. Critical Care Medicine, 2017, 45, 1805-1812.	0.9	60
30	A Prospective Study of Nighttime Vital Sign Monitoring Frequency and Risk of Clinical Deterioration. JAMA Internal Medicine, 2013, 173, 1554.	5.1	57
31	Effect of Timing of and Adherence to Social Distancing Measures on COVID-19 Burden in the United States. Annals of Internal Medicine, 2021, 174, 50-57.	3.9	57
32	Relationship Between ICU Bed Availability, ICU Readmission, and Cardiac Arrest in the General Wards. Critical Care Medicine, 2014, 42, 2037-2041.	0.9	52
33	Real-Time Risk Prediction on the Wards: A Feasibility Study. Critical Care Medicine, 2016, 44, 1468-1473.	0.9	52
34	Predictors of In-Hospital Mortality After Rapid Response Team Calls in a 274 Hospital Nationwide Sample*. Critical Care Medicine, 2018, 46, 1041-1048.	0.9	49
35	Validating the Electronic Cardiac Arrest Risk Triage (eCART) Score for Risk Stratification of Surgical Inpatients in the Postoperative Setting. Annals of Surgery, 2019, 269, 1059-1063.	4.2	48
36	Impact of Vasoactive Medications on ICU-Acquired Weakness in Mechanically Ventilated Patients. Chest, 2018, 154, 781-787.	0.8	47

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37	Association Between Survival and Time of Day for Rapid Response Team Calls in a National Registry. Critical Care Medicine, 2017, 45, 1677-1682.	0.9	43
38	Implications of Centers for Medicare & Medicaid Services Severe Sepsis and SepticÂShock Early Management Bundle andÂlnitial Lactate Measurement on the Management of Sepsis. Chest, 2018, 154, 302-308.	0.8	41
39	Phenotypic Clusters Predict Outcomes in a Longitudinal Interstitial Lung Disease Cohort. Chest, 2018, 153, 349-360.	0.8	40
40	Patient Outcomes and Cost-Effectiveness of a Sepsis Care Quality Improvement Program in a Health System*. Critical Care Medicine, 2019, 47, 1371-1379.	0.9	38
41	Association Between In-Hospital Critical Illness Events and Outcomes in Patients on the Same Ward. JAMA - Journal of the American Medical Association, 2016, 316, 2674.	7.4	33
42	Characteristics of Rapid Response Calls in the United States: An Analysis of the First 402,023 Adult Cases From the Get With the Guidelines Resuscitation-Medical Emergency Team Registry. Critical Care Medicine, 2019, 47, 1283-1289.	0.9	33
43	Skewed Lung CCR4 to CCR6 CD4+ T Cell Ratio in Idiopathic Pulmonary Fibrosis Is Associated with Pulmonary Function. Frontiers in Immunology, 2016, 7, 516.	4.8	29
44	Comparison of mentalâ€status scales for predicting mortality on the general wards. Journal of Hospital Medicine, 2015, 10, 658-663.	1.4	28
45	Obstructive sleep apnea and adverse outcomes in surgical and nonsurgical patients on the wards. Journal of Hospital Medicine, 2015, 10, 592-598.	1.4	25
46	Association of Transplant Center With Survival Benefit Among Adults Undergoing Heart Transplant in the United States. JAMA - Journal of the American Medical Association, 2019, 322, 1789.	7.4	25
47	Characteristics and outcomes of maternal cardiac arrest: A descriptive analysis of Get with the guidelines data. Resuscitation, 2018, 132, 17-20.	3.0	23
48	Risk factors for infection and evaluation of Sepsis-3 in patients with trauma. American Journal of Surgery, 2019, 218, 851-857.	1.8	23
49	Age-dependent trends in survival after adult in-hospital cardiac arrest. Resuscitation, 2020, 151, 189-196.	3.0	23
50	Racial disparities in outcomes following PEA and asystole in-hospital cardiac arrests. Resuscitation, 2015, 87, 69-74.	3.0	22
51	Geographic Variation in the Treatment of U.S. Adult Heart Transplant Candidates. Journal of the American College of Cardiology, 2018, 71, 1715-1725.	2.8	21
52	Do Sex Differences Exist in the Establishment of "Do Not Attempt Resuscitation―Orders and Survival in Patients Successfully Resuscitated From Inâ€Hospital Cardiac Arrest?. Journal of the American Heart Association, 2020, 9, e014200.	3.7	21
53	Trends, Cost, and Mortality From Sepsis After Trauma in the United States: An Evaluation of the National Inpatient Sample of Hospitalizations, 2012–2016. Critical Care Medicine, 2020, 48, 1296-1303.	0.9	20
54	Predicting clinical deterioration with Q-ADDS compared to NEWS, Between the Flags, and eCART track and trigger tools. Resuscitation, 2020, 153, 28-34.	3.0	20

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55	Accuracy Comparisons between Manual and Automated Respiratory Rate for Detecting Clinical Deterioration in Ward Patients. Journal of Hospital Medicine, 2018, 13, 486-487.	1.4	20
56	Combining patient visual timelines with deep learning to predict mortality. PLoS ONE, 2019, 14, e0220640.	2.5	19
57	Variation in Best Practice Measures in Patients With Severe Hospital-Acquired Acute Kidney Injury: A Multicenter Study. American Journal of Kidney Diseases, 2021, 77, 547-549.	1.9	19
58	Association Between Opioid and Benzodiazepine Use and Clinical Deterioration in Ward Patients. Journal of Hospital Medicine, 2017, 12, 428-434.	1.4	18
59	Analysis of the Proportional Hazards Model With Sparse Longitudinal Covariates. Journal of the American Statistical Association, 2015, 110, 1187-1196.	3.1	17
60	The Value of Clinical Judgment in the Detection of Clinical Deterioration. JAMA Internal Medicine, 2015, 175, 456.	5.1	11
61	Moving Beyond Single-Parameter Early Warning Scores for Rapid Response System Activation*. Critical Care Medicine, 2016, 44, 2283-2285.	0.9	11
62	Potential impact of a shock requirement on adult heart allocation. Journal of Heart and Lung Transplantation, 2017, 36, 1013-1016.	0.6	11
63	Sequential Organ Failure Assessment Score Modified for Recent Infection in Patients With Hematologic Malignant Tumors and Severe Sepsis. American Journal of Critical Care, 2016, 25, 409-417.	1.6	10
64	Validation of Early Warning Scores at Two Long-Term Acute Care Hospitals. Critical Care Medicine, 2019, 47, e962-e965.	0.9	10
65	Comparison of Machine Learning Methods for Predicting Outcomes After In-Hospital Cardiac Arrest. Critical Care Medicine, 2022, 50, e162-e172.	0.9	8
66	Risk Factors for Cardiovascular Collapse during Tracheal Intubation of Critically III Adults. Annals of the American Thoracic Society, 2020, 17, 1021-1024.	3.2	8
67	Association of COVID-19 Infection With Survival After In-Hospital Cardiac Arrest Among US Adults. JAMA Network Open, 2022, 5, e220752.	5.9	8
68	Allergic Immune Diseases and the Risk of Mortality Among Patients Hospitalized for Acute Infection*. Critical Care Medicine, 2019, 47, 1735-1742.	0.9	6
69	Sifting through the heterogeneity of the Rapid Response System literature. Resuscitation, 2012, 83, 1419-1420.	3.0	5
70	The Laboratory-Based Intermountain Validated Exacerbation (LIVE) Score Identifies Chronic Obstructive Pulmonary Disease Patients at High Mortality Risk. Frontiers in Medicine, 2018, 5, 173.	2.6	5
71	Detecting Sepsis: Are Two Opinions Better Than One?. Journal of Hospital Medicine, 2017, 12, 256-258.	1.4	5
72	Life Expectancy Predictions for Older Diabetic Patients as Estimated by Physicians and a Prognostic Model. MDM Policy and Practice, 2017, 2, 238146831771371.	0.9	4

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73	Measuring and Rewarding Quality in the ICU: The Yardstick Is Not As Straight As We Wish. American Journal of Respiratory and Critical Care Medicine, 2012, 185, 3-4.	5.6	3
74	In search of the optimal rapid response system bundle. Journal of Hospital Medicine, 2015, 10, 411-411.	1.4	3
75	1584: PATIENTS WITH TYPE 2-MEDIATED IMMUNE DISEASES ARE PROTECTED FROM DYING OF SEPSIS. Critical Care Medicine, 2019, 47, 767-767.	0.9	3
76	Safety and efficacy of catheter-directed therapy versus anticoagulation alone in a higher-risk acute pulmonary embolism population. Journal of Thrombosis and Thrombolysis, 2021, 52, 1151-1159.	2.1	2
77	In response to "Obstructive sleep apnea and adverse outcomes in surgical and nonsurgical patients on the wards― Journal of Hospital Medicine, 2016, 11, 157-157.	1.4	1
78	Electronic cardiac arrest triage score best predicts mortality after intervention in patients with massive and submassive pulmonary embolism. Catheterization and Cardiovascular Interventions, 2018, 92, 366-371.	1.7	1
79	Reply to Leijte et al.: Fever in Sepsis: Still a Hot Topic. American Journal of Respiratory and Critical Care Medicine, 2019, 200, 264-265.	5.6	1
80	Recommended Reading from the University of Chicago Pulmonary and Critical Care Fellows. American Journal of Respiratory and Critical Care Medicine, 2010, 182, 1453-1454.	5.6	0
81	Using Electronic Health Record Data to Develop and Validate a Prediction Model for Adverse Outcomes on the Wards. Chest, 2012, 142, 279A.	0.8	0
82	Obstructive Sleep Apnea as a Predictor of Clinical Deterioration in Hospitalized Patients on the Wards. Chest, 2014, 146, 502A.	0.8	0
83	Testing the functional assessment of mentation: A mobile application based assessment of mental status. Journal of Hospital Medicine, $2016, 11, 463-466$.	1.4	0
84	Response. Chest, 2018, 154, 1462.	0.8	0
85	RISKS OF CATHETER-DIRECTED THERAPY VERSUS CONSERVATIVE THERAPY IN A HIGH RISK, PROPENSITY-SCORE MATCHED PULMONARY EMBOLISM PATIENT POPULATION. Journal of the American College of Cardiology, 2018, 71, A1945.	2.8	0
86	Reply. Journal of the American College of Cardiology, 2018, 72, 703-704.	2.8	0
87	PREDICTING BACTEREMIA USING ELECTRONIC HEALTH RECORD DATA. Chest, 2019, 156, A1607.	0.8	0
88	TEMPERATURE TRAJECTORY MAY BE AN INDICATOR OF BACTEREMIA IN PATIENTS WITH SEPTIC SHOCK. Chest, 2020, 158, A592-A593.	0.8	0