Jayasree Chakraborty

List of Publications by Year in descending order

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| # | Article | IF | CITATIONS |
|----|---|-----|-----------|
| 1 | Machine learning radiomics can predict early liver recurrence after resection of intrahepatic cholangiocarcinoma. Hpb, 2022, 24, 1341-1350. | 0.3 | 7 |
| 2 | Rectal MRI radiomics inter- and intra-reader reliability: should we worry about that?. Abdominal Radiology, 2022, , . | 2.1 | 0 |
| 3 | A multi features based background modeling approach for moving object detection. Optik, 2022, 260, 168980. | 2.9 | 4 |
| 4 | CT radiomics to predict early hepatic recurrence after resection for intrahepatic cholangiocarcinoma. , 2022, , . | | 0 |
| 5 | Recurrence After Resection of Pancreatic Cancer: Can Radiomics Predict Patients at Greatest Risk of Liver Metastasis?. Annals of Surgical Oncology, 2022, 29, 4962-4974. | 1.5 | 11 |
| 6 | ASO Visual Abstract: Recurrence After Resection of Pancreatic Cancer – Can Radiomics Predict Patients at Greatest Risk of LiverÂMetastasis?. Annals of Surgical Oncology, 2022, , . | 1.5 | 0 |
| 7 | Enhancement of Hazy Images Using Atmospheric Light Estimation Technique. Journal of Circuits, Systems and Computers, 2021, 30, 2150078. | 1.5 | 1 |
| 8 | MhURI:A Supervised Segmentation Approach to Leverage Salient Brain Tissues in Magnetic Resonance Images. Computer Methods and Programs in Biomedicine, 2021, 200, 105841. | 4.7 | 10 |
| 9 | Differences in Liver Parenchyma are Measurable with CT Radiomics at Initial Colon Resection in Patients that Develop Hepatic Metastases from Stage II/III Colon Cancer. Annals of Surgical Oncology, 2021, 28, 1982-1989. | 1.5 | 15 |
| 10 | Quantitative Computed Tomography Image Analysis to Predict Pancreatic Neuroendocrine Tumor Grade. JCO Clinical Cancer Informatics, 2021, 5, 679-694. | 2.1 | 5 |
| 11 | Radiomic feature reproducibility in contrast-enhanced CT of the pancreas is affected by variabilities in scan parameters and manual segmentation. European Radiology, 2020, 30, 195-205. | 4.5 | 58 |
| 12 | Multi-Resolution Analysis of Edge-Texture Features for Mammographic Mass Classification. Journal of Circuits, Systems and Computers, 2020, 29, 2050156. | 1.5 | 0 |
| 13 | Multimodal radiomics and cyst fluid inflammatory markers model to predict preoperative risk in intraductal papillary mucinous neoplasms. Journal of Medical Imaging, 2020, 7, 1. | 1.5 | 8 |
| 14 | A combined radiomics and cyst fluid inflammatory markers model to predict preoperative risk in pancreatic cystic lesions. , 2020, , . | | 1 |
| 15 | Quantitative imaging features of pretreatment CT predict volumetric response to chemotherapy in patients with colorectal liver metastases. European Radiology, 2019, 29, 458-467. | 4.5 | 10 |
| 16 | Preoperative risk prediction for intraductal papillary mucinous neoplasms by quantitative CT image analysis. Hpb, 2019, 21, 212-218. | 0.3 | 36 |
| 17 | A Screening CAD Tool for the Detection of Microcalcification Clusters in Mammograms. Journal of Digital Imaging, 2019, 32, 728-745. | 2.9 | 6 |
| 18 | Radiomics-based prediction of microsatellite instability in colorectal cancer at initial computed tomography evaluation. Abdominal Radiology, 2019, 44, 3755-3763. | 2.1 | 74 |

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|----|---|-----|-----------|
| 19 | CT radiomics associations with genotype and stromal content in pancreatic ductal adenocarcinoma. Abdominal Radiology, 2019, 44, 3148-3157. | 2.1 | 37 |
| 20 | Brain Tumor Classification Using ResNet-101 Based Squeeze and Excitation Deep Neural Network. , 2019, , . | | 60 |
| 21 | Dynamic background modeling using intensity and orientation distribution of video sequence. Multimedia Tools and Applications, 2019, 78, 22537-22554. | 3.9 | 6 |
| 22 | A Deep Adaptive Convolutional Network for Brain Tumor Segmentation from Multimodal MR Images. , 2019, , . | | 0 |
| 23 | Computer-Aided Detection of Mammographic Masses Using Hybrid Region Growing Controlled by Multilevel Thresholding. Journal of Medical and Biological Engineering, 2019, 39, 352-366. | 1.8 | 15 |
| 24 | Abstract 2444: The use of CT radiomics to predict immune infiltrate in pancreatic ductal adenocarcinoma. , 2019, , . | | 1 |
| 25 | Abstract 2444: The use of CT radiomics to predict immune infiltrate in pancreatic ductal adenocarcinoma. , 2019, , . | | Ο |
| 26 | Quantitative Imaging Features and Postoperative Hepatic Insufficiency: A Multi-Institutional Expanded Cohort. Journal of the American College of Surgeons, 2018, 226, 835-843. | 0.5 | 7 |
| 27 | Computer-aided detection and diagnosis of mammographic masses using multi-resolution analysis of oriented tissue patterns. Expert Systems With Applications, 2018, 99, 168-179. | 7.6 | 38 |
| 28 | Survival Prediction in Pancreatic Ductal Adenocarcinoma by Quantitative Computed Tomography Image Analysis. Annals of Surgical Oncology, 2018, 25, 1034-1042. | 1.5 | 92 |
| 29 | Neighborhood Structural Similarity Mapping for the Classification of Masses in Mammograms. IEEE Journal of Biomedical and Health Informatics, 2018, 22, 826-834. | 6.3 | 26 |
| 30 | Edge Weighted Local Texture Features for the Categorization of Mammographic Masses. Journal of Medical and Biological Engineering, 2018, 38, 457-468. | 1.8 | 8 |
| 31 | <scp>CT</scp> radiomics to predict highâ€risk intraductal papillary mucinous neoplasms of the pancreas. Medical Physics, 2018, 45, 5019-5029. | 3.0 | 76 |
| 32 | Short-term reproducibility of radiomic features in liver parenchyma and liver malignancies on contrast-enhanced CT imaging. Abdominal Radiology, 2018, 43, 3271-3278. | 2.1 | 46 |
| 33 | Influence of CT acquisition and reconstruction parameters on radiomic feature reproducibility. Journal of Medical Imaging, 2018, 5, 1. | 1.5 | 61 |
| 34 | Multi-resolution analysis using integrated microscopic configuration with local patterns for benign-malignant mass classification. , 2018, , . | | 3 |
| 35 | Deep convolutional neural network for the classification of hepatocellular carcinoma and intrahepatic cholangiocarcinoma. , 2018, , . | | 6 |
| 36 | Quantitative CT analysis for the preoperative prediction of pathologic grade in pancreatic neuroendocrine tumors. , 2018, , . | | 0 |

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|----|--|-----|-----------|
| 37 | Preoperative assessment of microvascular invasion in hepatocellular carcinoma. , 2017, , . | | 0 |
| 38 | Quantification of CT images for the classification of high- and low-risk pancreatic cysts. Proceedings of SPIE, 2017, , . | 0.8 | 4 |
| 39 | Preoperative Prediction of Microvascular Invasion in Hepatocellular Carcinoma Using Quantitative Image Analysis. Journal of the American College of Surgeons, 2017, 225, 778-788e1. | 0.5 | 66 |
| 40 | Preoperative prediction of microvascular invasion in hepatocellular carcinoma using quantitative image analysis. Hpb, 2017, 19, S48. | 0.3 | 2 |
| 41 | Behind the cyst: predicting grade of dysplasia in intraductal papillary mucinous neoplasms (IPMNs) by quantitative image analysis. Hpb, 2017, 19, S22. | 0.3 | 0 |
| 42 | Video error concealment through 3-D face model. Multimedia Tools and Applications, 2017, 76, 23931-23955. | 3.9 | 1 |
| 43 | Quantitative Imaging Features of Preoperative Computed Tomography Images Predict Post-Hepatectomy Liver Insufficiency: A Multi-Institutional Expansion Cohort. Journal of the American College of Surgeons, 2017, 225, S137. | 0.5 | 0 |
| 44 | Analysis of 2D singularities for mammographic mass classification. IET Computer Vision, 2017, 11, 22-32. | 2.0 | 13 |
| 45 | Texture analysis of gradient images for benign-malignant mass classification. , 2017, , . | | 2 |
| 46 | Preliminary study of tumor heterogeneity in imaging predicts two year survival in pancreatic cancer patients. PLoS ONE, 2017, 12, e0188022. | 2.5 | 69 |
| 47 | Texture analysis for survival prediction of pancreatic ductal adenocarcinoma patients with neoadjuvant chemotherapy. , 2016, , . | | 4 |
| 48 | Benign-malignant mass classification in mammogram using edge weighted local texture features. , 2016, , . | | 1 |
| 49 | A Study of Different Texture Features Based on Local Operator for Benign-malignant Mass Classification. Procedia Computer Science, 2016, 93, 389-395. | 2.0 | 14 |
| 50 | Classification of benign and malignant masses in mammograms using multi-resolution analysis of oriented patterns. , 2015, , . | | 18 |
| 51 | Detection of the nipple in mammograms with Gabor filters and the Radon transform. Biomedical Signal Processing and Control, 2015, 15, 80-89. | 5.7 | 8 |
| 52 | Video error concealment using Speeded Up Robust Features and affine transformation. , 2014, , . | | 1 |
| 53 | Face detection using skin color modeling and geometric feature. , 2014, , . | | 7 |
| 54 | A Heuristic Approach to Automated Nipple Detection in Digital Mammograms. Journal of Digital Imaging, 2013, 26, 932-940. | 2.9 | 16 |

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| 55 | Automatic localization of the nipple in mammograms using Gabor filters and the Radon transform. , 2013, , . | | 1 |
| 56 | Measures of divergence of oriented patterns for the detection of architectural distortion in prior mammograms. International Journal of Computer Assisted Radiology and Surgery, 2013, 8, 527-545. | 2.8 | 23 |
| 57 | Automatic characterization of masses in mammograms. , 2013, , . | | 14 |
| 58 | Statistical measures of orientation of texture for the detection of architectural distortion in prior mammograms of interval-cancer. Journal of Electronic Imaging, 2012, 21, 033010-1. | 0.9 | 27 |
| 59 | Detection of architectural distortion in prior mammograms using statistical measures of orientation of texture. Proceedings of SPIE, 2012, , . | 0.8 | 10 |
| 60 | Detection of masses in mammograms using region growing controlled by multilevel thresholding. , 2012, , . | | 8 |
| 61 | Detection of architectural distortion using coherence in relation to the expected orientation of breast tissue. , 2012, , . | | 3 |
| 62 | Automatic Detection of Pectoral Muscle Using Average Gradient and Shape Based Feature. Journal of Digital Imaging, 2012, 25, 387-399. | 2.9 | 50 |
| 63 | A robust cooperative multi-robot path-planning in noisy environment. , 2010, , . | | 5 |
| 64 | Rotation and translation selective Pareto optimal solution to the box-pushing problem by mobile robots using NSGA-II. , 2009, , . | | 12 |
| 65 | Cooperative multi-robot path planning using differential evolution. Journal of Intelligent and Fuzzy Systems, 2009, 20, 13-27. | 1.4 | 57 |
| 66 | A Multi-Objective Pareto-Optimal Solution to the Box-Pushing Problem by Mobile Robots. , 2008, , . | | 5 |
| 67 | Distributed cooperative multi-robot path planning using differential evolution. , 2008, , . | | 30 |