

APPENDIX for

“Chronic Pain Diagnosis based on Artificial Intelligence: A System Review and Research Vision”

Appendix Table 1. Inclusion and exclusion criteria

| No. | Category of Criteria | Include Criteria | Exclude Criteria |
|-----|-----------------------------|---|---|
| 1 | Types of Pain | Studies focused on the chronic pain | Studies focused on the acute pain, overlapping pain, traumatic brain injury |
| 2 | Task of the Study | Studies focused on the prediction, classification, identification, assessment and diagnosis of chronic pain | Studies focused on the classification and identification of chronic pain-related information rather than the diagnosis of chronic pain, as shown the follows: <ul style="list-style-type: none">• chronic pain medicine usage• chronic pain intensity prediction• complications of chronic pain (e.g., multiple sclerosis, depression, insomnia, brain dysfunction, etc.)• chronic pain prevention• treatments for chronic pain• pain management• pain relief• development of chronic pain• severity of chronic pain• chronic pain related operation• clinical notes identification• sex differences of chronic pain• identification of subgroups of chronic pain patients• diet of the chronic pain patients• pain induced |
| 3 | Machine Learning Techniques | Modern machine learning techniques: Adaptive boosting, Bayesian, decision trees, fuzzy logic, gradient boosting, k-means clustering, natural language processing, nearest neighbors, neural networks, principal component analysis, random forests, reinforcement learning, simulated treatment learning, support vector machines | Not use artificial intelligence approaches, or only use traditional machine learning approaches, such as: Rule-based systems, Logistic and linear regression, Linear mixed-effect model, ANOVA |
| | | | Sample size less than 10 patients |
| | | | Did not report standard performance metrics (e.g., accuracy or AUC) or comparisons with a control group for at least one modern AI model. |

| | | | |
|---|--------------------|--|---------------------------------------|
| 4 | Article Type | Peer-reviewed articles describing: - Original research - Structured reviews of the literature reported in accordance with PRISMA guidelines | Not published in the English language |
| 5 | Species of Samples | Human | Different animals except human |
| 6 | Publication Year | Articles published between 2012 to 2022 (Last ten years) | Articles published before 2012 |

Appendix Note 1. PubMed (NCBI) Search Strategy

("Chronic Pain/classification"[Mesh] OR "Chronic Pain/diagnosis"[Mesh] OR "Chronic Pain/diagnostic imaging"[Mesh] OR Chronic Pain Diagnosis OR Chronic Pain Assessment OR Chronic Pain Classification) AND (Artificial Intelligence[Mesh] OR Artificial Intelligence OR Machine Learning OR Deep Learning OR Support Vector Machine OR Transfer Learning OR Transformer OR Graph Neural Network OR Adaptive boosting OR Bayesian OR Decision Trees OR Fuzzy Logic OR Gradient Boosting OR k-means Clustering OR Natural Language Processing OR Nearest Neighbors OR Neural Networks OR principal component analysis OR random forests OR reinforcement learning OR Simulated Treatment Learning) NOT ((Therapy[Title]) OR (Treatment[Title]) OR (Opioid[Title]) OR (Opiate[Title]) OR (Relief)[Title] OR (Intervention[Title]) OR (Prevention[Title]))

Restricted to ENGLISH LANGUAGE AND SPECIES Human AND PUBLISHED January 1, 2012 to Aug 2, 2022

Appendix Note 2. Web of Science (WOS) Search Strategy

TI=((((chronic pain diagnosis) OR (chronic low back pain diagnosis) OR (chronic pain assessment) OR (chronic pain prediction)) AND ((machine learning) OR (deep learning) OR (classification) OR (Support Vector Machine) OR (Transfer Learning) OR Transformer OR (Graph Neural Network))) OR AB = (((chronic pain diagnosis) OR (chronic low back pain diagnosis) OR (chronic pain assessment) OR (chronic pain prediction)) AND ((machine learning) OR (deep learning) OR (classification) OR (Support Vector Machine) OR (Transfer Learning) OR Transformer OR (Graph Neural Network))) NOT TI = ((therapy) OR (treatment) OR (opioid) OR (opiate) OR (relief) OR (intervention) OR (prevention)) NOT SO=CLINICAL REHABILITATION NOT WC= Rehabilitation AND PY=(2012-2022)

Appendix Note 3. Google Scholar Search Strategy

| No. | Search Query | Search Result No. |
|-----|---|-------------------|
| 1 | (intitle:chronic pain assessment) OR (intitle:chronic pain diagnosis) AND (intitle:deep learning) OR (intitle:machine learning) | 87 |
| 2 | (intitle:pain assessment) OR (intitle:pain diagnosis) AND (intitle:deep learning) OR (intitle:machine learning) | 147 |
| 3 | chronic pain diagnosis AND "deep learning" | 24 |
| 4 | chronic pain diagnosis AND "machine learning" | 65 |

Appendix Note 4. Survey Paper List

| No. | First Author | Year | Title | Database | PMID |
|-----|-----------------|------|--|----------|----------|
| 1 | D'Antoni, F | 2022 | Artificial Intelligence and Computer Aided Diagnosis in Chronic Low Back Pain: A Systematic Review | WoS | 35627508 |
| 2 | D'Antoni, F | 2021 | Artificial Intelligence and Computer Vision in Low Back Pain: A Systematic Review | WoS | 34682647 |
| 3 | Jenssen, MDK | 2021 | Machine Learning in Chronic Pain Research: A Scoping Review | WoS | 33259458 |
| 4 | Boissoneault, J | 2017 | Biomarkers for Musculoskeletal Pain Conditions: Use of Brain Imaging and Machine Learning | WoS | 28144827 |

Appendix Table 2. Excluding Paper and Corresponding Excluding Reasons during Full-text Reading (N = 21)

| No. | First Author | Year | Title | Database | PMID | Exclude Reason |
|-----|--------------|------|--|----------|----------|--|
| 1 | Abdollah V | 2021 | Texture analysis in the classification of T (2) - weighted magnetic resonance images in persons with and without low back pain | PubMed | 33247597 | This paper only uses the random forests algorithm to select the most promising classifiers. But they use the linear mixed-effect model for the low back pain daignosis. |
| 2 | Moustafa S | 2020 | Accurate diagnosis of endometriosis using serum microRNAs | PubMed | 32165186 | According to ICD-11, endometriosis is belong to diseases of the genitourinary system, rather than chronic pain. |
| 3 | Yang Z | 2020 | Combining deep learning with token selection for patient phenotyping from electronic health records | PubMed | 31996705 | This paper aims to identify disease phenotypes from EHR. Chronic pain is one of the 10 diseaese (including Depression, Psychiatric Disorders, Obesity, Substance Abuse, Alcohol Abuse, Chronic Pain, Chronic Neuro, Adv. Lung Disease, Adv. Heart Disease, Adv. Cancer). So this article is not focused on chronic pain diagnosis. |
| 4 | Strik C | 2019 | Risk of Pain and Gastrointestinal Complaints at 6Months After Elective Abdominal Surgery | PubMed | 30107242 | This study aims to assess risk factors instead of diagnosing or identifying chronic postoperative abdominal pain (CPAP). |
| 5 | D'Antoni, F | 2022 | Artificial Intelligence and Computer Aided Diagnosis in Chronic Low Back Pain: A Systematic Review | WoS | 35627508 | Survey paper |
| 6 | Keller, AV | 2022 | Unsupervised Machine Learning on Motion Capture Data Uncovers Movement Strategies in Low Back Pain | WoS | 35497350 | This study presented a biomechanical biomarker that could potentially identify LBP subjects. The performance of this biomaker is unknown as no classification experiments or accuracy is reported. |
| 7 | Nephew, BC | 2022 | Depression Predicts Chronic Pain | WoS | 34908146 | This study focuses on the potential relationship |

| | | | | | | |
|----|-----------------|------|--|----------------|----------|---|
| | | | Interference in Racially Diverse, Income-Disadvantaged Patients | | | between depression and chronic pain, which does not match the topics of chronic pain diagnosis or classification. |
| 8 | D'Antoni, F | 2021 | Artificial Intelligence and Computer Vision in Low Back Pain: A Systematic Review | WoS | 34682647 | Survey paper |
| 9 | Jenssen, MDK | 2021 | Machine Learning in Chronic Pain Research: A Scoping Review | WoS | 33259458 | Survey paper |
| 10 | Brown, TT | 2020 | The FUTUREPAIN study: Validating a questionnaire to predict the probability of having chronic pain 7-10 years into the future | WoS | 32817710 | This study aims to predict the probability of chronic pain in the future 7–10 years, which does not match the topic of diagnosing or identifying chronic pain. And |
| 11 | Boissoneault, J | 2017 | Biomarkers for Musculoskeletal Pain Conditions: Use of Brain Imaging and Machine Learning | WoS | 28144827 | Survey paper |
| 12 | Mauricio, A | 2020 | Chronic Pain Estimation Through Deep Facial Descriptors Analysis | Google Scholar | N.A. | Pain intensity prediction |
| 13 | Grauhan, NF | 2021 | Deep learning for accurately recognizing common causes of shoulder pain on radiographs | Google Scholar | 33611622 | This paper aim to recognize for common causes of shoulder pain (including both acute and chronic causes), not for chronic pain diagnosis. |
| 14 | Guan, B | 2022 | Deep learning approach to predict pain progression in knee osteoarthritis | Google Scholar | 33835240 | This paper aim to predict pain progression (the changes in pain score between baseline and two or more follow-up time over the first 48-months) in knee osteoarthritis, rather than chronic pain diagnosis. |
| 15 | Schmidt, D | 2021 | Deep learning takes the pain out of back breaking work - Automatic vertebral segmentation and attenuation measurement for osteoporosis | Google Scholar | 34598006 | This paper only use the deep learning technique to segment CT image for “vertebrae” identification, which is a part of the knee osteoporosis diagnosis. |
| 16 | Ibrahim, Said A | 2021 | Artificial intelligence for disparities in knee pain assessment | Google Scholar | 33442017 | This paper focus on the exploration of racial disparities in the assessment of knee osteoarthritis, rather than diagnosis of knee osteoarthritis |
| 17 | Lukkahatai N | 2018 | A predictive algorithm to identify genes that discriminate individuals with fibromyalgia syndrome diagnosis | Survey Paper | 30538537 | Machine learning is just part of the methods they used, not for the main diagnosis task |

| | | | | | | |
|----|------------------|------|--|--------------|----------|--|
| | | | from healthy control subjects | | | |
| 18 | Ultsch A | 2016 | A data science approach to candidate gene selection of pain regarded as a process of learning and neural plasticity | Survey Paper | 27548044 | Used machine learning to combine the knowledge to identify the genes relevant to pain, rather than diagnosis. |
| 19 | Lee J | 2019 | Machine learning-based prediction of clinical pain using multimodal neuroimaging and autonomic metrics. | Survey Paper | 30540621 | Predict pain intensity, rather than diagnosis. |
| 20 | B Mathew | 1998 | Artificial intelligence in the diagnosis of low-back pain and sciatica | Survey Paper | 2970122 | This paper is published in 1988 |
| 21 | Masoud Abdollahi | 2020 | Using a Motion Sensor to Categorize Nonspecific Low Back Pain Patients: A Machine Learning Approach | Survey Paper | 32604794 | This paper focuses on categorization of nonspecific low back pain patients to low, medium and high risk categories, rather than chronic pain diagnosis. |
| 22 | J B Bishop | 1997 | Classification of low back pain from dynamic motion characteristics using an artificial neural network | Survey Paper | 9431637 | This paper is published in 1997 |
| 23 | Karabulut, EM | 2014 | Effective automated prediction of vertebral column pathologies based on logistic model tree with SMOTE preprocessing | Survey Paper | 24753003 | This paper focus on the identification of Vertebral Column Pathologies, which may cause acute pain or chronic pain, rather than diagnosis of chronic pain. |
| 24 | N W Sanders | 2000 | Automated scoring of patient pain drawings using artificial neural networks: Efforts toward a low back pain triage application | Survey Paper | 10913774 | This paper is published in 2000 |

Appendix Table 3. Data Abstraction of all Including Papers (N = 55)

| No. | Authors | Year | Title | Database | PMID | Country | Organization | Journal /Conference | DOI |
|-----|--|------|--|---------------|----------|---------|----------------------------|------------------------|---------------------------|
| 1 | Lamichhane B, Jayasekera D, Jakes R, Ray WZ, Leuthardt EC, Hawasli AH. | 2021 | Functional Disruptions of the Brain in Low Back Pain: A Potential Imaging Biomarker of Functional Disability | Survey Papers | 34335444 | USA | Washington University | Frontiers in Neurology | 10.3389/fneur.2021.669076 |
| 2 | Ung H, Brown JE, Johnson | 2014 | Multivariate | PubMed | 23246778 | USA | University of Pennsylvania | Cerebral Cortex | 10.1093/ |

| | | | | | | | | | |
|---|---|------|--|----------------|----------|---------|---------------------------|--|----------------------------|
| | KA, Younger J, Hush J, Mackey S. | | classification of structural MRI data detects chronic low back pain | | | | | | cerc or/bhs378 |
| 3 | Ketola, J. H., Inkinen, S. I., Karppinen, J., Niinimäki, J., Tervonen, O., & Nieminen, M. T. | 2021 | T2-weighted magnetic resonance imaging texture as predictor of low back pain: A texture analysis-based classification pipeline to symptomatic and asymptomatic cases | Survey Papers | 33368707 | Finland | University of Oulu | Journal of Orthopaedic Research | N.A. |
| 4 | Athertya JS, Saravana Kumar G. | 2021 | Classification of certain vertebral degenerations using MRI image features | PubMed | 33984847 | India | IIT - Madras | Biomedical Physics Engineering Express | 10.1088/2057-1976/ac00d2 |
| 5 | Shen W, Tu Y, Gollub RL, Ortiz A, Napadow V, Yu S, Wilson G, Park J, Lang C, Jung M, Gerber J, Mawla I, Chan ST, Wasan AD, Edwards RR, Kaptchuk T, Li S, Rosen B, Kong J. | 2019 | Visual network alterations in brain functional connectivity in chronic low back pain: A resting state functional connectivity and machine learning study | PubMed | 30927604 | China | Hainan Medical University | Neuroimage Clinical | 10.1016/j.nicl.2019.101775 |
| 6 | Kulkarni, K. R., Gaonkar, A., Vijayarajan, V., & Manikandan, K | 2014 | Analysis of lower back pain disorder using deep learning | Google Scholar | N.A. | India | VIT University | IOP Conference Series: Materials Science and Engineering | N.A. |
| 7 | Torrado-Carvajal A, Toschi N, Albrecht DS, Chang K, | 2021 | Thalamic neuroinflammation as a reproduci | PubMed | 33065737 | USA | Harvard Medical School | Pain | 10.1097/j.pain.00000000 |

| | | | | | | | | | |
|----|---|------|---|----------------|--------------|-------|---------------------------------------|----------------------------------|---|
| | Akeju O, Kim M, Edwards RR, Zhang Y, Hooker JM, Duggento A, Kalpathy-Cramer J, Napadow V, Loggia ML. | | ble and discriminating signature for chronic low back pain | | | | | | 0000 0210 8 |
| 8 | Lamichhane B, Jayasekera D, Jakes R, Glasser MF, Zhang J, Yang C, Grimes D, Frank TL, Ray WZ, Leuthardt EC, Hawasli AH. | 2021 | Multi-modal biomarkers of low back pain: A machine learning approach | Survey Papers | 3333 8968 | USA | Washington University | NeuroImage: Clinical | 10.1 016/j .ncl. 2020 .102 530 |
| 9 | Tan, W. K., Hassanpour, S., Heagerty, P. J. | 2018 | Comparison of natural language processing rules-based and machine-learning systems to identify lumbar spine imaging findings related to low back pain | Google Scholar | 2960 5561 | USA | University of Washington | Academic Radiology | N.A. |
| 10 | Owari Y, Miyatake N. | 2019 | Prediction of Chronic Lower Back Pain Using the Hierarchical Neural Network: Comparison with Logistic Regression-A Pilot Study | PubMed | 3118 1815 | Japan | Shikoku Medical College | Medicina | 10.3 390/ medi cina 5506 0259 |
| 11 | Parsaeian M, Mohammad K, Mahmoudi M, Zeraati H | 2012 | Comparison of logistic regression and artificial neural network in low back pain prediction: Second national health survey | Survey Papers | 2311 3198 | Iran | Tehran University of Medical Sciences | Iranian Journal of Public Health | N.A. |

| | | | | | | | | | |
|----|---|------|---|-------------------|--------------|-----------------|--|---|---|
| 12 | Judd, M; Zulkernine, F; Wolfrom, B; Barber, D; Rajaram, A | 2018 | Detecting Low Back Pain from Clinical Narratives Using Machine Learning Approach es | Web of Science | N.A. | Canada | Queen's University, Kingston | International Conference on Database and Expert Systems Applications: Dexa 2018 International Workshops | 10.1 007/ 978- 3- 319- 9913 3- 7_10 |
| 13 | Hu B, Kim C, Ning X, Xu X. | 2018 | Using a deep learning network to recognise low back pain in static standing | PubMed | 2979 2576 | USA | Harvard T.H. Chan School of Public Health | Ergonomics | 10.1 080/ 0014 0139 .201 8.14 8123 0 |
| 14 | Ashouri, S., Abedi, M., Abdollahi, M., Manshadi, F. D., Parnianpour, M., & Khalaf, K. | 2017 | A novel approach to spinal 3D kinematic assessme nt using inertial sensors: Towards effective quantitativ e evaluation of low back pain in clinical settings | Survey Papers | 2880 0443 | Iran | Sharif University of Technology | Computers in Biology and Medicine | N.A. |
| 15 | Thiry, P; Houry, M; Philippe, L; Nocent, O; Buisseret, F; Dierick, F; Slama, R; Bertucci, W; Thevenon, A; Simoneau- Buessinger, E | 2022 | Machine Learning Identifies Chronic Low Back Pain Patients from an Instrumen ted Trunk Bending and Return Test | Web of Science | 3580 8522 | France | Université Polytechnique Hauts-de- France | Sensors (Basel) | 10.3 390/ s221 3502 7 |
| 16 | Chan, H; Zheng, HR; Wang, HY; Sterritt, R; Newell, D | 2013 | Smart Mobile Phone Based Gait Assessme nt of Patients with Low Back Pain | Web of Science | 2608 9700 | UK | University of Ulster | International Conference on Natural Computation | N.A. |
| 17 | Staatjes VE, Quddusi A, Klukowska AM, Schröder ML. | 2020 | Initial classificati on of low back and leg pain | PubMed | 3207 2271 | Switzerl and | University of Zurich | European Spine Journal | 10.1 007/ s005 86- 020- |

| | | | | | | | | | |
|----|--|------|---|----------------|-----------|--------|---|--|-----------------------------------|
| | | | based on objective functional testing: a pilot study of machine learning applied to diagnostics | | | | | | 0634 3-5 |
| 18 | Bernard X W Liew, David Rugamer, Alessandro Marco De Nunzio, Deborah Falla | 2019 | Interpretable machine learning models for classifying low back pain status using functional physiological variables | Survey Papers | 3212 4044 | UK | University of Essex | European Spine Journal | 10.1 007/ s005 86- 020- 0635 6-0 |
| 19 | Du, WJ; Omisore, OM; Li, HH; Ivanov, K; Han, SP; Wang, L | 2018 | Recognition of Chronic Low Back Pain during Lumbar Spine Movements Based on Surface Electromyography Signals | Web of Science | 3281 7710 | China | Shenzhen Institutes of Advanced Technology, Chinese Academy of Sciences | IEEE Access | 10.1 109/ ACC ESS. 2018 .287 7254 |
| 20 | Caza-Szoka M, Massicotte D, Nougrou F, Descarreaux M. | 2016 | Surrogate analysis of fractal dimensions from SEMG sensor array as a predictor of chronic low back pain | PubMed | 2826 9714 | Canada | University of Quebec, Tri-Rivers City | International Conference of The IEEE Engineering in Medicine & Biology Society | 10.1 109/ EMB C.20 16.7 5921 95 |
| 21 | Wang, N., Zhang, Z., Xiao, J., & Cui, L | 2019 | DeepLap: A Deep Learning based Non-Specific Low Back Pain Symptomatic Muscles Recognition System | Google Scholar | N.A. | China | Chinese Academy of Sciences | IEEE International Conference on Sensing, Communication, and Networking | N.A. |

| | | | | | | | | | |
|----|---|------|---|----------------|----------|--------|---|---|-----------------------------|
| 22 | Robinson ME, O'Shea AM, Craggs JG, Price DD, Letzen JE, Staud R. | 2015 | Comparison of machine classification algorithms for fibromyalgia: neuroimages versus self-report | PubMed | 25704840 | USA | University of Florida | Journal of Pain | 10.1016/j.jpain.2015.02.002 |
| 23 | Behr M, Saiel S, Evans V, Kumbhare D. | 2020 | Machine Learning Diagnostic Modeling for Classifying Fibromyalgia Using B-mode Ultrasound Images | PubMed | 32174253 | Canada | University Health Network | Ultrasonic Imaging | 10.1177/0161734620908789 |
| 24 | Alves MVS, Maciel LIL, Ramalho RRF, Lima LAS, Vaz BG, Morais CLM, Passos JOS, Pegado R, Lima KMG. | 2021 | Multivariate classification techniques and mass spectrometry as a tool in the screening of patients with fibromyalgia | PubMed | 34799667 | Brazil | Federal University of Rio Grande do Norte | Scientific Reports | 10.1038/s41598-021-02141-1 |
| 25 | Emir, B; Masters, ET; Mardekian, J; Clair, A; Kuhn, M; Silverman, SL | 2015 | Identification of a potential fibromyalgia diagnosis using random forest modeling applied to electronic medical records | Web of Science | 26089700 | USA | Pfizer Inc. | Journal of Pain Research | 10.2147/JPR.S82566 |
| 26 | Andrés-Rodríguez L, Borràs X, Feliu-Soler A, Pérez-Aranda A, Rozadilla-Sacanell A, Arranz B, Montero-Marin J, García-Campayo J, Angarita- | 2019 | Machine Learning to Understand the Immune-Inflammatory Pathways in Fibromyalgia | PubMed | 31470635 | Spain | Institut de Recerca Sant Joan de Déu | International Journal of Molecular Sciences | 10.3390/ijms20174231 |

| | | | | | | | | | |
|----|--|------|--|----------------|----------|---------|---|--|---------------------------------|
| | Osorio N, Maes M, Luciano JV. | | | | | | | | |
| 27 | Sundermann B, Burgmer M, Pogatzki-Zahn E, Gaubitz M, Stüber C, Wessolleck E, Heuft G, Pfeleiderer B. | 2014 | Diagnostic classification based on functional connectivity in chronic pain: model optimization in fibromyalgia and rheumatoid arthritis | PubMed | 24507423 | Germany | University Hospital Munster, Albert-Schweitzer-Campus | Academic Radiology | 10.1016/j.acra.2013.12.003 |
| 28 | Wang, R., Xu, K., Feng, H., & Chen, W | 2020 | Hybrid RNN-ANN Based Deep Physiological Network for Pain Recognition | Google Scholar | 33019243 | China | Fudan University | International Conference of The IEEE Engineering in Medicine & Biology Society | N.A. |
| 29 | Fodeh SJ, Finch D, Bouayad L, Luther S, Kerns RD, Brandt C. | 2017 | Classifying Clinical Notes with Pain Assessment using Machine Learning | PubMed | 29295346 | USA | Yale University School of Medicine | Medical & Biological Engineering & Computing | N.A. |
| 30 | Gilam, G; Cramer, EM; Webber, KA; Ziadni, MS; Kao, MC; Mackey, SC | 2021 | Classifying chronic pain using multidimensional pain-agnostic symptom assessments and clustering analysis | Web of Science | 34516888 | USA | Stanford University School of Medicine | Science Advances | 10.1126/sciadv.abj0320 |
| 31 | Gaynor, SM; Bortsov, A; Bair, E; Fillingim, RB; Greenspan, JD; Ohrbach, R; Diatchenko, L; Nackley, A; Tchivileva, IE; Whitehead, W; Alonso, AA; Buchheit, TE; Boortz-Marx, RL; Liedtke, W; Park, JJ; Maixner, W; Smith, SB | 2021 | Phenotypic profile clustering pragmatically identifies diagnostically and mechanistically informative subgroups of chronic pain patients | Web of Science | 33259458 | USA | Harvard T.H. Chan School of Public Health | Pain | 10.1097/j.pain.0000000000002153 |

| | | | | | | | | | |
|----|---|------|---|----------------|----------|-------|---|--|--------------------------------|
| 32 | Antonucci, LA; Taurino, A; Laera, D; Taurisano, P; Losole, J; Lutricuso, S; Abbatantuono, C; Giglio, M; De Caro, MF; Varrassi, G; Puntillo, F | 2020 | An Ensemble of Psychological and Physical Health Indices Discriminates Between Individuals with Chronic Pain and Healthy Controls with High Reliability: A Machine Learning Study | Web of Science | 32880867 | Italy | University of Bari Aldo | Pain and Therapy | 10.1007/s40122-020-00191-3 |
| 33 | Zhang W, Bianchi J, Turkestani NA, Le C, Deleat-Besson R, Ruellas A, Cevdanes L, Yatabe M, Goncalves J, Benavides E, Soki F, Prieto J, Paniagua B, Najarian K, Gryak J, Soroushmehr R. | 2021 | Temporomandibular Joint Osteoarthritis Diagnosis Using Privileged Learning of Protein Markers | PubMed | 34891638 | USA | University of Michigan, Ann Arbor | International Conference of The IEEE Engineering in Medicine & Biology Society | 10.1109/EMBC46164.2021.9629990 |
| 34 | Bianchi J, de Oliveira Ruellas AC, Gonçalves JR, Paniagua B, Prieto JC, Styner M, Li T, Zhu H, Sugai J, Giannobile W, Benavides E, Soki F, Yatabe M, Ashman L, Walker D, Soroushmehr R, Najarian K, Cevdanes LHS. | 2020 | Osteoarthritis of the Temporomandibular Joint can be diagnosed earlier using biomarkers and machine learning | PubMed | 32415284 | USA | School of Dentistry | Scientific Reports | 10.1038/s41598-020-64942-0 |
| 35 | Mao CP, Chen FR, Huo JH, Zhang L, Zhang GR, Zhang B, Zhou XQ. | 2020 | Altered resting-state functional connectivity and effective connectivity of the | PubMed | 32488929 | China | Second Affiliated Hospital of Xi'an Jiaotong University | Human Brain Mapping | 10.1002/hbm.25038 |

| | | | | | | | | | |
|----|--|------|---|---------------|----------|--------|---|---|---------------------------------|
| | | | habenula in irritable bowel syndrome: A cross-sectional and machine learning study | | | | | | |
| 36 | Labus JS, Van Horn JD, Gupta A, Alaverdyan M, Torgerson C, Ashe-McNalley C, Irimia A, Hong JY, Naliboff B, Tillisch K, Mayer EA. | 2015 | Multivariate morphological brain signatures predict chronic abdominal pain patients from healthy control subjects. | Survey Papers | 25906347 | USA | University of California at Los Angeles | Pain | 10.1097/j.pain.0000000000000196 |
| 37 | Lin YC, Yu NY, Jiang CF, Chang SH. | 2018 | Characterizing the SEMG patterns with myofascial pain using a multi-scale wavelet model through machine learning approaches | Survey Papers | 29890503 | China | National Cheng Kung University | Journal of Electromyography and Kinesiology | 10.1016/j.jelekin.2018.05.004 |
| 38 | Behr M, Noseworthy M, Kumbhare D. | 2019 | Feasibility of a Support Vector Machine Classifier for Myofascial Pain Syndrome : Diagnostic Case-Control Study | PubMed | 30614553 | Canada | University of Toronto | Journal Of Ultrasound in Medicine | 10.1002/jum.14909 |
| 39 | Callan D, Mills L, Nott C, England R, England S. | 2014 | A tool for classifying individuals with chronic back pain: using multivariate pattern analysis | PubMed | 24905072 | Japan | Osaka University | PLoS One | 10.1371/journal.pone.0098007 |

| | | | | | | | | | |
|----|---|------|---|----------------|----------|-------|--|---------------------------|------|
| | | | with functional magnetic resonance imaging data | | | | | | |
| 40 | Santana, A. N., Cifre, I., De Santana, C. N. | 2019 | Using Deep Learning and Resting-State fMRI to Classify Chronic Pain Conditions | Google Scholar | 31920483 | Spain | University of the Balearic Islands | Frontiers In Neuroscience | N.A. |
| 41 | Santana, A. N., de Santana, C. N., & Montoya, P | 2020 | Chronic Pain Diagnosis Using Machine Learning, Questionnaires, and QST: A Sensitivity Experiment | Google Scholar | 33212774 | Spain | University of the Balearic Islands | Diagnostics (Basel) | N.A. |
| 42 | Tan, W. K., & Heagerty, P. J. (2022) | 2020 | Surrogate-guided sampling designs for classification of rare outcomes from electronic medical records data | Google Scholar | N.A. | USA | University of Washington | Biostatistics | N.A. |
| 43 | Lee, J. J., Liu, F., Majumdar, S., & Pedoia, V | 2021 | An ensemble clinical and MR-image deep learning model predicts 8-year knee pain trajectory: Data from the osteoarthritis initiative | Google Scholar | N.A. | USA | University of California San Francisco | Osteoarthritis Imaging | N.A. |
| 44 | Chang, G. H., Felson, D. T., Qiu, S., Capellini, T. D | 2018 | Assessment of bilateral knee pain from MR imaging | Google Scholar | N.A. | USA | Boston University | Biorxiv | N.A. |

| | | | | | | | | | |
|----|---|------|---|----------------|----------|----------|---|--|---------------------------------|
| | | | using deep neural networks | | | | | | |
| 45 | Barroso J, Vigotsky AD, Branco P, Reis AM, Schnitzer TJ, Galhardo V, Apkarian AV. | 2020 | Brain gray matter abnormalities in osteoarthritis pain: a cross-sectional evaluation | PubMed | 32379222 | Portugal | Universidade do Porto | Pain | 10.1097/j.pain.0000000000001904 |
| 46 | Wang Y, Zhu Y, Xue Q, Ji M, Tong J, Yang JJ, Zhou CM. | 2021 | Predicting chronic pain in postoperative breast cancer patients with multiple machine learning and deep learning models | PubMed | 34364190 | China | The First Affiliated Hospital of Zhengzhou University | Journal Of Clinical Anaesthesia | 10.1016/j.jclinane.2021.110423 |
| 47 | Kartal, E; Kocoglu, FO; Ozen, Z; Emre, IE; Gungor, G; Bozkurt, PS | 2022 | AN INTELLIGENT POSTOPERATIVE CHRONIC PAIN PREDICTION SYSTEM (I-POCPP) | Web of Science | N.A. | Turkiye | Istanbul University | Journal of Istanbul Faculty of Medicine-Istanbul Tıp Fakultesi Dergisi | 10.26650/IUITFD.972738 |
| 48 | He M, Wang X, Zhao Y. | 2021 | A calibrated deep learning ensemble for abnormality detection in musculoskeletal radiographs | PubMed | 33907257 | USA | Fordham University | Scientific Reports | 10.1038/s41598-021-88578-w |
| 49 | Lendaro E, Balouji E, Baca K, Muhammad AS, Ortiz-Catalan M. | 2021 | Common Spatial Pattern EEG decomposition for Phantom Limb Pain detection | PubMed | 34891394 | Sweden | Chalmers University of Technology | International Conference of The IEEE Engineering in Medicine & Biology Society | 10.1109/EMBC46164.2021.9630561 |
| 50 | Yang M, Zheng H, Wang H, McClean S, | 2012 | A machine learning approach | Survey Papers | 21996355 | UK | University of Ulster | Medical Engineering & Physics | 10.1016/j.medengp |

| | | | | | | | | | |
|----|---|------|--|--------|----------|--------|------------------------------------|---|------------------------------------|
| | Hall J, Harris N. | | to assessing gait patterns for Complex Regional Pain Syndrome | | | | | | hy.2011.09.018 |
| 51 | Bagarinao E, Johnson KA, Martucci KT, Ichesco E, Farmer MA, Labus J, Ness TJ, Harris R, Deutsch G, Apkarian VA, Mayer EA, Clauw DJ, Mackey S. | 2014 | Preliminary structural MRI based brain classification of chronic pelvic pain: A MAPP network study | PubMed | 25242566 | USA | Stanford University Medical Center | Pain | 10.1016/j.pain.2014.09.002 |
| 52 | Cheng Y, Jin Z, Zhou X, Zhang W, Zhao D, Tao C, Yuan J. | 2022 | Diagnosis of Metacarpophalangeal Synovitis with Musculoskeletal Ultrasound Images | PubMed | 34930637 | China | Nanjing University | Ultrasound In Medicine & Biology | 10.1016/j.ultrasmedbio.2021.11.003 |
| 53 | Jiménez-Grande D, Farokh Atashzar S, Martinez-Valdes E, Marco De Nunzio A, Falla D. | 2021 | Kinematic biomarkers of chronic neck pain measured during gait: A data-driven classification approach | PubMed | 33581443 | UK | University of Birmingham | Journal Of Biomechanics | 10.1016/j.jbiomech.2020.110190 |
| 54 | Zhong J, Chen DQ, Hung PS, Hayes DJ, Liang KE, Davis KD, Hodaie M. | 2018 | Multivariate pattern classification of brain white matter connectivity predicts classic trigeminal neuralgia | PubMed | 29905649 | Canada | University Health Network | Pain | 10.1097/j.pain.0000000000001312 |
| 55 | Garcia-Chimeno Y, Garcia-Zapirain B, Gomez-Beldarrain M, Fernandez- | 2017 | Automatic migraine classification via feature selection committee | PubMed | 28407777 | Spain | Avda. Universidades | BMC Medical Informatics and Decision Making | 10.1186/s12911-017-0434-4 |

| | | | | | | | | | |
|--|-----------------------------------|--|---|--|--|--|--|--|--|
| | Ruanova B, Garcia-Monco JC. | | and machine learning technique s over imaging and questionn aire data | | | | | | |
|--|-----------------------------------|--|---|--|--|--|--|--|--|

Appendix Table 4. List of Abbreviation

| No. | Abbreviation | Definition |
|--------------------------------|--------------|--|
| Data Types | | |
| 1 | ECG | Electrocardiography |
| 2 | EMG | Electromyography |
| 3 | sEMG | Surface electromyography |
| 4 | EEG | Electroencephalo-graph |
| 5 | EMR | Electronic medical record |
| 6 | SC | Skin conductance |
| 7 | fMRI | Functional magnetic resonance imaging |
| 8 | MRI | Magnetic resonance images |
| 9 | QSTs | Quantitative sensory tests |
| 10 | BDI | Beck depression inventory II |
| 11 | STAI | State–trait anxiety inventory |
| 12 | GSRs | Gastrointestinal symptom rating scale |
| 13 | IMGV | Integrative medical group visits |
| 14 | TSPO | 18 kda translocator protein |
| 15 | CT | Computerised tomography |
| Machine Learning Models | | |
| 16 | CART | Classification and regression tree |
| 17 | DT | Decision tree |
| 18 | ETC | Extra trees classifier |
| 19 | KNN | K-nearest neighbors |
| 20 | KRVFL | Kernel-based random vector functional link network |
| 21 | LASSO | Least absolute shrinkage and selection operator algorithm |
| 22 | LDA | Linear discriminant analysis |
| 23 | LR | Logistic regression |
| 24 | LUPI | Learning using privileged information |
| 25 | MLP | Multi-layer perceptron |
| 26 | MPC | Multivariate pattern classification |
| 27 | NB | Naive bayes |
| 28 | PCA | Principle component analysis |
| 29 | RF | Random forest |
| 30 | SVC | Support vector classifier |
| 31 | SVM | Support vector machine |
| 32 | MVPA | Multivariate pattern analysis |
| 33 | MR | Multiple regression |
| 34 | AHC | Agglomerative hierarchical clustering |
| 35 | sPLS-DA | Sparse partial least squares for discrimination analysis |
| 36 | HRGAN | High-resolution Generative Adversarial Network |
| 37 | DB | Data boosting |
| 38 | SPA-LDA | Successive projections algorithm with linear discriminant analysis |
| 39 | NCA-K-NN | Neighbourhood component analysis - k-nearest neighbour |
| Others | | |
| 40 | CP | Chronic pain group |
| 41 | HC | Healthy control group |
| 42 | OA | Osteoarthritis |
| 43 | TMJ | Temporomandibular joint |
| 44 | RA | Rheumatoid arthritis |
| 45 | POCP | Postoperative chronic pain |
| 46 | CPP | Chronic pelvic pain |
| 47 | FMS | Fibromyalgia |

| | | |
|----|--------|--------------------------------|
| 48 | PPT | Pain threshold |
| 49 | LF | Lateral flexion |
| 50 | 5R-STs | Five-repetition sit-to-stand |
| 51 | IBS | Irritable bowel syndrome |
| 52 | MPS | Myofascial pain syndrome |
| 53 | CPSP | Chronic postsurgical pain |
| 54 | MSDs | Musculoskeletal disorders |
| 55 | PLP | Phantom limb pain |
| 56 | CRPS | Complex regional pain syndrome |
| 57 | CPP | Chronic pelvic pain |
| 58 | MCP | Metacarpophalangeal |
| 59 | TN | Trigeminal neuralgia |
| 60 | CNP | Chronic neck pain |