



18022 IVVES

ITEA3

Project details

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Name: Synthetic 3D brain data

Input(s):	Main feature(s)	Output(s):	
Real 3D bra scansMRI, CT,	n • Data preprocessing, registration, augmentation • Multi-GPU GAN training	 Synthetic 3D brain data 	
Unique Selling Proposition(s):	 Full pipeline from data preprocessing t generation Designed to work with moderately low of input data 	·	
Integration constraint(s):	 Implemented as Docker images, which Singularity images Multiple GPUs and lots of RAM (>100 		
Intended user(s):	 Researchers, data scientists 	 Researchers, data scientists 	
Provider:	• VTT	• VTT	
Contact point:	 Harri Pölönen, <u>harri.polonen@vtt.fi</u> Mark.Pijnenburg@philips.com 		
Condition(s) for reuse:	 Mostly built on open-source technology Component licenses: Apace 2.0 and M 		
		Latest update: 29 Nov 2022	





Name:	RELOAD

Input(s):	Main feature(s)	Output(s):
The URL of Web-based SUT + the s jmx files recorded by JMeter representing initial load to plans for ear functional us case of SUT	st h	■ Load test report ■
Unique Selling Proposition(s):	 Automated performance and load testin reinforcement learning 	g of a system using
Integration constraint(s):	■ Can be provided as SaaS	
Intended user(s):	Software testersProduct managersSoftware engineers	
Provider:	RISE Research Institutes of Sweden	
Contact point:	 Mahshid Helali Moghadam <u>Mahshid.helali.moghadam@ri.se</u> Mehrdad Saadatmand <u>mehrdad.saadatmand@ri.se</u> 	
Condition(s) for reuse:	Licensing and permission required	
		Latest update: Nov 2022



Name:	DeepAD
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Input(s):	Main feature(s)	Output(s):
 Sensors' daren data from ot sources in timeseries format 	Detection tool	 Discovers patterns in data that do not conform to the expected normal behaviour
Unique Selling Proposition(s):	 Condition monitoring Predictive maintenance Abnormal behavior detection 	
Integration constraint(s):	 Input data in a sequence format with time sample 	
Intended user(s):	Software developers and designers of industrial application need to implement Anomaly Detection techniques. It has applications in cyber-security intrusion detection, fraud detection, fault detection, system health monitoring, event detection in sensor networks, detecting ecosystem disturbances, defect detection, and medical diagnosis.	
Provider:	■ RISE Research Institutes of Sweden	
Contact point:	 Sima Sinaei <u>sima.sinaei@ri.se</u> Mehrdad Saadatmand <u>mehrdad.saadatmand@ri.se</u> 	
Condition(s) for reuse:	 Licensing and permission required 	

Latest update: Nov 2022



Name: Generatin	ı Adversarial	Examples
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Input(s):		Main feature(s)	Output(s):
■ Image datasets		Enhancing security and robustness of Neural Networks especially in the face of an adversary who wishes to fool the model A slightly perturbed image, still easily recognizable by huma observers with the good of producing a wrong output from the correct target class	
Unique Selling Proposition(s):	adversarial attacks and defence in Machine Learning applications		
Integration constraint(s):	A labeled dataset and a primary machine learning model for classification are needed. The security and Robustness of this Neural Network can be improved by generating an adversarial dataset		
Intended user(s):		Al-based system's developer	
Provider:	RISE Research Institutes of Sweden		
Contact point:	 Sima Sinaei <u>sima.sinaei@ri.se</u> Mehrdad Saadatmand <u>mehrdad.saadatmand@ri.se</u> 		
Condition(s) for reuse:	•	Licensing and permission required	
			Latest update: Nov 2022





Input(s):	Main feature(s)	Output(s):
 Tabular data (incl. synthe) Text data Images Audio files 		■ EDA report ■ Training data
Unique Selling Proposition(s):	 Automated EDA and preprocessing Easy report generation and sharing Educational component of the solution 	
Integration constraint(s):	App is deployed on Streamlit cloud andAdding new features in app requires cod	•
Intended user(s):	Students and ML practitioners	
Provider:	■ Sogeti NL	
Contact point:	■ Tijana Nikolic – <u>tia.nikolic@sogeti.com</u> ■ Almira Pillay – <u>almira.pillay@sogeti.com</u>	
Condition(s) for reuse:	 MIT license 	
		Latest update: Nov 2022





ITEA3

Input(s):	Main feature(s)	Output(s):	
 Code repository Code Complexity prediction Explainable Al layer 		Code complexityExplainable Al artefacts	
Unique Selling Proposition(s):	 Improve peer review process and code maintainability Shift-left 		
Integration constraint(s):		Code repository needs to be in a standardized format, tagged with bugs and the pull requests need to follow a naming convention	
Intended user(s):	 Developers 		
Provider:	■ Sogeti NL		
Contact point:	•	Tijana Nikolic – <u>tia.nikolic@sogeti.com</u> Almira Pillay – <u>almira.pillay@sogeti.com</u>	
Condition(s) for reuse:	Commercial product		
		Latest update: Nov 2022	



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Input(s):		Main feature(s)	Output(s):	
 XML files 		ETL pipelineDashboard	 Visualization 	
Unique Selling Proposition(s):	•	Improve move towards shift-left	dashboard	
Integration constraint(s):		AzureGrafana		
Intended user(s):	-	- '		
Provider:	-	Sogeti NL		
Contact point:	•	Tijana Mikono <u>da:mkono@oogoti.com</u>		
Condition(s) for reuse:	•	Commercial product		
			Latest update: Nov 2022	





	Name:	Quality	/ AI	Framework
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Input(s):	Main feature(s)	Output(s):
■ Al Use Case	 A framework to help organize Al projects Education of practitioners on quality and ethical Al aspects 	 A set of tools and methods to ensure quality and ethical Al development
Unique Selling Proposition(s):	 Ethical AI considerations Quality AI considerations Education of practitioners 	
Integration constraint(s):	 None from a technical perspective, this is a framework Agile way of work 	
Intended user(s):	 Product owners and project managers ML Practitioners Students 	
Provider:	Sogeti NL	
Contact point:	 Tijana Nikolic – <u>tia.nikolic@sogeti.com</u> Almira Pillay – <u>almira.pillay@sogeti.com</u> 	
Condition(s) for reuse:	■ Free to reuse, it is a framework	
		Latest update: Nov 2022





Name: Fa	ilure	anal	vsis
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Input(s):		Main feature(s)	Output(s):	
xunit test result files		 Tests failure analysis provides fast and reliable way to find and group similar failures in your CI pipeline 	 Find similar errors between tests 	
Unique Selling Proposition(s):	•	When failure grouping and similarity scoring a machine, it will free resources from dever fix the most important failures in their CI/C work for a human to download, open and rand analyze which failures belong to the standard package solves this problem by pand failures found within by calculating the using cosine similarity.	lopment team member to D pipeline. It is tedious ead all the test failures ame group. The failure- rocessing xunit xml files	
Integration constraint(s):	 Python 3.8 or never numpy 1.23.4 or never pandas 1.5.1 or never sklearn 1.1 or never lxml 4.9.1 or never drain3 0.9.11 or never 			
Intended user(s):	 Software development team members, like developers people. Teams with CI pipeline with decent amount of automation result. 		•	
Provider:	•	Open Source Code: https://github.com/F-S	Secure/failures-analysis	
Contact point:	•	https://github.com/F-Secure/failures-analys	sis/issues	
Condition(s) for reuse:	■ Apache License 2.0			
			Latest update: Nov 2022	





Input(s):	Main feature(s)	Output(s):
xunit test restilescsv file	ult Visualize tests which state changes most often.	 Calculation of test names which state changes most often Heatmap
Unique Selling Proposition(s):	· ·	
Integration constraint(s):	 Python 3.7 or never pandas junitparser seaborn matplotlib 	
Intended user(s):	 Software development team members, like developers and QA people. Teams with CI pipeline with decent amount of test automation result. 	
Provider:	 Open Source Code: https://github.com/F-Secure/flaky-tests- detection 	
Contact point:	 https://github.com/F-Secure/flaky-tests-detection/issues 	
Condition(s) for reuse:	■ Apache License 2.0	
		Latest update: November 23, 2022





Name.	Annotated	software	requirement	cornus
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Input(s):		Main feature(s) Output(s):	
N/A		Software and system requirement descriptions annotated in universal dependencies	
Unique Selling Proposition(s):	English software requirement text corpus annotated with universal dependencies syntactic and part-of-speech information. Requirements are taken from a variety of domains contained in the open source PURE corpus (accessible at http://nlreqdataset.isti.cnr.it/).		ation. Requirements are
Integration constraint(s):	None		
Intended user(s):	Machine learning or natural language processing experts that require training or evaluation data for automatic analysis of software requirements in universal dependencies grammar as a standalone task or as part of a multiobjective task. This corpus can help improve natural language understanding tasks aiming to interpret, validate or analyse software requirements in a requirement validation or verification scenario.		
Provider:	 Computer Research Institute of Montréal Data available: https://github.com/UniversalDependencies/UD_English-CTeTex/ 		/UD_English-CTeTex/
Contact point:	Pierre André Ménard - pierre-andre.menard@crim.ca		rd@crim.ca
Condition(s) for reuse:	Open source licence <u>CC BY-SA 4.0</u>		
		Lates	t update: November 22, 2022



Name:	TESTAR
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Input(s):	Main feature(s)	Output(s):	
 An application with a graph user interfact (GUI) 	cal generation on GUI level	 Reports on test sequences leading to failures 	
Unique Selling Proposition(s):	 Open source tool for automated robustness testing on GUI level Run-time (scriptless) test generation 		
Integration constraint(s):	 The system under test has to be one of Windows GUI application Web application Mobile (Android or iOS) applica 	Ç	
Intended user(s):	 Software developers of applications wit 	h a GUI	
Provider:	• OUNL	OUNL	
Contact point:	■ Tanja.Vos@ou.nl	Tanja.Vos@ou.nl	
Condition(s) for reuse:	■ BSD-3 open source license		
	La	test update: November 15, 2022	



Name: Conformal Prediction Framework

Input(s):	Main feature(s)	Output(s):
 Regression datasets 	 Machine Learning framework for constructing predictive models that can estimate the confidence of their own predictions Used on top of machine learning models, as built-in quality assurance Good for safety-critical systems 	 Model predictions in the form of confidence intervals
Unique Selling Proposition(s):	 Small memory and energy footprint Works on top of any machine learning model of Ekkono's SD Aids towards validation of evolving systems Confidence bounds are guaranteed to contain the correct output value with some pre-defined probability Necessary for sensitive and high-risk applications 	
Integration constraint(s):	 The framework is part of the Ekkono SDK Modeling: .NET 2.0+ or Python 3.7+ (Windows, macOS, or Linux) Deployment: C++17 (or C++14 with included MPark.Variant), delivered as source to be compile the customer Integration will be done through Ekkono's C++ AP the compiled library 	
Intended user(s):	 Data Scientists, machine learning engineers, and software developers that want to run machine learning and conforma prediction on any type of device 	
Provider:	■ Ekkono Solutions	
Contact point:	Rikard@ekkono.aiEva@ekkono.ai	
Condition(s) for reuse:	Commercial license	
	Latest	update: November 28, 2022





Name: 0	Change-	Analvzer
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Input(s):	Main feature(s)	Output(s):	
 Software Application with GUI (Windows / Web) 	 Software Application exploration using Reinforcement Learning Automated Test Case generation done while exploring the Software Application Change Detection done by comparing collected data from different Software versions 	 Generated Test Cases Change Detection report comparing two different Software versions 	
Unique Selling Proposition(s):	 Open-source framework which explores Software Applications using Reinforcement Learning and uses the collected data to generate Test Cases and to detect changes across different Software versions 		
Integration constraint(s):	 Python 3.7 or newer Pandas Matplotlib LudwigAl OpenAl PyTorch 		
Intended user(s):	■ Software development team members, QA & DEV		
Provider:	Open Source Code: https://github.com/F-Secure/change-analyzer		
Contact point:	Sorin Patrasoiu, <u>sorin.patrasoiu@witr</u>	Sorin Patrasoiu, sorin.patrasoiu@withsecure.com	
Condition(s) for reuse:	■ Apache License 2.0		
	Latest	update: November 23, 2022	



Name: Anomaly Detector

Input(s):	Main feature(s)	Output(s):
 Regression datasets 	 Multi-variate algorithm that detects sudden or instantaneous deviations from a normal state Monitors the health or performance of a device 	 Anomaly score between 0 and 1, indicating how likely (1) or unlikely (0) the observed data point is an anomaly
Unique Selling Proposition(s):	 Unsupervised algorithm, no need for labeled data Algorithm is trained on normal state data, then used in unseen data to detect deviations Applicable to streaming and offline datasets 	
Integration constraint(s):	 The anomaly detector is part of the Ekkono SDK Modeling: .NET 2.0+ or Python 3.7+ (Windows, macOS, or Linux) Deployment: C++17 (or C++14 with included MPark.Variant), delivered as source to be compiled by the customer Integration will be done through Ekkono's C++ API of the compiled library 	
Intended user(s):	 Data Scientists, machine learning engineers, and software developers that want to run machine learning and anomaly detection on any type of device 	
Provider:	■ Ekkono Solutions	
Contact point:	Rikard@ekkono.aiEva@ekkono.ai	
Condition(s) for reuse:	 Commercial license 	
	Latest	update: November 28, 2022