

Exploitable Results by Third Parties

SAMUEL 17010

Project details

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Name: 3DPartFinder Geometric Search with Mesh 3D Models support

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> Mesh based 3D Models (e.g. STL) 	<ul style="list-style-type: none"> Extraction 3DPartFinder-type geometric descriptor 	<ul style="list-style-type: none"> Geometric Search Results Geometric database analytics report (HTML)
Unique Selling Proposition(s):	<ul style="list-style-type: none"> Extraction of fully-featured 3DPartFinder geometric descriptor of mesh-based files (STL, MAGICS, MATAMX). Usable in conjunction with CAD models. Cross-compatible. Corner stone of part design reuse. Basis for inventory control and duplicate parts reduction. 	
Integration constraint(s):	<ul style="list-style-type: none"> Up-to-date Windows installation: Windows 10 and up 64-bit Quad Core Processor 8 GB RAM Memory (16 GB recommended for 3d descriptors extraction) 	
Intended user(s):	<ul style="list-style-type: none"> Any Users in needs of Geometric search to find similar 3D models or information related or associated to the 3D models Software Editors seeking Geometric search capabilities 	
Provider:	<ul style="list-style-type: none"> 3DSemantix 	
Contact point:	<ul style="list-style-type: none"> Alain Coulombe - alain.coulombe@3dsemantix.com 	
Condition(s) for reuse:	<ul style="list-style-type: none"> Subscriptions OEM Contract 	

Latest update: 2022-09-23

Name: AM Partner Search Platform

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> ▪ 3D Models (STL, STEP, Native CAD) ▪ Production data in excel format or database plugin (Material, equipment used, quantity produced) 	<ul style="list-style-type: none"> ▪ Find AM manufacturers based on production experience of targeted part to be 3D printed. ▪ Consult manufacturer's profile and have "at-glance" look at their production distribution (material, lot size, AM Processes.) 	<ul style="list-style-type: none"> ▪ HTML GUI (Online platform)
Unique Selling Proposition(s):	<ul style="list-style-type: none"> ▪ Locate Part Manufacturer based on production experience vs target part ▪ Customized analysis according to reference 3D Model to produce and required characteristics (material, build process). 	
Integration constraint(s):	<ul style="list-style-type: none"> ▪ For users using the platform: <ul style="list-style-type: none"> ○ Access to internet ▪ For data-submitting manufacturer: <ul style="list-style-type: none"> ○ See requirements for <i>3DPartFinder AM Data Valorisation Solution</i> ○ Access to internet 	
Intended user(s):	<ul style="list-style-type: none"> ▪ OEM ▪ AM Manufacturers ▪ Online platform publishers seeking Experience-based Partner Search integration 	
Provider:	<ul style="list-style-type: none"> ▪ 3DSemantix 	
Contact point:	<ul style="list-style-type: none"> ▪ Alain Coulombe - alain.coulombe@3dsemantix.com 	
Condition(s) for reuse:	<ul style="list-style-type: none"> ▪ Subscription 	

Latest update: 2022-09-23

Name: 3DPartFinder AM Data Valorisation Solution

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> ▪ 3D Models (STL, STEP, Native CAD, Native Magics formats) ▪ Production data in excel format or database plugin (Material, equipment used, quantity produced) 	<ul style="list-style-type: none"> ▪ Geometry-based search engine. ▪ Presentation of related production data. ▪ Visualization of geometric search results 3d models directly in Materialise Magics or Third-party viewer. ▪ AI-based Build Time estimation. ▪ AI-based Build Orientation suggestions. 	<ul style="list-style-type: none"> ▪ Desktop GUI ▪ Exportable search results (CSV) ▪ Geometric database analytics report (HTML)
Unique Selling Proposition(s):	<ul style="list-style-type: none"> ▪ Valorisation of the company historical production data. Know-how reuse. ▪ Quick, efficient and cross-compatible geometric search ▪ Target specific data subset, specific to a use case or context (costing, process planning, etc...) ▪ Generate relevant and precise build time estimation for part or build platform. Returned from AI model trained on user's company's production data. ▪ Receive useful build orientation suggestions. Orientations calculated by AI model trained on user's company's production data. 	
Integration constraint(s):	<ul style="list-style-type: none"> ▪ Up-to-date Windows installation: Windows 10 and up ▪ 64-bit Quad Core Processor ▪ 8 GB RAM Memory (16 GB recommended for 3d descriptors extraction) 	
Intended user(s):	<ul style="list-style-type: none"> ▪ Part Designers ▪ RFQ Response Team ▪ AM Production Team 	
Provider:	<ul style="list-style-type: none"> ▪ 3DSemantix 	
Contact point:	<ul style="list-style-type: none"> ▪ Alain Coulombe - alain.coulombe@3dsemantix.com 	
Condition(s) for reuse:	<ul style="list-style-type: none"> ▪ Subscription 	

Latest update: 2022-09-23

Name: AI module to predict AM Build time

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> 3D Models (STL, STEP, Native CAD) Production data or database plugin) 	<ul style="list-style-type: none"> Train AI model on a set of data to predict Build time of a 3D model. 	<ul style="list-style-type: none"> Build time estimate
Unique Selling Proposition(s):	<ul style="list-style-type: none"> Predicts precise AM build time of individual parts or whole build platforms. AI model trained offline on company's own production data and 3D models geometric characteristics and geometric descriptor. Online, global AI model also available. 	
Integration constraint(s):	<ul style="list-style-type: none"> Organized, quality AM production data 	
Intended user(s):	<ul style="list-style-type: none"> AM Manufacturers Software Editors 	
Provider:	<ul style="list-style-type: none"> 3DSemantix 	
Contact point:	<ul style="list-style-type: none"> Alain Coulombe - alain.coulombe@3dsemantix.com 	
Condition(s) for reuse:	<ul style="list-style-type: none"> Subscription OEM Contract 	

Latest update: 2022-09-23

Name: AI module to predict AM Part orientation

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> 3D Models (STL, STEP, Native CAD) Production data or database plugin) 	<ul style="list-style-type: none"> Train AI model on a set of data to predict AM Part orientation of a 3D model. 	<ul style="list-style-type: none"> 3D model suggested orientation
Unique Selling Proposition(s):	<ul style="list-style-type: none"> Predicts part build orientation, taking into account part geometry, material, required supports, build process and more. Suggests various relevant possible orientations when appropriate. AI model trained offline on company's own production data and 3D models geometric characteristics and geometric descriptor. Online, global AI model also available. 	
Integration constraint(s):	<ul style="list-style-type: none"> Organized, quality AM production data 	
Intended user(s):	<ul style="list-style-type: none"> AM Manufacturers Software Editors 	
Provider:	<ul style="list-style-type: none"> 3DSemantix 	
Contact point:	<ul style="list-style-type: none"> Alain Coulombe - alain.coulombe@3dsemantix.com 	
Condition(s) for reuse:	<ul style="list-style-type: none"> Subscription OEM Contract 	

Latest update: 2022-09-23

Name: Automatic build preparation

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> ▪ 3D Models (STL, STEP, Native CAD) ▪ Production data or database plugin 	<ul style="list-style-type: none"> ▪ Train AI model on a set of data to propose build preparatory steps. 	<ul style="list-style-type: none"> ▪ Build parameters: orientation and toolpath optimisation
Unique Selling Proposition(s):	<ul style="list-style-type: none"> ▪ Determining the build parameters is currently a process that often involves a few iterations (trial and error). The number of iterations can be greatly reduced. This reduces cost as well as waste. 	
Integration constraint(s):	<ul style="list-style-type: none"> ▪ The resulting proposition will be integrated into the current Build Processor suite of Materialise (BP). 	
Intended user(s):	<ul style="list-style-type: none"> ▪ AM users and service bureaus ▪ Existing (software) customers 	
Provider:	<ul style="list-style-type: none"> ▪ Materialise 	
Contact point:	<ul style="list-style-type: none"> ▪ Michel Janssens – michel.janssens@materialise.be 	
Condition(s) for reuse:	<ul style="list-style-type: none"> ▪ different business models can be applied: license, pay-per-use... ▪ OEM Contract 	
	<i>Latest update: 2022-09-27</i>	

Name: Build time estimation (BTE) for (high-end) FDM		
Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> ▪ 3D Models (STL, STEP, Native CAD) ▪ Production data or database plugin 	<ul style="list-style-type: none"> ▪ Train AI model on a set of data to predict the FDM build time of a 3D model. 	<ul style="list-style-type: none"> ▪ The build time
Unique Selling Proposition(s):	<ul style="list-style-type: none"> ▪ The current (before the project) accuracy of BTE for FDM had 65% predictions within tolerance. Early tests show that 88% is feasible with a trained AI model. ▪ Two variants can be provided: a pretrained model on a large database present at Materialise (the basic user model) and tools to train models on the customers database (the advanced user model) ▪ The system can improve over time as more data becomes available. 	
Integration constraint(s):	<ul style="list-style-type: none"> ▪ Organized, quality AM production data: this is only limited available. Legacy data is not always useable. ▪ There is currently not a (commercial) system to use customer specific data. ▪ For the training of the AI-model at the user, training algorithms need to be provided. At this moment, free to use packages are used (Keras...) but this might change in the future. 	
Intended user(s):	<ul style="list-style-type: none"> ▪ AM users and service bureaus ▪ Existing (software) customers 	
Provider:	<ul style="list-style-type: none"> ▪ Materialise 	
Contact point:	<ul style="list-style-type: none"> ▪ Michel Janssens – michel.janssens@materialise.be 	
Condition(s) for reuse:	<ul style="list-style-type: none"> ▪ different business models can be applied: license, pay-per-use... ▪ OEM Contract 	

Latest update: 2022-09-27

Name: Build Quality Surveillance

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> Production monitoring data 	<ul style="list-style-type: none"> Train AI model on a set of monitoring data 	<ul style="list-style-type: none"> Detected anomalies Identified anomalies (diagnostic) Specific corrective actions.
Unique Selling Proposition(s):	<ul style="list-style-type: none"> The tool provides new insights and a better understanding in the physics of the AM process under surveillance. 	
Integration constraint(s):	<ul style="list-style-type: none"> The method requires full access to monitoring data. Most of the legacy systems are “black box”. Materialise has an open controller (MCP) enabling this. The solution still requires insight and is intended as a tool for the AM expert rather than the user of AM machines. Integration with open AI environments is still complex (the interactive notebook of Sirris might help here) 	
Intended user(s):	<ul style="list-style-type: none"> AM OEMs AM experts 	
Provider:	<ul style="list-style-type: none"> Materialise 	
Contact point:	<ul style="list-style-type: none"> Michel Janssens – michel.janssens@materialise.be 	
Condition(s) for reuse:	<ul style="list-style-type: none"> OEM Contract, bilateral collaboration 	
	<i>Latest update: 2022-09-27</i>	

Name: Part recognition

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> 3D Models (STL, STEP, Native CAD) Cameras with edge computing capabilities 	<ul style="list-style-type: none"> Train AI model on a limited set of virtual images of 3D model. 	<ul style="list-style-type: none"> Part ID
Unique Selling Proposition(s):	<ul style="list-style-type: none"> Reduces mix-ups of parts during manufacturing. Speeds up part identification 	
Integration constraint(s):	<ul style="list-style-type: none"> The functionality is still limited The system must still be “pressure tested” on robustness. The result is hardware specific 	
Intended user(s):	<ul style="list-style-type: none"> AM users and service bureaus 	
Provider:	<ul style="list-style-type: none"> Materialise 	
Contact point:	<ul style="list-style-type: none"> Michel Janssens – michel.janssens@materialise.be 	
Condition(s) for reuse:	<ul style="list-style-type: none"> different business models can be applied : license, pay-per-use... OEM Contract 	
	<i>Latest update: 2022-09-27</i>	

Name: A data-driven AI workflow for build-time estimation (BTE)

Input(s):	Main feature(s)	Output(s):
<ul style="list-style-type: none"> ▪ 3D Objects (STL, Native CAD) ▪ Correct BTE 	<ul style="list-style-type: none"> ▪ Extraction of features characterizing the 3D objects and automatic selection of the most important features for estimating the objects' build-time ▪ Use data-driven methods to divide the heterogenous set of 3D objects into homogeneous subsets ▪ Train independent AI models on the identified subsets for estimating build-times of the 3D objects ▪ Incremental learning and performance improvement as more data becomes available ▪ Capturing any validated BTE estimation into a reference repository 	<ul style="list-style-type: none"> ▪ Reference repository composed of 3D objects – features - AI BTE models ▪ Estimation of the build-time of 3D objects ▪ Interactive notebook implementing the validated AI workflow to be used for research experimentation
Unique Selling Proposition(s):	<ul style="list-style-type: none"> ▪ Build-times of 3D models are mainly predicted employing a global AI modeling approach using all parts in the heterogenous dataset ▪ As an alternative, the dataset can be divided into subsets of homogenous parts whose characteristics and building times are comparable ▪ This helps a data-driven algorithm to better learn the mapping between the 3D objects' characteristics and their printing time. ▪ Allows to gradually construct and maintain a reference repository composed of 3D objects, their characteristic features and the associated AI models for BTE estimation 	
Integration constraint(s):	<ul style="list-style-type: none"> ▪ Access to a large dataset of 3D objects to build independent AI models on the identified subsets of the data. This might not be readily available ▪ The AI workflow should be trained on a dataset in which the 3D objects were printed in the correct/optimal orientation and their build-times were computed accurately to make a reliable ground-truth 	
Intended user(s):	<ul style="list-style-type: none"> ▪ AM research labs ▪ AM users and service bureaus ▪ Existing (software) customers 	
Provider:	<ul style="list-style-type: none"> ▪ Sirris 	
Contact point:	<ul style="list-style-type: none"> ▪ Mahdi Tabassian – mahdi.tabassian@sirris.be 	
Condition(s) for reuse:	<ul style="list-style-type: none"> ▪ Different business models can be applied: license, pay-per-use... ▪ OEM contract 	

Latest update: 2022-09-27