



## EDISON Discussion Document

### Data Science Professional profiles definition (DSP): Proposal and Working Version

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## Executive summary

The EDISON project is designated to create a foundation for establishing a new profession of Data Scientist for European research and industry. The EDISON vision for building the Data Science profession will be enabled through the creation of a comprehensive framework for Data Science education and training that includes such components as Data Science Competence Framework (CF-DS), Data Science Body of Knowledge (DS-BoK), Data Science Model Curriculum (MC-DS), and Data Science Professional profiles definition.

The intended EDISON framework comprising of the mentioned above components will provide a guidance and a basis for universities to define their Data Science curricula and courses selection, on one hand, and for companies to better define a set of required competences and skills for their specific industry domain in their search for Data Science talents, on the other hand.

This document presents the results of the research and development in the EDISON project to define the Data Science Professional (DSP) profiles that important for defining the Data Scientist roles in the organisation and their alignment with the organizational goals and mission. The Data Science Professional definition is done in the context of the whole EDISON Data Science Framework.

The proposed DSP profiles are defined as an extension to current ESCO (European Skills, Competences, Qualifications and Occupations) taxonomy and is intended to be proposed for formal inclusion of the new Data Science professions family into the future ESCO taxonomy edition.

The proposed DSP profiles when adopted by the community will have multiple uses. First of all they will help organisations to plan their staffing for data related functions when migrating to agile data driven organizational model. The Human Resource (HR) departments can effectively use DSP profiles for vacancy description design and candidates assessment.

When used together with CF-DS, the DSP profiles can provide a basis for building interactive/web based tool for individual competences benchmarking against selected (or desirable) professional profiles as well as advising practitioners on the (up/re-) skilling path.

The definition of the Data Science Professional profiles together with other EDSF components will provide a formal basis for Data Science professional certification, organizational and individual skills management and career transferability.

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# 1 Introduction

The revolutionary value of data in modern computer powered e-Science is recognized in early works by technology visionaries. It is first described in the book by Tony Hey and others “The Fourth Paradigm” [3] and confirmed in the HLEG report “Riding the wave: How Europe can gain from the rising tide of scientific data” [4], that computational (and statistical) methods and data mining on large sets of scientific and experimental data will play a key role in discovering hidden and obscure relationships between processes and events that are necessary in order to make new scientific discoveries and support innovation in industry and the modern digital economy. Industry also recognises the benefits of Big Data technologies and the use of scientific methods in business/operational data analysis and in problem solving for managing enterprise operations, staying innovative and competitive, and being able to provide advanced customer-centric service delivery. Modern agile data driven companies are transforming their organizational to reflect the important role of data in optimizing business and operational processes. These changes have increased the demand for new types of specialists with strong technical background and deep knowledge of the data intensive technologies. This has been defined as a new profession of the Data Scientist.

This document presents the results of the research and development in the EDISON project to define the Data Science Professional profiles that important for defining the Data Scientist roles in the organisation and their alignment with organizational goals and mission. The Data Science Professional definition is done in the context of the whole EDISON Data Science Framework that includes such components as Data Science Competence Framework (CF-DS), Data Science Body of Knowledge (DS-BoK), Model Curriculum (MC-DC).

The intended EDISON framework comprising of the mentioned above components will provide a guidance and a basis for universities to define their Data Science curricula and courses selection, on one hand, and for companies to better define a set of required competences and skills for their specific industry domain in their search for Data Science talents, on the other hand. Similar to e-CF3.0, the proposed CF-DS, will provide a basis for building interactive/web based tool for building custom Data Science profiles and (self-)evaluate candidates compliance with a created profile.

The document has the following structure. Section 2 describes the EDSF and its components. Section 3 provides an overview of existing profession profiles definition frameworks for ICT and Data Science competences and skills definition including e-CF3.0, CWA 16458 (2012) European ICT profiles, European Skills, European Competences, Qualifications and Occupations (ESCO) framework. Section 4 presents the initial definition of the proposed DSP profiles as an extension to ESCO taxonomy. It also provides example mapping different profiles to CF-DS competences what can be used for building curricula and training programs customised for specific professional profiles and roles. The document concludes with the suggested further development to finalise the DSP profiles definition.

## 2 EDISON Data Science Framework

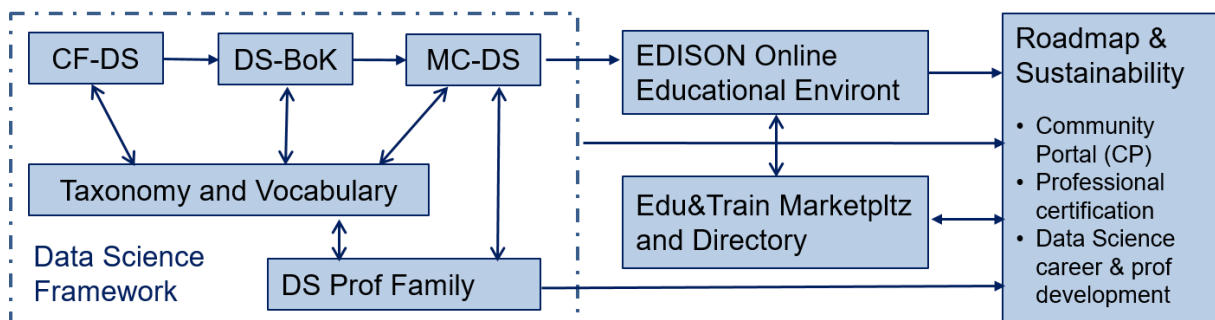
The EDISON project [1] is designated to create a foundation for establishing a new profession of Data Scientist for European research and industry. The EDISON vision for building the Data Science profession will be enabled through the creation of a comprehensive framework for Data Science education and training that includes such components as Data Science Competence Framework (CF-DS), Data Science Body of Knowledge (DS-BoK) and Data Science Model Curriculum (MC-DS).

Figure 1 below illustrates the main components of the EDISON Data Science Framework (EDSF) and their inter-relations that provides conceptual basis for the development of the Data Science profession (including reference to published discussion documents [2]):

- CF-DS – Data Science Competence Framework [3]
- DS-BoK – Data Science Body of Knowledge [4]
- MC-DS – Data Science Model Curriculum[5]
- Data Science Professional profiles and occupations taxonomy (DSP) [6]
- Data Science Taxonomy and Scientific Disciplines Classification

The proposed framework provides basis for other components of the Data Science professional ecosystem such as

- EDISON Online Education Environment (EOEE)
- Education and Training Marketplace and Directory
- Data Science Community Portal (CP) that also includes tools for individual competences benchmarking and personalized educational path building
- Certification Framework for core Data Science competences and professional profiles



**Figure 1 EDISON Data Science Framework components.**

The CF-DS provides the overall basis for the whole framework, its first version has been published in November 2015 and was used as a foundation for all following EDSF components developments. The CF-DS has been widely discussed at the numerous workshops, conferences and meetings, organised by the EDISON project and where the project partners contributed. The core CF-DS includes common competences required for successful work of Data Scientist in different work environments in industry and in research and through the whole career path. The future CF-DS development will include coverage of the domain specific competences and skills and will involve domain and subject matter experts. To allow easy use throughout all EDSF and in particular in DS-BoK and MC-DS, the CF-DS competences are enumerated (refer to recent CF-DS version at [3]).

The DS-BoK defines the Knowledge Areas (KA) for building Data Science curricula that are required to support required Data Science competences. DS-BoK is organised by Knowledge Area Groups (KAG) that correspond to the CF-DS competence groups. DS-BoK follows the same approach to collect community feedback and contribution: Open Access CC-BY community discussion document is published on the project website [2]. DS-BoK incorporates best practices in Computer Science and domain specific BoK's and includes KAs defined based on the Classification Computer Science (CCS2012) [7], components taken from other BoKs and proposed new KA to incorporate new technologies used in Data Science and their recent developments. The revised and updated

DS-BoK version used in this deliverable is presented in Appendix C and will be published as a next version of the DS-BoK discussion document after discussion with the EDISON Liaison Group (ELG) experts.

The MC-DS presented in this deliverable is built based on CF-DS and DS-BoK where Learning Outcomes are defined based on CF-DS competences and Learning Units are mapped to Knowledge Units in DS-BoK. Three mastery (or proficiency) levels are defined for each Learning Outcome to allow for flexible curricula development and profiling for different Data Science professional profiles. The proposed Learning outcomes are enumerated to have direct mapping to the enumerated competences in CF-DS. The preliminary version of MC-DS has been discussed at the first EDISON Champions Conference in June 2016 and collected feedback is incorporated in current version of MC-DS.

The DSP profiles and Data Science occupations taxonomy are defined based on and as an extension to European Skills, Competences, Qualifications and Occupations (ESCO) [8]. DSP profiles definition will create an important instrument to define effective organisational structures and roles related to Data Science positions and can be also used for building individual career path and corresponding competences and skills transferability between organisations and sectors.

The Data Science Taxonomy and Scientific Disciplines Classification will serve to maintain consistency between four core components of EDSF: CF-DS, DS-BoK, MC-DS, and DSP profiles. It is anticipated that successful acceptance of the proposed EDSF and its core components will require standardisation and contacts with the European and international standardisation bodies and professional organisations. This work is being done by the project as a part of the Dissemination and communication activity.

The EDISON Data Science professional ecosystem illustrated in Figure 1 uses core EDSF components shape and profiles the offered services and ensure the EDISON project sustainability. In particular, CF-DS and DS-BoK are used for individual competences and knowledge benchmarking and they are instrumental for constructing personalised learning path and professional (up/re-) skilling based on MC-DS.

### 3 Existing frameworks for ICT and Data Science competences and skills definition

This section provides a brief overview of existing standard and commonly accepted frameworks for defining professional profiles for general ICT occupations and currently defined data handling related professions. Appendix A provides additional overview of earlier works and publications that attempted to define required Data Science competences, skills and organisational roles.

#### 3.1 CWA 16458 (2012): European ICT Professional Profiles

The European ICT Professional Profiles CWA 17458 (2012) was created to provide a basis for compatible ICT profiles definition by organisations and a basis for defining new profiles by European stakeholders [9].

The CWA defines 23 main ICT profiles the most widely used by organisations by defining organisational roles for ICT worker, that are grouped into the six ICT Profile families:

- Business Management
- Technical Management
- Design
- Development
- Service and Operation
- Support

The European ICT Profile descriptions are reduced to core components and constructed to clearly differentiate profiles from each other. Further context-specific elements can be added to the Profiles according to the specific environments in which the Profiles are to be integrated. Figure 2 illustrates six ICT profile families and related main profiles which are non-exhaustive.

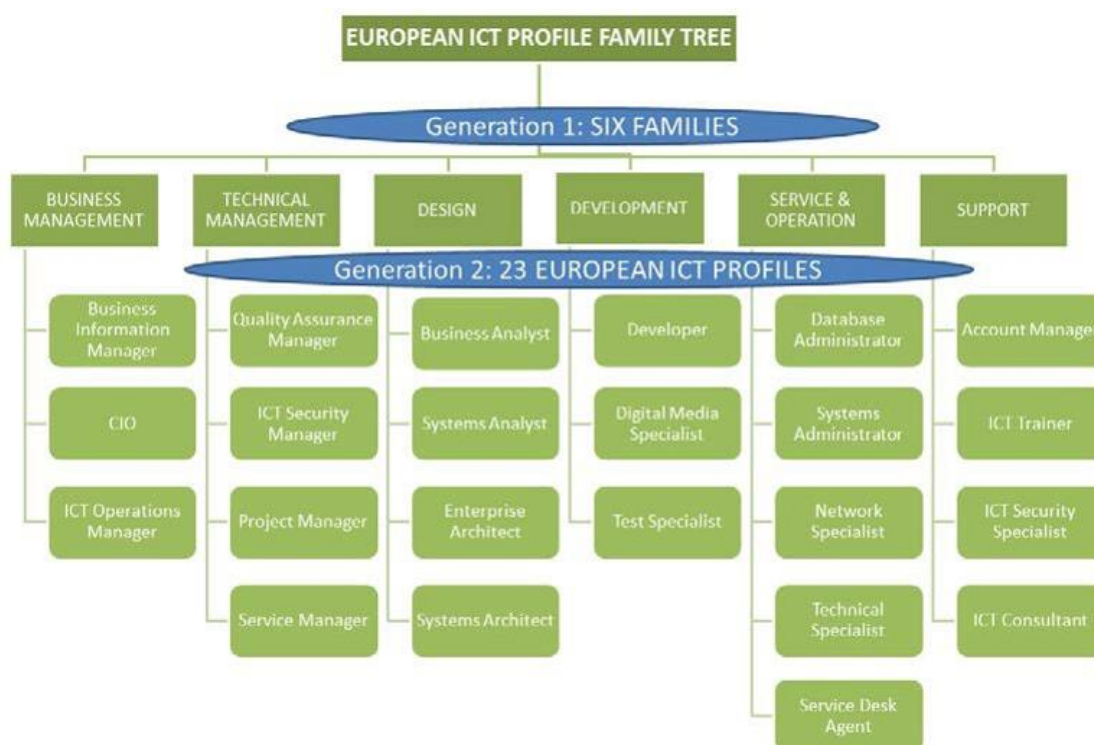


Figure 2. European ICT Profile Family Tree – Generation 1 and 2 as a shared European reference [9]

The 23 profiles constructed in CWA combined with e- competences from the e-CF3.0 [10], provide a pool for the development of tailored profiles that may be developed by European ICT sector players in specific contexts and with higher levels of granularity. The 23 Profiles cover the full ICT Business process; positioning them



into the e-CF Dimension 1 demonstrates this. Figure 4 below illustrates this together with the ICT Profiles family structure).

Figure 3 illustrates mapping between CWA families and e-CF3.0 competence areas and also CWA ICT profiles allocation to families and competence areas.

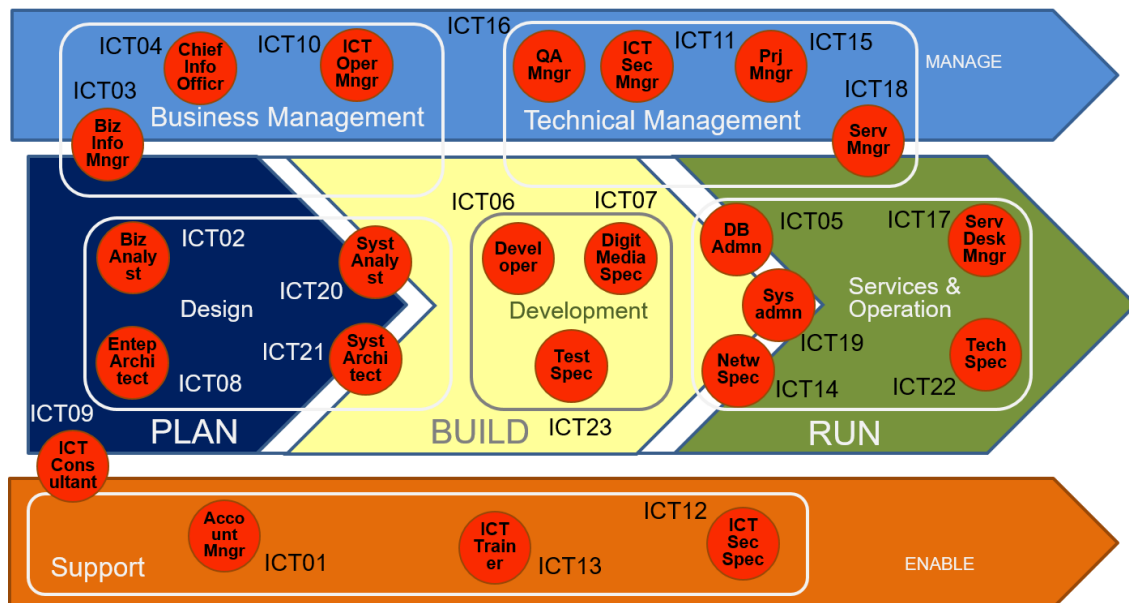


Figure 3. European ICT Professional Profiles structured by six families and positioned within the ICT Business Process (e-CF Dimension 1) (adopted from [5] and extended)

### 3.2 ESCO (European Skills, Competences, Qualifications and Occupations) framework and platform [8]

The Commission services launched the ESCO project in 2010 with an open stakeholder consultation. Currently, DG Employment, Social Affairs and Inclusion coordinates the development of ESCO with the support by the European Centre for the Development of Vocational Training Cedefop. Stakeholders are closely involved in the development and dissemination of ESCO.

The ESCO classification identifies and categorises skills, competences, qualifications and occupations relevant for the EU labour market and education and training. It systematically shows the relationships between the different concepts. ESCO has been developed in an open IT format, is available for use free of charge by everyone and can be accessed via the ESCO portal.

The first version of ESCO v0 was published on 23 October 2013. This version is based on the EURES classification but includes an enhanced semantic structure, cross-sector skills and competences and an initial small sample of qualifications. It includes the results of the Cross-Sector Reference Group, but not yet any sectoral updates.

ESCO v0 contains 4 761 occupations, around 5 000 skills and competences as well as some qualifications. As each concept in ESCO exists in all ESCO languages this amounts to more than 250 000 terms.

The qualifications pillar in ESCO v0 contains a small sample list of qualifications regulated at European level, international qualifications and certificates and licences linked to tasks, technologies, occupations or sectors. The list will be further developed in the next releases of ESCO. In addition, National Qualifications databases developed by the Member States and referenced to the European Qualifications Framework (EQF) [11] will, in the future, feed into the development of ESCO.

ESCO called for community review and contribution with the deadline of 31 December 2015<sup>1</sup>. Until end of 2016 the classification will be completely revised. The final product will be launched in 2017 as ESCO v1.

Table 1 contains data related occupations extracted from the ESCO classification together with related hierarchies. Table 2.3 is included for reference purposes to presents the ESCO top level occupations classification where data related occupations of different groups are highlighted in bold.

Table 1. Data related occupations in ESCO (2015) taxonomy

Occupations	Skills/Comp group	Hierarchy	Hierarchy	Top hierarchy
Security director (data processing/IT)	Database and network professionals not elsewhere classified	Database and network professionals	Information and communications technology professionals	Professionals
Security analyst (data processing/IT)				
Supervisor (data processing)				
Data processing investigator				
Data recorder	Database designers and administrators			
Operations manager (data processing)				
Data processing manager				
Data processing analyst				
Data processing supervisor	Systems administrators			
Data processing consultant				
Data processing strategist	Systems analysts	Software applications developers and analysts		
Operations technician (data processing)	Information and communications technology operations technicians	Information and communications technology operations and user support technicians	Information and communications technicians	Technicians and associate professionals
Access supervisor, data processing/IT	Information and communications technology service managers	Production and specialised services managers		Managers

Table 2. ESCO top occupation hierarchy

ESCO Occupations top level hierarchy	
Armed forces occupations	
<b>Clerical support workers</b>	
	Numerical and material recording clerks
	Other clerical support workers
	Customer services clerks
	<b>General and keyboard clerks</b>
Craft and related trades workers	
Elementary occupations	
<b>Managers</b>	

<sup>1</sup> EDISON project provided a number of comments and suggestions to Data Science and education methods related terms and definitions.

	Administrative and commercial managers
	Chief executives, senior officials and legislators
	Hospitality, retail and other services managers
	<b>Production and specialised services managers</b>
	Plant and machine operators and assemblers
	<b>Professionals</b>
	Teaching professionals
	<b>Science and engineering professionals</b>
	Health professionals
	Legal, social and cultural professionals
	Business and administration professionals
	<b>Information and communications technology professionals</b>
	Service and sales workers
	Skilled agricultural, forestry and fishery workers
	<b>Technicians and associate professionals</b>
	Health associate professionals
	<b>Information and communications technicians</b>
	Legal, social, cultural and related associate professionals
	<b>Science and engineering associate professionals</b>
	Business and administration associate professionals

### 3.3 Existing definitions of Data Scientist

There is no well established definition of the Data Scientist due to a number of competences and skills expected from these specialists. We will take as a basis the definition provided in the NIST SP1500-1 document [12]:

*“A **Data Scientist** is a practitioner who has sufficient knowledge in the overlapping regimes of expertise in business needs, domain knowledge, analytical skills, and programming and systems engineering expertise to manage the end-to-end scientific method process through each stage in the **big data lifecycle**.”* The document defines the following groups of skills required from the Data Scientists: domain experience, statistics and data mining, and engineering skills [12].

Initial attempt to define the Data Scientist has been made by O’Reilly Strata Survey (2013) (see [13] and Appendix A) which recognised creativity as an important feature of Data Scientist.

Other definitions [14, 15] admit such desirable features as ability to solve variety of business problems, optimize performance and suggest new services for the organisation employing Data Scientist. Many practitioners admit a need for a successful Data Scientist to develop a special mindset, to be statistically minded, understand raw data and “appreciate data as a first class product” [16].

The qualified Data Scientist should be capable of working in different roles in different projects and organisations such as Data Engineer, Data Analyst or Data Architect, Data Steward, etc., and possess the necessary skills to effectively operate components of the complex data infrastructure and processing applications through all stages of the Data lifecycle till the delivery of expected scientific and business values to science and/or industry.

## 4 Defining Data Science Professional profiles

This section presents initial results on defining the Data Science Professional profiles that can be also called Data relates occupations family. They are defined extension to the ESCO occupations taxonomy. The proposed new occupations are placed in four top classification groups: managers (for managerial roles); Professionals (for applications developers and for infrastructure engineers); technicians and associate professionals (for operators and technicians); and clerical support workers (for data curators and stewards).

### 4.1 Taxonomy of Data Science Occupations according to ESCO Hierarchy

The presented here initial taxonomy of Data Science professional roles is based on the ESCO occupations classification and their competences and organisational roles are defined similar to CWA 16458 ICT profiles. Table 3 presents them in the context of the ESCO classification hierarchy, only Data Science related top level groups are presented (for overall top level ESCO occupations hierarchy refer to section 3.2 and Table 2).

The following suggestions were used when constructing the proposed taxonomy:

- Data Scientist occupations depending on organisational role can be placed in the following top level hierarchies:
  - Managers (for managerial roles);
  - Professionals (for analytics applications developers and for infrastructure and datacenter engineers);
  - Technicians and associate professionals (for operators and technicians)
- Correspondingly, new 3rd level occupation groups are proposed:
  - Data Science/Big Data Infrastructure Managers
  - Data Science Professionals
  - Data Science technology professionals
  - Data and information entry and access
- Group of occupations related to digital librarians, data archives management, data curations and support currently placed in the 3rd group *“Professionals > Information and communications technology professionals > Data Science technology professionals > Data handling professionals not elsewhere classified”*, however potentially it can also put in a new 2nd level group *“Clerical support workers > Data handling support workers (alternative)”*. Motivation for this is growing need for data support workers in all domain of human activities in the digital data driven economy.
- It is recognised that existing ESCO group “Database and network professionals” should extended with new occupations (or professions) related to Big Data and scientific data related profiles which examples are included in the table: Large scale (cloud) database administrator/operator and Scientific database administrator/operator, however further identification of such occupations need to be done.

Table 3. Data Science occupations extension to ESCO classification

Top level	Hierarchies existing and new		Occupations group (if any)	Occupations
Managers				
	Production and specialised services managers	Data Science/Big Data Infrastructure Managers		DSP01 Data Science (group) Manager
			Research Infrastructure Managers	DSP02 Data Science Infrastructure Manager
				DSP03 Research Infrastructure Manager
Professionals				
	Science and engineering professionals	Data Science Professionals	Data science professionals not elsewhere classified	DSP04 Data Scientist
				DSP05 Data Science Researcher
				DSP06 Data Science Architect
				DSP07 Data Science (Application) Programmer/Engineer
				DSP08 (Big) Data Analyst
				DSP09 Business Analyst
	Information and communications technology professionals	Data Science technology professionals	Data handling professionals not elsewhere classified	DSP10 Data Steward
				DSP11 Digital Data Curator
				DSP12 Digital Librarian
				DSP13 Data Archivist
	Science and engineering professionals	Database and network professionals	Large scale (cloud) data storage designers and administrators	DSP14 Large scale (cloud) database designer*)
			Database designers and administrators	DSP15 Large scale (cloud) database administrator*)
			Database and network professionals not elsewhere classified	DSP16 Scientific database administrator*)
Technicians and associate professionals				
	Science and engineering associate professionals	Data Science Technology Professionals	Data Infrastructure engineers and technicians	DSP17 Big Data facilities Operators
				DSP18 Large scale (cloud) data storage operators
			Database and network professionals not elsewhere classified	DSP19 Scientific database operator*)
Clerical support workers				
	General and keyboard clerks			
		Data handling and support workers	Data and information entry and access	DSP20 Data entry/access desk/terminal workers

				DSP21 Data entry field workers
				DSP22 User support data services

## 4.2 Definition of the Data Science Professional profiles

This section provides definition of the Data Science Professional profiles by defining their competences and organisational roles. The proposed definition can be instrumental in defining education and training profiles for students and for practitioners to acquire necessary competences and knowledge for specific professional profiles or occupations. It can be also used for defining certification profiles or career path building.

The Data Scientist occupation groups are placed in the following top level ESCO hierarchies:

- Managers (for managerial roles);
- Professionals (for analytics applications developers and for infrastructure and datacenter engineers);
- Technicians and associate professionals (for operators and technicians)
- Optionally, some data management occupations can be also placed into the Clerical support workers group such as digital data archivist, digital librarians.

Correspondingly, the following new 3rd level occupation groups are proposed:

- Data Science/Big Data Infrastructure Managers
- Data Science Professionals
- Data Science technology professionals
- Data and information entry and access (this is a candidate group under Clerical support workers top level hierarchy)

It is proposed that the existing ESCO group “Database and network professionals” should be extended with new occupations (or professions) related to Big Data or cloud based databases: Large scale (cloud) database administrator/operator and Scientific database administrator/operator, however further identification of such occupations needs to be done.

A group of occupations related to digital librarians, data archives management, data stewardship and data curation are currently placed in the 3rd proposed group:

*Professionals > Information and communications technology professionals > Data Science technology professionals > Data handling professionals not elsewhere classified,*

however potentially it can also be added in a new 2nd level group “Clerical support workers > Data handling support workers (alternative)”. The motivation for this is a growing need for data support workers in all domains of human activities in the digital data driven economy.

To ensure a smooth Data Science professions acceptance by industry and employment bodies, the proposed profiles should be compatible with the relevant standards ESCO, eCFv3.0 (future CEN standard EN 16324) [10], CWA 16458 2012 ICT Profiles [9].

Table 4 provides an initial definition of the identified Data Science professional profiles collected from job advertisements, blogs and recent discussions at different forums, in particular, with the Research Data Alliance, and digital curation and data preservations communities.

**Table 4 Data Science professional profiles definition**

Profile ID	Data Science Profile title	Data Science Profile Summary statement	Alternative titles and legacy titles
<b>Managers</b>			

<b>DSP01</b>	Data Science (group) Manager	Proposes, plans and manages functional and technical evolutions of the data science operations within the relevant domain (technical, research, business).	Data analytics department manager
<b>DSP02</b>	Data Science Infrastructure Manager	Proposes plans and manages functional and technical evolutions of the big data infrastructure within the relevant domain (technical, research, business).	Big Data Infrastructure Manager
<b>DSP03</b>	Research Infrastructure Manager	Proposes plans and manages functional and technical evolutions of the research infrastructure within the relevant scientific domain.	Research Infrastructure data storage facilities manager
<b>Professionals</b>			
<b>DSP04</b>	Data Scientist	Data scientists find and interpret rich data sources, manage large amounts of data, merge data sources, ensure consistency of data-sets, and create visualisations to aid in understanding data. Build mathematical models, present and communicate data insights and findings to specialists and scientists, and recommend ways to apply the data.	Data Analyst
<b>DSP05</b>	Data Science Researcher	Data Science Researcher applies scientific discovery research/process, including hypothesis and hypothesis testing, to obtain actionable knowledge related to scientific problem, business process, or reveal hidden relations between multiple processes.	Data Analyst
<b>DSP06</b>	Data Science Architect	Designs and maintains the architecture of Data Science applications and facilities. Creates relevant data models and processes workflows.	System Architect, Applications architect
<b>DSP07</b>	Data Science (Application) Programmer/Engineer	Designs/develops/codes large data (science) analytics applications to support scientific or enterprise/business processes.	Scientific Programmer
<b>DSP08</b>	Data Analyst	Analyses large variety of data to extract information about system, service or organisation performance and present them in usable/actionable form	
<b>DSP09</b>	Business Analyst	Analyses large variety of data Information System for improving business performance.	Business Development Manager (Data science role)
<b>Professional (data handling/management)</b>			
<b>DSP10</b>	Data Stewards	Plans, implements and manages (research) data input, storage, search, presentation; creates data model for domain specific data; support and advice domain scientists/researchers	
<b>DSP11</b>	Digital data curator	Finds, selects, organises, shares (exhibits) digital data collections, maintains their integrity, up-to-date status and freshness, discoverability	Digital curator, digital archivist, digital librarian

<b>DSP12</b>	Digital Librarians	Selection, acquisition, organization, accessibility and preservation of digital information/library. Manages digital materials, takes a lead role in the creation, maintenance and stewardship of digital collections, including the digitization of special collections. Develops strategies for effective management and preservation of library digital assets.	Digital data curator
<b>DSP13</b>	Data Archivists	Maintain historically significant collections of datasets, documents and records, other electronic data, and seek out new items for archiving.	Digital Archivists
<b>Professional (database)</b>			
<b>DSP14</b>	Large scale (cloud) database designer	Designs/develops/codes large scale data bases and their use in domain/subject specific applications according to the customer needs.	Large scale (cloud) database developer
<b>DSP15</b>	Large scale (cloud) database administrator	Designs and implements, or monitors and maintains large scale cloud databases	
<b>DSP16</b>	Scientific database administrator	Designs and implements, or monitors and maintains large scale scientific databases	Large scale (cloud) database administrator
<b>Technicians and associate professionals</b>			
<b>DSP17</b>	Big Data facilities Operator	Manages daily operation of facilities, resources, and responds to customer requests. Includes all operations related to data management and data lifecycle	
<b>DSP18</b>	Large scale (cloud) data storage operator	Manages daily operation of cloud storage, Including related to data lifecycle, and responds to requests from storage users	
<b>DSP19</b>	Scientific database operator	Manages daily operation of scientific databases, Including related to data lifecycle, and responds to requests from database users	Large scale (cloud) data storage operators
<b>Clerical support workers (general and keyboard workers)</b>			
<b>DSP20</b>	Data entry/access worker	Enter data into data management systems directly reading them from source, documents or obtained from people/users	Data entry desk/terminal worker
<b>DSP21</b>	Data entry field workers	The same work done on field when collecting data from disconnected sensors or doing direct counting or reading	
<b>DSP22</b>	User support data services	Provides support to users to entry their data into governmental service and user facing applications	



Table 5 provides a mapping between professional profiles and Data Science competence groups, which are Defined in CF-DS [3] together with the suggested ranking the relevance of different competence groups to corresponding Data Science profiles (where 1 is less relevant and 5 is highly relevant).

The CF-DS competence groups are defined as follows (for full definition of the CF-DS competence see CF-FS document [3] and Appendix)

**Data Analytics (DSDA)**

Use appropriate statistical techniques and predictive analytics on available data to deliver insights and discover new relations

**Data Management (DSDM)**

Develop and implement a data management strategy for data collection, storage, preservation, and availability for further processing.

**Data Science Engineering (DSENG)**

Use engineering principles to research, design, develop and implement new instruments and applications for data collection, analysis and management

**Scientific and Research Methods (DSRM) for research domain and Business Process Management (DSBP)**

Create new understandings and capabilities by using the scientific method (hypothesis, test/artefact, evaluation) or similar engineering methods to discover new approaches to create new knowledge and achieve research or organizational goals

**Data Science Domain Knowledge (DSDK)**

Use domain knowledge (scientific or business) to develop relevant data analytics applications, and adopt general Data Science methods to domain specific data types and presentations, data and process models, organizational roles and relations

Table 1 Mapping Data Science competence groups to the proposed profiles

Profile ID	Data Science Profile title	Data Science Competences Groups (relevance 1 - low, 5 – high)				
		DSDA Data Analytics	DSDM Data Managem ent	DSENG Data Science Engineering	DSRM Research Methods, Business methods	DSDK Subject Domain
Managers						
DSP01	Data Science (group) Manager	3	4	3	3	2
DSP02	Data Science Infrastructure Manager	2	4	4	2	2
DSP03	Research Infrastructure Manager	2	4	4	3	2
Professionals						
DSP04	Data Scientist	5	3	4	5	3
DSP05	Data Science Researcher	4	3	2	5	4
DSP06	Data Science Architect	4	3	5	3	3
DSP07	Data Science (Application) Programmer/Engineer	4	2	5	3	4
DSP08	Data Analyst	5	3	3	3	4
DSP09	Business Analyst	5	3	3	4	5
Professional (data handling/ management)						
DSP10	Data Stewards	3	5	3	3	3
DSP11	Digital data curator	1	5	2	2	3
DSP12	Digital Librarians	2	5	2	2	3
DSP13	Data Archivists	1	5	1	1	3
Professional (database)						
DSP14	Large scale (cloud) database designer	2	4	4	3	3
DSP15	Large scale (cloud) database administrator	2	4	3	2	3
DSP16	Scientific database administrator	2	4	3	2	3
Technicians and associate professionals						
DSP17	Big Data facilities Operator	1	4	4	2	3
DSP18	Large scale (cloud) data storage operator	1	4	3	1	1
DSP19	Scientific database operator	1	4	3	2	3
Clerical support workers (general and keyboard workers)						

DSP20	Data entry/access worker	2	1	2
DSP21	Data entry field workers	2	1	2
DSP22	User support data services	3	2	2

## 5 Conclusion and further developments

This document provided information about the Data Science Professional profiles definition as a part of the overall EDISON Data Science Framework. The presented DSP profiles are defined based on and as an extension to ESCO taxonomy. They are enumerated and includes the following groups: Managers (DSP01-DSP03), Professionals (DSP04-DSP09), Professional Data Management/Handling (DSP10-DSP13), Professional Technical (DSP14-DSP16), Professional Technicians (DSP17-DSP19), Support and clerical workers (DSP20 – DSP22).

The document also illustrates example how identified in CF-DS competences can be assigned to different profiles.

The presented information is a result of the EDISON project team discussion and a subject for further review and discussion by the research and industry community.

Further developments will be focused on the following activities:

- Finalise the Data Science Professional profiles definition by collecting feedback by consulting ESCO committee and practitioners from research and industry on their Human Resource management practices. Provide suggestion for ESCO extension with Data Science and data related occupations
- Run the community survey and use a customisable questionnaire to run few key interviews, primarily with experts and top executives at universities and companies.

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## Acronyms

Acronym	Explanation
ACM	Association for Computer Machinery
BABOK	Business Analysis Body of Knowledge
CCS	Classification Computer Science by ACM
CF-DS	Data Science Competence Framework
CS	Computer Science
DM-BoK	Data Management Body of Knowledge by DAMAI
DS-BoK	Data Science Body of Knowledge
ETM-DS	Data Science Education and Training Model
EUDAT	<a href="http://eudat.eu/what-eudat">http://eudat.eu/what-eudat</a>
EGI	European Grid Initiative
ELG	EDISON Liaison Group
EOSC	European Open Science Cloud
ERA	European Research Area
ESCO	European Skills, Competences, Qualifications and Occupations
ICT	Information and Communication Technologies
IEEE	Institute of Electrical and Electronics Engineers
IPR	Intellectual Property Rights
LIBER	Association of European Research Libraries
MC-DS	Data Science Model Curriculum
NIST	National Institute of Standards and Technologies of USA
PM-BoK	Project Management Body of Knowledge
PRACE	Partnership for Advanced Computing in Europe
RDA	Research Data Alliance
SWEBOK	Software Engineering Body of Knowledge

## Appendix A. Overview: Studies, reports and publications related to Data Science competences and skills definition

### A.1. O'Reilly Strata Survey (2013)

O'Reilly Strata industry research [25] defines the four Data Scientist profession profiles and their mapping to the basic set of technology domains and competencies as shown in Figure A.1. The four profiles are defined based on the Data Scientists practitioners self-identification:

- Data Businessperson
- Data Creative
- Data Developer
- Data Researcher

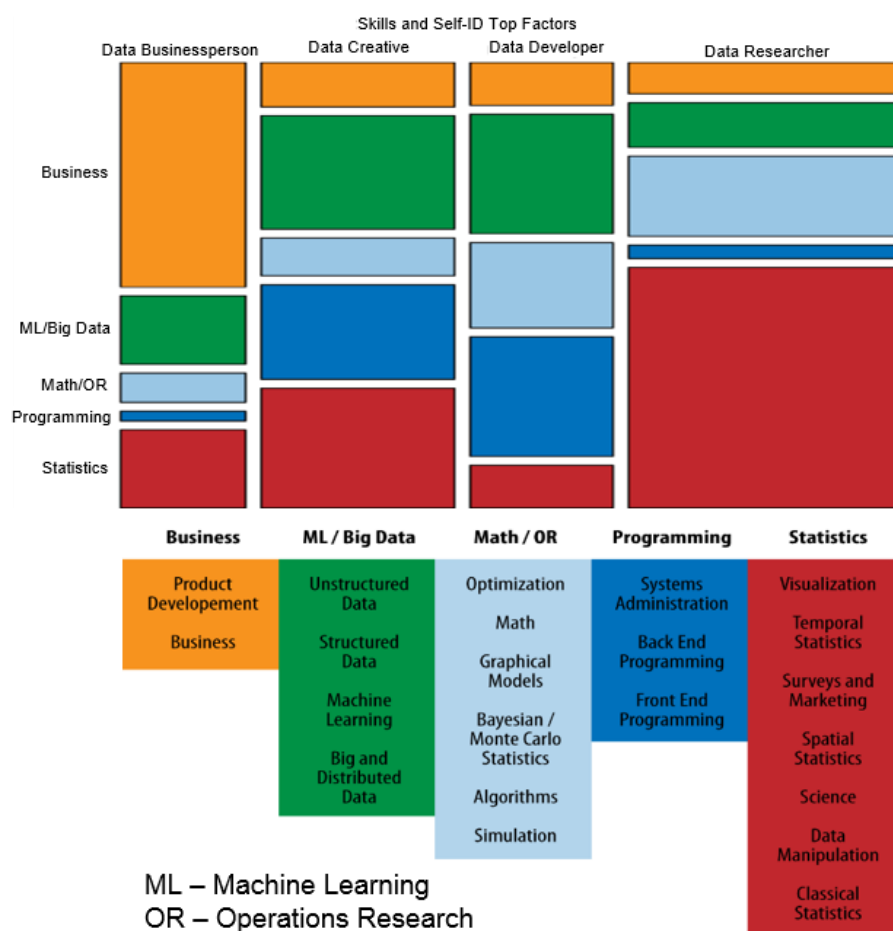


Figure A.1. Data Scientist skills and profiles according to O'Reilly Strata survey [25]

Table A.1 below lists skills for Data Science that are identified in the study. They are very specific in technical sense but provide useful information when mapped to the mentioned above Data Science profiles. We will refer to this study in our analysis of CF-DS and related competence groups.

Table A.1. Data Scientist skills identified in the O'Reilly Strata study (2013)

Data Science Skills	Examples -> Knowledge and skills
Algorithms	computational complexity, CS theory
Back-End Programming	JAVA/Rails/Objective C

Bayesian/Monte-Carlo Statistics	MCMC, BUGS
Big and Distributed Data	Hadoop, Map/Reduce
Business	management, business development, budgeting
Classical Statistics	general linear model, ANOVA
Data Manipulation	regexes, R, SAS, web scraping
Front-End Programming	JavaScript, HTML, CSS
Graphical Models	social networks, Bayes networks
Machine Learning	decision trees, neural nets, SVM, clustering
Math	linear algebra, real analysis, calculus
Optimization	linear, integer, convex, global
Product Development	design, project management
Science	experimental design, technical writing/publishing
Simulation	discrete, agent-based, continuous)
Spatial Statistics	geographic covariates, GIS
Structured Data	SQL, JSON, XML
Surveys and Marketing	multinomial modeling
Systems Administration	*nix, DBA, cloud tech.
Temporal Statistics	forecasting, time-series analysis
Unstructured Data	noSQL, text mining
Visualization	statistical graphics, mapping, web-based data visualisation

### A.3. UK Study on demand for Big Data Analytics Skills (2014)

The study “Big Data Analytics: Assessment of demand for Labour and Skills 2013-2020” [30] provided extensive analysis of the demand side for Big Data specialists in UK in forthcoming year. Although majority of roles are identified as related to Big Data skills, it is obvious that all these roles can be related to more general definition of the Data Scientist as an organisational role working with Big Data and Data Intensive Technologies.

The report lists the following Big Data roles:

- Big Data Developer
- Big Data Architect
- Big Data Analyst
- Big Data Administrator
- Big Data Consultant
- Big Data Project Manager
- Big Data Designer
- Data Scientist

### A.4. IWA Data Science profile

Italian Web Association (IWA) published the WSP-G3-024. Data Scientist Profile for web related projects [31]. It provide a god example of domain specific definition of the Data Science competences, skills and organisational responsibilities, it suggests also mapping to e-CF3.0 competences.

The Data Scientist is defined as “Professional that owns the collection, analysis, processing, interpretation, dissemination and display of quantitative data or quantifiable organization for analytical, predictive or strategic.”

The profile contains the following sections:



- Concise definition
- Mission
- Documentation produced
- Main tasks
- Mapping to e-CF competences
- Skills and knowledge
- Application area of KPI
- Qualifications and certifications (informational)
- Personal attitudes (informational)
- Reports and reporting lines (informational)

For reference purpose, it is worth to mention that IWA Data Scientist profile maps its competences and skills to the following e-CF3.0 competences:

- A.6. Application design: Level e-3
- A.7. Monitoring of technological Bertrand: Level e-4
- B.1. Development of applications: Level e-2
- B.3. Testing: Level e-3
- B.5. Production of documentation: Level e-3
- C.1. User assistance: Level e-3
- C.3. Service Delivery: Level e-3
- C.4. Management Problem: Levels e-3, e-4.

## Appendix B. Data Science Competence Framework (CF-DS) Excerption

This Appendix contains excerption from the original CF-DS document [3] that describes identified Data Science competences. The full CF-DS definition including both competences and skills is available in the CF-DS document.

### B.1. Identified Data Science Competence Groups

The results of analysis presented here provides a basis and justification for defining two (new) competence areas that have not been explicitly defined in previous studies and frameworks. In particular, the proposed CF-DS competence and skills groups include:

3 competence groups identified in the NIST document and confirmed by analysis of collected data:

- Data Analytics including statistical methods, Machine Learning and Business Analytics
- Engineering: software and infrastructure
- Subject/Scientific Domain competences and knowledge

2 new identified competence groups that are highly demanded and are specific to Data Science

- *Data Management, Curation, Preservation (new)*
- *Scientific or Research Methods (new)*

Data Management, curation and preservation is already being conducted within existing (research) data related professions such as data archivist, data manager, digital data librarian, and others. Data management is also important component of European Research Area policy. It is extensively addressed by the Research Data Alliance and supported numerous projects, initiatives and training programmes<sup>2</sup>.

Knowledge of the scientific research methods and techniques is something that makes Data Scientist profession different from all previous professions. Data Scientist is expected to have ability to find hidden value in raw data collected from scientific experiments or organisational activity. Tasks that are quite similar to researcher's work.

From the education and training point of view the identified competences can be treated or linked to expected learning or training outcomes. This aspect is discussed in details in other EDSF documents DS-BoK and MC-DS.

New identified competence areas provide a better basis for defining education and training programmes for Data Science related jobs, re-skilling and professional certification.

Table B.1 provides an example of competences definition for different groups that are extracted from the collected information. It indicates that all identified groups are demanded by companies and they have expected place in the corporate structure and activities.

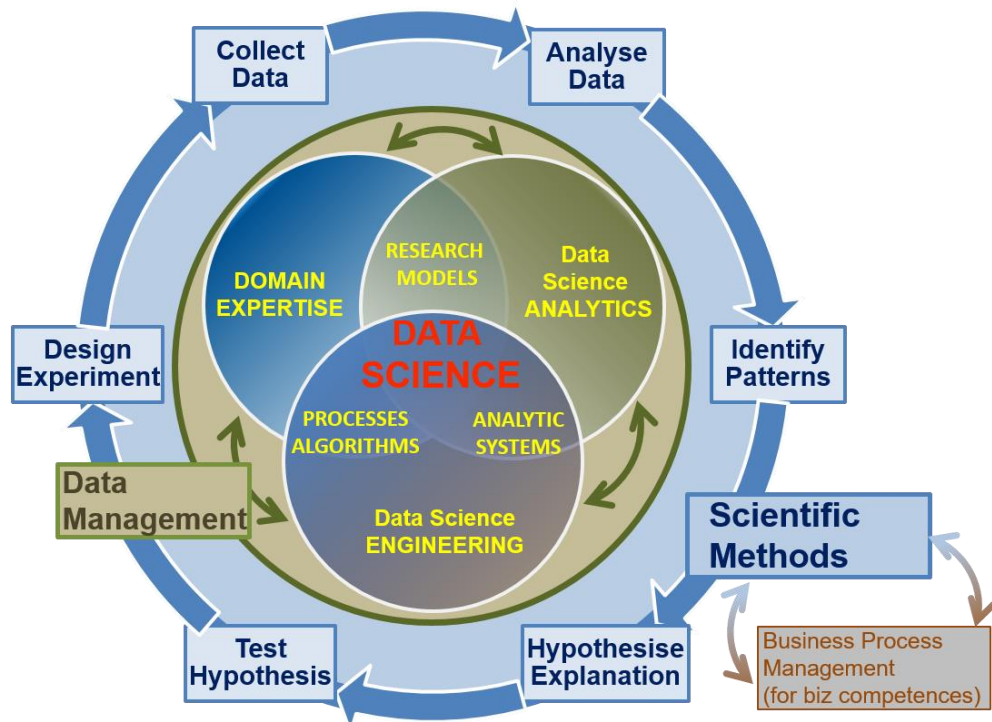
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<sup>2</sup> Research Data Alliance Europe <https://europe.rd-alliance.org/>

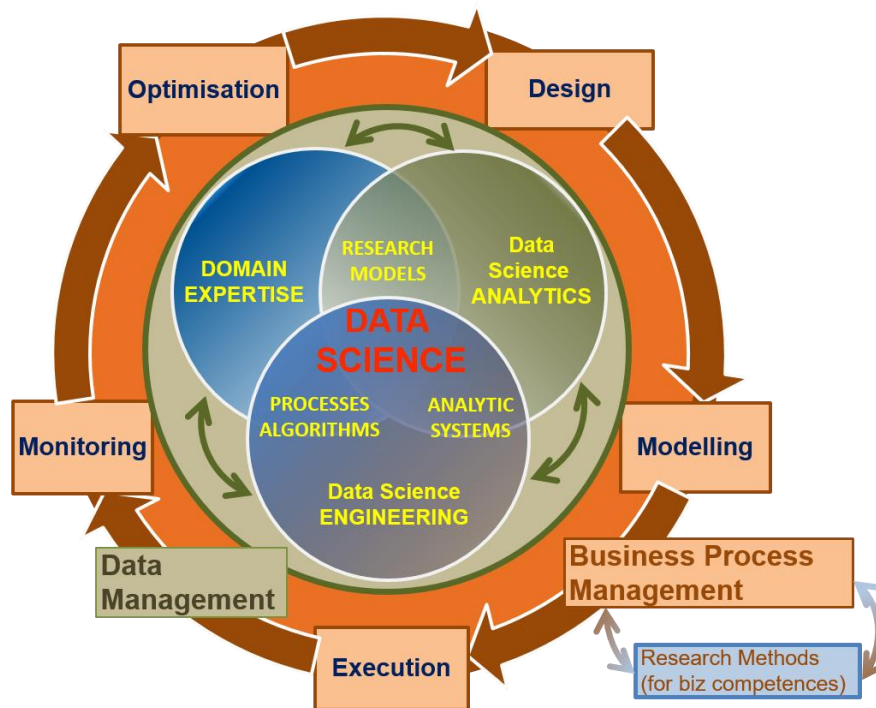
Table B.1. Enumerated competences definition for different Data Science competence groups

<b>Data Science Analytics (DSDA)</b>	<b>Data Management (DSDM)</b>	<b>Data Science Engineering (DSENG)</b>	<b>Scientific/ Research Methods (DSRM)</b>	<b>DS Domain Knowledge, e.g., Business Apps (DSDK)</b>
Use appropriate statistical techniques and predictive analytics on available data to deliver insights and discover new relations	Develop and implement data management strategy for data collection, storage, preservation, and availability for further processing.	Use engineering principles to research, design, develop and implement new instruments and applications for data collection, analysis and management	Create new understandings and capabilities by using the scientific method (hypothesis, test/artefact, evaluation) or similar engineering methods to discover new approaches to create new knowledge and achieve research or organisational goals	Use domain knowledge (scientific or business) to develop relevant data analytics applications, and adopt general Data Science methods to domain specific data types and presentations, data and process models, organisational roles and relations
DSDA01 Use predictive analytics to analyse big data and discover new relations	DSDM01 Develop and implement data strategy, in particular, in a form of Data Management Plan (DMP)	DSENG01 Use engineering principles to research, design, prototype, data analytics applications, or develop structures, instruments, machines, experiments, processes, systems	DSRM01 Create new understandings and capabilities by using the scientific method (hypothesis, test, and evaluation) or similar engineering research and development methods	DSDK01 Understand business and provide insight, translate unstructured business problems into an abstract mathematical framework
DSDA02 Use appropriate statistical techniques on available data to deliver insights	DSDM02 Develop and implement relevant data models, including metadata	DSENG02 Develop and apply computational solutions to domain related problems using wide range of data analytics platforms	DSRM02 Direct systematic study toward a fuller knowledge or understanding of the observable facts, and discovers new approaches to achieve research or organisational goals	DSDK02 Use data to improve existing services or develop new services
DSDA03 Develop specialized analytics to enable agile decision making	DSDM03 Collect and integrate different data source and provide them for further analysis	DSENG03 Develops specialized data analysis tools to support executive decision making	DSRM03 Undertakes creative work, making systematic use of investigation or experimentation, to discover or revise knowledge of reality, and uses this knowledge to devise new applications	DSDK03 Participate strategically and tactically in financial decisions that impact management and organizations
DSDA04 Research and analyze complex data sets, combine different sources and types of data to improve analysis.	DSDM04 Develop and maintain a historical data repository of analysis results (data provenance)	DSENG04 Design, build, operate relational non-relational databases	DSRM04 Ability to translate strategies into action plans and follow through to completion.	DSDK04 Provides scientific, technical, and analytic support services to other organisational roles
DSDA05 Use different data analytics platforms to process complex data	DSDM05 Ensure data quality, accessibility, publications (data curation)	DSENG05 Develop solutions for secure and reliable data access	DSRM05 Contribute to and influence the development of organizational objectives	DSDK05 Analyse customer data to identify/optimize customer relations actions
DSDA06 Visualise complex and variable data.	DSDM06 Manage IPR and ethical issues in data management	DSENG06 Prototype new data analytics applications	DSRM06 Apply ingenuity to complex problems, develop innovative ideas	DSDK06 Analyse multiple data sources for marketing purposes

Figures B.1 (a) and (b) provide graphical presentation of relations between identified competence groups as linked to Scientific Methods or to Business Process Management. The figure illustrates importance of Data Management competences and skills and Scientific/Research Methods or Business Processes knowledge for all categories and profiles of Data Scientists.



(a) Data Science competence groups for general or research oriented profiles.



(b) Data Science competence groups for business oriented profiles.

Figures B.1. Relations between identified Data Science competence groups for (a) general or research oriented and (b) business oriented professions/profiles: Data Management and Scientific/Research Methods or Business Processes Management competences and knowledge are important for all Data Science profiles.

The Scientific Methods typically include the following stages (see section 3.2.2 for reference to existing scientific methods definitions):

- Design Experiment
- Collect Data
- Analyse Data
- Identify Patterns
- Hypothesise Explanation
- Test Hypothesis

There are a number Business Process Operations models depending on their purpose but typically they contain the following stages that are generally similar to those for Scientific methods, in particular in collecting and processing data (see reference to exiting definitions in section 3.2.3):

- Design
- Model/Plan
- Deploy & Execute
- Monitor & Control
- Optimise & Re-design

The identified demand for general competences and knowledge on Data Management and Research Methods needs to be implemented in the future Data Science education and training programs, as well as to be included into re-skilling training programmes. It is important to mention that knowledge of Research Methods does not mean that all Data Scientists must be talented scientists; however, they need to know general research methods such as formulating hypothesis, applying research methods, producing artefacts, and evaluating hypothesis (so called 4 steps model). Research Methods training are already included into master programs and graduate students.

In summary, consolidation of the presented initial version of CF-DS was challenging task due to variety of information and required expertise. It is anticipated that it will undergo further improvement involving external and subject domain experts via EDISON Liaison Group and community involvement. Other project activities will provide feedback on necessary improvements and details.

It is a challenging task to include all required subjects and knowledge into education and training programs. It will require search for new approaches in Data Science education what will be a subject for subsequent EDISON project activities and work items.

## **B.2. Identified Data Science Skills**

For identified Data Science skills and technical platforms knowledge refer to the original CF-DS document [3].