# Enabling a Data-informed Public Sector: An Introductory Course to

**BDTI Essentials** 



Session 2: Data Cleaning and Transformation

### Welcome! The webinar will begin shortly.

#### **DG CNECT**

Directorate-General for Communications Networks, Content and Technology **DG DIGIT** Directorate-General for Digital Services



DIGITA

PROGRAMMI

# Some housekeeping

- The Chat and Q&A widgets are located to the right of the presenter screen
- Please post session-related questions in Q&A
- Microphones and video are off by default, attendees cannot turn them on
- The session will be recorded and available shortly after the end of the session on our website

### Do

- Participate in the chat during the session
- Ask questions during the Q&A portion at the end
- Give feedback after the session
- Join the dedicated discussion board

### Don't

- Self-promote
- Disrespect anyone







### Enabling a Data-informed Public Sector: An Introductory Course to

### **BDTI Essentials**



Session 2: Data Cleaning and Transformation

### Agenda

- Welcome [11:00-11:05]
- Session 1 Recap [11.05-11.15]
- $\circ$  Data cleaning and transformation [11:15-12:00]
- O Q&A [12:00-12:10]
- Wrap up [12:10 12:15]



🗏 🏵 ⊾ 🕲 🖶

### **Course discussion board**





Here, you can ask questions and discuss topics related to the BDTI Essentials Online Course. The board is moderated by the BDTI team, so your questions will be answered quickly. We also encourage members to discuss and help each other where possible to build a collaborative space and rewarding community.







#### Helpful links



# A bit about you









# Session 1 – BDTI intro, Data Access and Exploration Recap and Exercise review



### What is the Big Data Test Infrastructure (BDTI)?

Not <u>only</u> for big data, for public sector in general (open data)

You have the key ingredients (datasets), we equip you with the best **open-source tool** to create amazing recipes for **public good**.

**Six months free of charge service** for EU public administrations \* Ready-to-use data analytics stack and support Cloud platform based on open-source tools



\* The cost of the pilot project must fit within the funding boundaries of the BDTI pilot budget



#### Open-source tools to support your data journey DIGITAL 100% MINIO Data Lake 5. Decision-Making 1. Collection UNIVERSAL SERVER mongoDB. **Database Orchestration** Metabase 4. Visualisation Apache **CO** Superset<sup>®</sup> elasticsearch KNIME Advanced 2. Processing Development Processing 3. Analysis Spark jupyter **Environments** Engines R Studio

European Commission

# Who is the Big Data Test Infrastructure (BDTI) for?



#### **European Public Administrations**

All European Public Administrations at local, regional and national level can independently apply for a BDTI pilot project



#### Ecosystem with academia and private sector

Academia, spin-off, startups can apply for pilot projects once there is a **clear collaboration** with a Public Administration which will be the main point of contact for the project (**Master/PhD, GovTech startups**)



Are you working for a public administration in need of infrastructure for data analytics?

https://big-data-test-infrastructure.ec.europa.eu/



# Use case: "Identify Innovative Green Energy Projects from EU Universities"

Zoi and her team, work in the Educational Department of a highly polluted EU region. Their **purpose** is to identify potential *"green energy partnerships"* between EU universities to address pollution issues. Zoi reached out the BDTI team to learn how to **connect the dots**, use the free BDTI playground to **create** *insights* and produce a *report*, which can be **easily updated** when new data is available.





They found some interesting **open data:** 

- List of innovation green energy projects across EU countries
- CO2 emissions of these countries



The Data Analysis Process is like cooking a dish



Business challenge/question/problem statement

### Ingredients + tools

- 1. Find the data you need  $\rightarrow$  gather the right ingredients, good quality
- 2. Get, clean and prepare your data  $\rightarrow$  slice and dice
- 3. Analyse your data  $\rightarrow$  mix ingredients together and try different combinations
- 4. **Present** the results and **create** knowledge  $\rightarrow$  serve and consume

### Data $\rightarrow$ Information $\rightarrow$ Knowledge













### Recap: Zoi Data Journey



21

BDTI - Big Data Test Infrastructure / BDTI Essentials Course

### https://code.europa.eu/bdti/bdti-essentials-course

B BDTI Essentials Course €	Ω ∽ ★ Unstar 1 ¥ Fork 0 5								
Update README.md     Maria Claudia BODINO authored 18 hours ago		85da3666 D							
master v bdti-essentials-course / + v		History Find file Edit v Code v							
README	icense 🕲	Auto DevOps enabled   Add CHANGELOG  Add CONTRIBUTING  Add Kubernetes cluster							
Add Wiki     Configure Integrations									
Name	Last	Update README.md     Maria Claudia BODINO authored 18 hours ago							
C Session 1: Data Access and Exploration	Updi								
Session 2: Data Cleaning and Transformati	Upda	E code owners insign users and groups as approvers for specific the changes, client more.							
15 img	Upto	naster - bdtl-essentials-course / Session 1: Data Access and Exploration / + - Lock History Find file							

Add

Upda



LICENSE

+++ README.md

Maria Claudia BODINO authored 18 hours ago						100
As Code owners. Assign users and groups as approvers for specific file change	ss. Learn more.			1	Aanage bra	inch rules
master - Ddtl-essentials-course / Session 1: Data Access and Exploration	1 + -	Lock	History	Find file	Edit ~	Code -
Name	Last commit				L	ast update
-						
Ding	Delete session1_4.png		1 week ago			
🔹 gitkeep	Add new directory for session 1		2 weeks ago			
BDTLEssentials_Session1_DataAccess_Exploration.pdf	Upload New File		1 week ago			
Exercise_\$1.knwf	Upload New File		1 week ago			
*** README.md	Update README.md		18	hours ago		
@ Session1.R	Upload New File		1 week ago			
C SessionT.jpynb	Upload New File		1 week ago			
C Solution_S1.knwf	Upload New File				3	l week ago

SSTATION D

#### Learning Resources

In this section you can find some open and online learning resources related to the tools we are going to use in the different sessions

#### Jupyter Notebooks

Phyton Data science Handbook

#### R Studio

R for Data Science

#### KNIME

- Extensive Resources for Learning KNIME
- Online self-paced courses
- Explore KNIME Hub for examples
- · Documentation to keep your team up to speed with best practices for data science
- KNIME free books
- Ask for help on KNIME Forum
- Videos on KNIMETV

#### Statistical learning

· An Introduction to Statistical learning in Python and R







# **Exercise Review**



# **BDTI access refresh**

- To have access, public administrations need to *apply on the official website*
- <u>https://big-data-test-</u> infrastructure.ec.europa.eu/apply-bdti\_en
- Once you have access follow these steps:
  - □ Step 1 > Access the portal via the website
  - Step 2 > Navigate to the "Service Catalog" and launch the tools
  - Step 3 > Go to "My Services" and open the tool you created



More share options

#### **BDTI Portal**

The BDTI portal is a web application which allows users to easily deploy and manage containerized data science workloads. In this section, you can access the portal and find documentation about the portal.

#### Access the BDTI Portal

Disclaimer: The BDTI portal is only available to users who have a BDTI pilot.



Big Data Test Infrastructure (BDTI)



How to import an existing KNIME workflow?

- Once you are in your working space click on "import workflow"
  - o .knwf = 'workflow'
  - .knar = 'archive'
- Remember to reconfigure when needed!







### **Exercise review**

#### Session 1 - Solution

#### **Data Access and Exploration**

Zoi downloads the data from <u>data reasona as</u> and from Dur World in Data. Your Test task is to start building the workflow by accessing the different data sources.

Now that the data access is established, Zoi needs to have a look of the data itself and decide what information is useful to complete her task.

#### Access the Data from File(s) and explore the data

1. Read Organization.cov with the CSV Reader node

2. Read Project joon with JSON Reader node and parse it to a table format with JSON Path and Ungroup nodes

- 3. Read The European Science Vocabulary (EuroSciVisc) euroSciVisc xisx with the Excel Reader node
- 4. Read the CD2 dataset with the CSV Reader node

5. Read the official European Names data with the Eacel Reader node









### Session 2: Data Cleaning and Transformation

# Session 2: Data Cleaning and Transformation



Session	Торіс	Duration
Webinar 1	Data Access and Exploration	45 min
Webinar 2	Data Cleaning and Transformation	75 min
Webinar 3	Data Blending and Storage	75 min
Webinar 4	Analytics	75 min
Webinar 5	Advanced Module: Gathering Data from the Web and Geo Visualisation	75 min

- Slides, hands-on exercises, and solutions will be provided on the webinar
- Gitlab repository <u>https://code.europa.eu/bdti/bdti-essentials-course</u>



# Recap of objectives

Objective	Level	
	EU Countries	
Visualise total Energy projects	Organisations in EU Countries	
Identify and Visualise CO2 emissions	EU Countries	
Visualise energy projects per year	EU Countries	
Explore the Relationship between CO2 emissions and Energy Project Funding	EU Countries	
Map Visualisation of funded projects	EU Universities	European

19

DIGITAL EUROPE PROGRAMME

# Outputs according to Zoi goals



### The report/dashboard needs to be informative!

- Insightful plots
- Tables with a rank of countries according to CO2 emissions
- Bar chart with CO2 emissions
- Line plot with both Horizon 2020 and 2021
- Locate on a map the Universities that got funding for Energy projects from Horizon



### **Example Dashboard Report with BDTI Framework**





### At the end of this session, you should be able to:

- Sign in into the BDTI framework and initialise KNIME.

- Prepare the data for analysis by **cleaning** the dataset.

- Prepare the data for subsequent analysis by **transforming** the dataset to the correct format.

- Address data quality issues.





# Session 2: Data Cleaning and Transformation

- The core of the data pipeline
- Why clean and transform?
  - Data are **rarely** clean
  - Remove not useful data
  - Remove repeated data
  - Handle missing values
- Extract more information
  - Define rules and mathematical operations
  - Transform at cell or row level
- Transform the data to the desired shape
  - Rename and resort table columns
  - Split and merge columns
  - Convert data types



#### **Use Case**

With an overview of the data, Zoi realises that it is not in the best shape to do an analysis, some columns are in the wrong format and she is also missing data.

In this session, she will learn techniques to clean and transform the data.

Data Access

Data Cleaning and Transformation Data Blending and Export

Analytics: aggregation, visualisation, reporting

Advanced Analytics







### Why?

- "Dirty data"
  - □ Unnecessary rows or columns

Item	Cost	Store Type	City	VAT	Delivery code
Table	Table50Couch100Couch100		NY	14%	XY
Couch			NY	14%	XX
Couch			NY	14%	XX
Cancelled	?	?	?	?	?
TV	70	Online	NY	14%	XY
Chair30Table?		Onsite	NY	14%	XY
		Online	NY	14%	XX





### Why?

- "Dirty data"
  - □ Unnecessary rows or columns
  - □ Missing or null values

Item	Cost	Store Type	City	VAT	Delivery code
Table	50	Onsite	NY	14%	XY
Couch	100	Online	NY	14%	XX
Couch	100	Online	NY	14%	XX
Cancelled	?	?	?	?	?
TV	70	Online	NY	14%	XY
Chair	30	Onsite	NY	14%	XY
Table	?	Online	NY	14%	XX





### Why?

- "Dirty data"
  - □ Unnecessary rows or columns
  - □ Missing or null values
  - Duplicates

Item	Cost	Store Type	City	VAT	Delivery code
Table	50	Onsite	NY	14%	XY
Couch 100		Online	N-Y	14%	XX
Couch	100	Online	N-Y	14%	XX
Cancelled	?	?	?	?	?
TV	70	Online	NY	14%	XY
Chair	30	Onsite	NY	14%	XY
Table	?	Online	NY	14%	XX





### Why?

• Keep only what is essential

Item	Cost	Store Type	City
Table	50	Onsite	NY
Couch	100	Online	NY
Couch	100	Online	NY
TV	70	Online	NY
Chair	30	Onsite	NY
Table	50	Online	NY









### Step 1: Explore data

Step 2: Identify which columns / rows you need and filter or sort

Step 3: Identify if there are missing values or duplicates you need to deal with





- Look at raw data (R-studio, Jupyter Notebook, KNIME)
- Review statics view (KNIME)

Step 1 - Explore Data

						File Table	(Statistics)						
Rows: 25   Col	lumita: 14												
Name	Туре	# Missing val	# Unique valu	Moleum	Maximum	25% Quantile	50% Quantile _	75% Quantile	Mean	Mean Absolut	Standard Dec.,	Sun	10 most cert
projectED	Number (inte	0	35385	115,797	101,051,997	717,171.5	786,731	871,260	11,555,105.079	19,242,039,454	31,055,440 111	2,054,890,555.	785,219 (210
projectAcrony.	String	0	32973	0	0	0	0	0	0	0	.0	0	GripheneCo
organisationiD	Number (inte	0	41534	882,147,877	999,997,930	949,799,169	998,801,629	999,902,773	974,468,313.5.	30,978,751.754	35,311,978,981	173,293,601,6	999,997,930
vatNumber	String	17946	34924	0	0	0	0	0	۲	0	0	0	FR40180089
name	String	0	41463	0.	0	6	0	0	0	0	0	0	CENTRE NAT
shortName	Going	37684	28911	0	0	0	0	0	۲	0	0	0	CNRS (1883)
SME	Daring	521	2	0	0	0	0		0	0	0	0	false (14235
activityType	String	1	5	0	0	0	0	0	0	0	0	0	PRC (59961)
street	String	1199	38227	0	0	0	0	0	0	0	0	0	ILLE MICHEL
postCode	String	1962	18971	0	0	0	0	0	0	0	0	0	1000 (2480)
city	Daving	923	13215	0	•	8	0	0	0	0	0	0	Parts (4548)
country	String	1	177	0	0	0	0	0	0	0	0	0	DE (20782, 1
nutsCode	Saving	177834	0	0	0	0	0	0	0	0	0	0	0
geolocation	taring	251	26044	0	0	0	0	0	0	0	0	0	48.84755265
arganizationU	String	62236	8745	0	0	0	0	0	0	0	0	0	HID./WWW.D
contactForm	String	0	177456	0		0	0	0	0	0	ė.	0	Intest/ec.ev
contentilpdat.	String	0	31103	0	0	0	0	0	0	0	C	0	2023-02-28 1
107	Number Gote.	0	41534	1,905,548	3,146,064	1,906,583	1,919,366	2,002,179	2,059,448.994	204,881.698	282,765.939	346,240,052,4	1,905,672 (1)
order	Number (inte.	0	261	1	1.001	2		12	14.535	15.401	74.853	2,584,823	1 (37472; 21
role .	String	0	5	0	0	0	0	0	0	0	0	0	participant ()
ecContribution	Number (dou	7748	80653		178,149,171	80,343.75	212,933.76	440,743.75	401,552.969	363,000.334	1,075,846,727	68,298,538,22	0 (19523; 11
setEcContribu.	Number (dou	48	86062		178,169,171	76,110.312	202,085.61	418,198,438	384,161.079	348,387,687	1,009,162,309	68,298,461,53.	0 (16824, 9.4
totalCost	String	191	83929	0	0	0	0	0	0	0	0	0	0 (12125; 8.8
andOfParticip.	String	0	2	0	0	0	0	0	0	0	0	0	false (17095
active	Shing	176225	1	0	0	0	0	0	0	0	0	0	false (1606)







Step 1: Explore data

### Step 2: Identify which columns / rows you need, and filter or sort

Step 3: Identify if there are missing values or duplicates you need to deal with



### Step 2 - Identify columns

 Horizon Data "Organisations"- Visualise Energy projects per EU Country and for Organisations per Country

(	Contraction of the			- 1000000000								
RowID	projectID Number (inte.	projectAc	organisat	vatNumber String	name 🥪	shortName String	SME String	setivityTy	street String	postCode	city String	country Stong
Rowt	879926	EEN SACHSEN	999913831	0	INDUSTRIE-U	IHK C	false	PUB	STRASSE DE_	09111	CHEMNITZ	DE
Row	879926	EEN SACHSEN	994556036	DE176093942	ZTS-ZENTRU_	ZTS	true	отн	INDUSTRIEST_	01612	GLAUBITZ	DE
Row?	879926	EEN SACHSEN	999797625	•	INDUSTINE-U.,	IHK DRESDEN	faise	PUB	LANGER WEG.	01239	DRESDEN	DE
Row3	879926	EEN SACHSEN	999940409	0	TT PTT	HWK DRESDE	false	PUB	AM LAGERPL.	0.2-	DRESDEN	DE
Rowl	879926	EEN SACHSEN	999877844	DE140857609	TI CI NIS CHE.	0	false	HES	STRASSE DE		Chemnitz	DE
Rows	879926	EEN SACHSEN	999914219	0	INDUSTRIE-U_	IIHK-L	false	PUB	GOERDELERR.	04109	LEIPZIG	DE
Rowb	879926	EEN SACHSEN	969246408	0	HOCHSCHUL	Hochschule Z	false	HES	THEODOR-KO_	02763	Zittau	DE
Row?	879926	EEN SACHSEN	999917905	DE141484057	AGENTUR FU	AGIL	false	PRC	LESSINGSTR	04109	LEIPZIG	DE
RowB	740689	HEIMDALL	997710476	IT015032900.	CENTRO INT	FONDAZIONE	false	REC	VIA ARMAND	17100	SAVONA	IT
Row	740689	HEIMDALL	940057847	DK37222135	FREDERIKSB_	FBBR	false	PRC	LOGISMOSE 3	3600	FREDERIKSS	DK
Row.	740689	HEIMDALL	999703438	ESS0811001G	Departament _	INT	false	PUB	Carrer Diputa	08009	Barcelona	ES
Row.	740689	HEIMDALL	919049102	IT136697210.	ASSOCIAZIO_	O	false	отн	VIA BERNAR	00151	Roma	IT
Row.	740689	HEIMDALL	999778322	EL094149709	SPACE HELL	SPACE HELL.	true	PRC	MESSOGION	153 41	Aghia Parask.,	EL.
Row.	740689	HEIMDALL	999638739	ESG62616586	CENTRE TEC	CTTC	false	REC	AVINGUDA C	08860	Castelldefels	ES
Row.	740689	HEIMDALL	996569950	FR44130005	UNIVERSITE	UNISTRA	false	HES	RUE BLAISE P.	67081	Strasbourg	FR
Row.	740689	HEIMDALL	935977542	ESQ0801980D	INSTITUT CA	ICGC	false	PUB	PARC DE MO	08038	Barcelona	ES
Row.	740689	HEIMDALL	997822026	ESB24352296	TECNOSYLVA	0	true	PRC	Parque Tecno.	24009	Leon	ES
Row.	740689	HEIMDALL	999981731	DE121965658	DEUTSCHES	DLR	false	REC	LINDER HOHE	51147	Koln	DE
Row.	740689	HEIMDALL	917614763	GB158535583	SCOTTISH FI_	SCOTTISH FI_	false	PUB	SCOTTISH FI	G72 7NA	CAMBUSLANG	UK
Row.	740689	HEIMDALL	989367991	GB830703457	AVANTI COM	0	true	PRC	COBHAM HO	EC4V 6EB	London	UK
Row.	740689	HEIMDALL	958896217	ESG55536098	FUNDACIO D'	PAU COSTA F	true	REC	AV. MOSSEN _	08552	Taradell	ES





<u>https://data.europa.eu/data/datasets/cordish2020projects?locale=en</u>

• https://data.europa.eu/data/datasets/cordis-eu-research-projects-under-horizon-europe-2021-2027?locale=en

# Step 2 - Identify columns



• Horizon Data "Projects" - Identify Energy **Projects** per Country per year

ids Number (inte	legalBasi	masterCa	natures String	×	objectives 🧹	rcns Number (inte_	startDates String	statuss String	Υ.	subCalls	titl	topicss	totalCosts 4 Number (double)
633053	H2020-Eurato	EURATOM-Ad			A Roadmap t	193159	2014-01-01	SIGNED		EURATOM-Ad	Imple_	EURATOM	1,329,639,211.5
807097	H2020-EU.3.4.	H2020-IBA-C			The challeng	218781	2014-01-09	CLOSED		H2020-IBA-C	Large	CS2-GAM	247.9 6 62.58
807085	H2020-EU.3.4.	H2020-IBA-C			Engines ITD	216640	2014-01-03	CLOSED		H2020-IBA-C	Engine	CS2-GAM	24.10 00.20 21
945583	H2020-EU.3.4.	H2020-IBA-C			Main objectiv	231017	2020-01-01	SIGNED		H2020-IBA-C	Large	CS2-GAM	228,776,991.75
807083	H2020-EU.3.4.	H2020-IBA-C			The Airframe	216645	2014-01-02	CLOSED		H2020-IBA-C	AIRFR_	CS2-GAM	210,183,580.85
853989	H2020-EU.3.1.	H2020-JTI-IM			The European	226660	2020-01-01	SIGNED		H2020-JTI-IM_	EURO	IMI2-2018	207,963,891
101007873	H2020-EU.3.1.	H2020-JTI-IM			Current anti-t	236634	2021-06-01	SIGNED		H2020-JTI-IM	ACAD_	IMI2-2020	185,000,000
777389	H2020-EU.3.1.	H2020-JTI-IM	crisisPrep	are	Paediatric me	218657	2018-05-01	SIGNED		H2020-JTI-IM	conect	IMI2-2016	182,018,216
737417	H2020-EU.2.1	H2020-ECSEL			R3-POWERUP	210524	2017-11-01	SIGNED		H2020-ECSEL	300m	ECSEL-20_	180,318,047.5
681463	H2020-EU.4.e.	H2020-Adhoc			COST is an in	198528	2015-05-01	CLOSED		H2020-Adhoc	COST	COST-Net	178,169,171
662338	H2020-EU.2.1	ECSEL-2014-2			The SeNaTe	198669	2015-04-01	CLOSED		ECSEL-2014-2	Seven	ECSEL-02	177,731,624.33
945542	H2020-EU.3.4.	H2020-IBA-C			The challeng	229352	2020-01-01	SIGNED		H2020-IBA-C	Fast R	CS2-GAM	171,739,896.46
807081	H2020-EU.3.4.	H2020-IBA-C			The Systems	216639	2014-01-01	CLOSED		H2020-IBA-C	Syste	CS2-GAM	158,177,923.26
881603	H2020-EU.1.2.	H2020-SGA-F			This proposal	227202	2020-04-01	SIGNED		H2020-SGA-F	Graph	SGA-FET	150,000,000
945539		H2020-SGA-F_			The last of fo	229984	2020-04-01	SIGNED		H2020-SGA-F	Huma	SGA-FETF	150,000,000
101018100	H2020-EU.3.1.	H2020-FPA-S	crisisResp	onse	Funding of re	232261	2020-07-24	SIGNED		H2020-SGA-S	PROP	SGA-SC1	150,000,000
692522	H2020-EU.2.1	H2020-ECSEL			The TAKE5 pr	203403	2016-04-01	CLOSED		H2020-ECSEL	Techn	ECSEL-15	149,882,181
101036970	H2020-EU.3.3.	H2020-LC-GD			REFHYNE II w	237006	2021-10-01	SIGNED		H2020-LC-GD	Clean	LC-GD-2-2	148,956,405

<u>https://data.europa.eu/data/datasets/cordish2020projects?locale=en</u>

• <u>https://data.europa.eu/data/datasets/cordis-eu-research-projects-under-horizon-europe-2021-2027?locale=en</u>



ommissio

# Step 2 – Identify columns

Horizon Data "EuroSciVoc" - *Identify* **Energy** *Projects per Country per year* <u>https://op.europa.eu/en/web/eu-vocabularies/euroscivoc</u>

 Reference vocabulary for the Open Science community and is aligned with Linked Open Data standards.

RowID	projectID String	~	euroSciVocCode	euroSciVocPath	euroSciVocTitle String
Row0	869855		/29/101/555/1359	/social sciences/sociolog	automation
Row1	869855		/25/63/399	/engineering and technol	textiles
Row2	869855		/25/73/453/58525161	/engineering and technol	sensors
Row3	869855		/25/73/453/459	/engineering and technol	robotics
Row4	869855		/29/93/47293307	/social sciences/psychol	ergonomics
Row5	870148		/25/63/409	/engineering and technol	colors
Row6	883285		/21/33/121/44109686/5	/medical and health scien	ebola
Row7	883285		/21/33/137/133/9678651	/medical and health scien	pandemics
Row8	883285		/21/33/121/44109686/7	/medical and health scien	influenza
Pow0	200005		/20/01/522/1212	(coolial eciences/econom	amalaumant

<u>https://data.europa.eu/data/datasets/cordish2020projects?locale=en</u>

• <u>https://data.europa.eu/data/datasets/cordis-eu-research-projects-under-horizon-europe-2021-2027?locale=en</u>





### Step 2 – Identify columns and rows



European

• CO2 emissions data: *Identify and Visualise* **EU Countries** with the Most and Least **CO2 Emissions** 

	country	year	iso_code	population	gdp	cement_co2	cement_co2_per_capita	cumulative_cement_co2	cumulative_co2
0	Afghanistan	1850	AFG	3752993.0	NaN	NaN	NaN	NaN	NaN
1	Afghanistan	1851	AFG	3767956.0	NaN	NaN	NaN	NaN	NaN
2	Afgle- Set in	1852	AFG	3783940.0	NaN	NaN	NaN	NaN	NaN
3	A av nistan	1853	AFG	3800954.0	NaN	NaN	NaN	NaN	NaN
4	Afghanistan	1854	AFG	3818038.0	NaN	NaN	NaN	NaN	NaN
		V						(C11	
50593	Zimbabwe	2017	ZWE	14751101.0	2.194784e+10	0.469	0.032	18.479	753.403
50594	Zimbabwe	2018	ZWE	15052191.0	2.271535e+10	0.558	0.037	19.037	765.199
50595	Zimbabwe	2019	ZWE	15354606.0	NaN	0.570	0.037	19.606	776.313
50596	Zimbabwe	2020	ZWE	15669663.0	NaN	0.570	0.036	20.176	786.921
50597	Zimbabwe	2021	ZWE	15993525.0	NaN	0.570	0.036	20.745	798.217

<u>https://ourworldindata.org/co2-emissions</u>



### Step 2 – Identify **rows**

• EU Country Names data: Filter EU countries (rows)

	RowID	AREA String		~	CODE String	. ~	String	
29	Row28	European Union (E	U)			UK		United Kingdom
30	Row29	European Free Trac	ie Asse	ociation (EFTA)		IS		Iceland
31	Row30	Eun 🖌 🔹 ,		tion (EFTA)		u		Liechtenstein
32	Row31	Eun 🏅	*	tion (EFTA)		NO		Norway
33	Row32	Eun *	.*	ition (EFTA)		СН		Switzerland
34	Row33	EU				TR		Turkey
35	Row34	European Neighbo	urhood	Policy (ENP)-South		IL.		Israel
36	Row35	Other European co	untries			RU		Russia
37	Row36	low36 Non-European countries			AU		Australia	
38	Row37	Non-European cou	ntries			BR		Brazil

https://ec.europa.eu/eurostat/statistics-explained/images/9/9f/Country\_Codes\_and\_Names.xlsx









Step 2: Identify which columns/ rows you need, and filter or sort

# Step 3: Identify if there are missing values or duplicates you need to deal with





### Step 3 - Missing Values

- Define how to handle missing values for all columns of a given type
  - Affect all columns that are not explicitly mentioned in the second tab
- Define how to handle missing values for each available column



EEN TOPIC2I	919796390	FR111300216
EEN TOPIC2I	952860392	$\bigcirc$
EEN TOPIC2I	939950468	0
EEN TOPIC2I	938560167	0
EEN TOPIC2I	999779971	FR211845001
INNO DK	902853206	DK40084711
INNO DK	901400437	DK40084606
INNO DK	902860578	DK40084746
EEN OUEST H	950310359	FR555136825
EEN OUEST H	950310359	FR555136825
EEN Northern	996747848	NL80144121
EEN-Ukraine	933428673	UA36368053
EEN-Ukraine	938295066	0
EEN-Ukraine	927605666	0
EEN-Ukraine	899486433	0
MEDAINNO	015515586	ED951200224



### Step 3 - Duplicates values

- Detect duplicate rows and apply a selected treatment
  - Select columns to check for duplicates
  - Provide options for treating duplicated values





### Data Cleaning: Recap

Step 1: Explore data

Step 2: Filter or sort

### Step 3: Missing values and duplicates

ltem	Cost	Store Type	City	VAT	Delivery code
Table	50	Onsite	NY	14%	ХҮ
Couch	100	Online	N-Y	14%	XX
Couch	100	Online	N-Y	14%	XX
Cancelled	?	?	?	?	?
TV	70	Online	NY	14%	ХҮ
Chair	30	Onsite	NY	14%	ХҮ
Table	?	Online	NY	14%	XX

	ltem	Cost	Store Type	City
	Table	50	Onsite	NY
	Couch	100	Online	NY
	Couch	100	Online	NY
	TV	70	Online	NY
	Chair	30	Onsite	NY
	Table	50	Online	NY



European Commission





- What do we transform?
  - Shape of the table (new columns)
  - Content of cells

ids Number (integer)	<ul> <li>startDates</li> <li>Local Date</li> </ul>	endDates Local Date	startYear	ų.
879926	2020-01-01	2021-12-31	2020	
740689	2017-05-01	2021-01-31	2020	
730082	2016-11-01	2019-04-30	. 2017	
101023342	2021-06-01	2024-11-30	2016	
831993	2019-01-01	2021-12-31	2021	
641451	2015-01-01	2016-12-31	2019	
731665	2017-01-01	2018-12-31	2015	
653866	2015-06-01	2017-11-30	2013	
770037	2017-11-01	2019-10-31	2017	
636906	2015-02-01	2017-01-31	2015	
825132	2019-01-01	2022-10-31	2017	
780785	2018-01-01	2021-12-31	2015	
19	2020-11-01	5029.10.31	2019	

RowID Country String	RowID	country String
Ro' Cz	Ro	Cz
Row. Czechia	Row	Czech Republic
Row. Czechia	Row	Czech Republic
Row. Czechia	Row	Czech Republic
		Court Description



0

2018

### Why transform data?

- Columns in mismatching data type that can lead to problems
- What are data types?
  1.String: Text or characters.
  - 2. Number: Numerical values.
  - 3. Date & Time: Points in time or durations.
- Horizon 2020 column "ID" = number
- Horizon 2021 column "ID" = string



RowID	projectID Number (integer)	RowID	projectID String	~
Row0	879926	Row0	101091483	
Row1	879926	Row1	101091483	
Row2	879926	Row2	101091483	
Row3	879926	Row3	101091483	
Row4	879926	Row4	101091483	
Row5	879926	Row5	101091483	
Row6	879926	Row6	101091483	



- Why transform data?
  - Need to extract information that is missing
    - We want the Start Year of the projects
    - "startDate" column, but it comes as a string
    - Need to transform to a date-time format and the extract year!

startDates String	<pre>endDates String</pre>	startYear String	~ 7
2020-01-01	2021-12-31	2020	
2017-05-01	2021-01-31	2017	
2016-11-01	2019-04-30	2016	
2021-06-01	2024-11-30	2021	
2019-01-01	2021-12-31	2019	
2015-01-01	2016-12-31	2015	
2017-01-01	2018-12-31	2017	
2015-06-01	2017-11-30		





- Why transform data?
  - One project can have multiple topic name, so we need to **group** the different topics per project

RowID	projectID String	euroSciVocTitle String
Row175	852096	renewable energy
Row177	852096	energy conversion
Row232	951513	renewable energy
Row248	759603	energy conversion
Row314	742829	energy and fuels
Row536	790458	energy and fuels
Row686	966685	solar energy
Row800	864660	energy conversion
Row811	724610	energy conversion

RowID	projectID ~~ String	projectsTopics String
RowID	852096	projectsTopics
Row2001	852096	renewable energy, energy conversion





- Why transform data?
  - One project can have multiple topic name, so we need to **group** the different topics per project

DucientID	<b>T</b> :+1 = =		Project ID	Concatenate(Titles)
 Project ID	lities		1	Renewable energy, Energy
1	Renewable energy		2	Solar energy
2	solar energy		3	Solar energy, energy conversion
3	Solar energy		4	Renewable energy
3	Energy conversion			
4	Renewable energy	Aggregated on Project ID(g Concatenation (aggregatior	group) by n method)	



European Commission

- Why transform data?
  - Some country names are incorrect and can also lead to problems later when merging with all the date
    - Transform string and get them in the correct format

AREA String	$\sim$	CODE String	$\sim$	COUNTRY NAME String	RowID	AREA String	$\sim$	CODE String	$\sim$	COUNTRY NAM
European Union (EU)		BE		Belgium	Row1	European Union (EU)		BE		Belgium
European Union (EU)		BG		Bulgaria	Row2	European Union (EU)		BG		Bulgaria
European Union (EU)		CZ		Czech Republic	Row3	European Union (EU)		CZ		Czech Republic
European Union (EU)		DK		Denmark	Row4	European Union (EU)		DK		Denmark
European Union (EU)		DE		Germany (including former GDR from 1991)	Row5	European Union (EU)		DE		Germany
European Union (EU)		88		Estonia	Row6	European Union (EU)		EE		Estonia
European Union (EU)		IE		Ireland	Row7	European Union (EU)		IE		Ireland
European Union (EU)		EL		Greece	Row8	European Union (EU)		EL		Greece
European Union (EU)		ES		Spain	Row9	European Union (EU)		ES		Spain
European Union (ELI)		CD 03		Eranon						



European







### Data Transformation: what and why?



### From raw data



# To transformed data in a format tailored to Zoi's needs

RowID	projectID String	euroSciVocCode	euroSciVocPath
Row0	869855	/29/101/555/1359	/social sciences/sociolo
Row1	869855	/25/63/399	/engineering and technol
Row2	869855	/25/73/453/58525161	/engineering and technol
Row3	869855	/25/73/453/459	/engineering and technol
Row4	869855	/29/93/47293307	/social sciences/psychol
Row5	870148	/25/63/409	/engineering and technol
Row6	883285	/21/33/121/44109686/5	/medical and health scien
Row7	883285	/21/33/137/133/9678651	/medical and health scien
Row8	883285	/21/33/121/44109686/7	/medical and health scie
David	000005	100/01/500/1010	Innelal anionana lananam

1	euroSciVocTitle String
g	automation
١	textiles
١.	sensors
١	robotics
۱.,	ergonomics
ι.	colors
n	ebola
n	pandemics
n	influenza
	amalaumant

RowID	string	~	projectsTopics String
Row0	101039110		energy conversion
Row1	101039446		energy and fuels
Row2	101039576		renewable energy
Row3	101039746		solar energy, energy conversion
Row4	101040341		geothermal energy
Row5	101040379		energy conversion
Row6	101040669		energy conversion
Row7	101040994		energy conversion
Row8	101041768		solar energy
Row9	101041809		solar energy
Row10	101042781		hydrogen energy
Row11	101043617		energy conversion
Row12	01043969		hydrogen energy

### • ID Column in Horizon 2020 to the correct data format

- We extracted the Start Year of projects
- Grouped multiple titles into a single row for each project
- Gave Countries the correct wording









Number to String

### A bit more cleaning

### Sometimes we need to clean again after doing some transformations!

country String	<ul> <li>year Number (megar)</li> </ul>	<ul> <li>iso_code</li> <li>strep</li> </ul>	<ul> <li>cumulative_co2</li> <li>Number (double)</li> </ul>
Afghanistan	1850	AFG	0
Afghanistan	1851	AFG	0
Afghanistan	1852	AFG	Ø
Afghanistan	1853	AFG	۲
Afghanistan	1854	AFG	0
Afghanistan	1855	AFG	٢
Afghanistan	1856	AFG	٢
Afghanistan	1857	AFG	0
	Cle	ean 📕	

RowID	country String	RowID	country String
Ro	Cz	Ro'	Cz
Row	Czechia	Row	Czech Republic
Row	Czechia	Row	Czech Republic
Row	Czechia	Row	Czech Republic
Deres	Orachia	Deres	On the Branchitz

Transform



Clean again!

country String	year Number (integer)	iso_code 🧹	cumulative_co2 Number (double)
Austria	2021	AUT	5,571.7
Belgium	2021	BEL	12,644.15
Bulgaria	2021	BGR.	3,873.285
Croatia	2021	HRV	1,118.099
Cyprus	2021	CYP	301.471
Czech Republic	2021	CZE	12,076.99
Denmark	2021	DNK	4,115.088
Estonia	2021	EST	1,640.459
Finland	2021	FIN	3,226.029
France	2021	FRA	39,105.766
Germany	2021	DEU	93,290.508
Greece	2021	GRC	4,132.681
Hungary	2021	HUN	5,058.661
Ireland	2021	IRL	2,263.349
Italy	2021	ITA	25,065.178 European

Commission

# Today's Output: Workflow

• All these steps are in one single workflow that we can re-run and do it again!







Now, you should be able to:

- Sign in into the BDTI framework and initialise KNIME

- Prepare the data for analysis by **cleaning** the dataset

- Prepare the data for subsequent analysis by **transforming** the dataset into the correct format

- Address data quality issues











# Practice, practice and practice!

 Slides, hands-on exercises, and solutions will be provided on the webinar Gitlab repository

https://code.europa.eu/bdti/bd ti-essentials-course

Maria	Claudia BODINO aufrirezi 4 tays age					900	1.949 0
ater -	bdl-essentials-course / + +		History	Find file	Edit v	à-	Cone
README	19 85D 3-Ctause "New" or "Revised" License	Auto DevOps enabled					
		Last conveit					Last update
Session 1	Data Access and Exploration	Upload banner session 1					4 days ago
BDTI,Ban	ner_generic prig	Upload New File					5 days ago
UCENSE		Add LICENSE					6 days ago
READINE	nd	Lipdate README red					5 stays age
REACHE	nd						
2822.1							
Lagyngre	2023 European Union						
Document	tation in this repository is licensed under the Creat	ove Commons Attribution 4.0 License, and co	de samples als Scensed under	the \$50 3-0	lause licens	ж.	



ommissio

### Practice, practice and practice!



 Slides, hands-on exercises, and solutions will be provided on the webinar Gitlab repository

https://code.europa.eu/bdti/bd ti-essentials-course

#### Session 2

#### **Data Cleaning**

With an overview of the data, Zoi asks your help in cleaning the data, since there seems to be duplicates, missing values, and some columns that are not needed.

#### Clean the data

#### Clean the Organizations, Projects and euroSc/Voc data

Select columns with the Column Filter node.
 In Organisations 2021, filter the row 645002 using the Row Filter node.
 Use the Column Researcer node to set the columns in the order you want.
 In EuroSchlod select only energy projects using the Row Filter node.

#### **Data Transformation**

Next, Zoi realises that some columns are in an incorrect data format, so she needs to transform them to be able to use the data in the correct

#### wey.

#### Transform the data

1. Transform ID column in Horizon2020 to string with the Number to String node

2. Transform the startDates and endDates columns format with the String to Date&Time node.

- 3. Extract the Start Year with the Extract Date& Time Fields node
- 4. Rename your new column to "startYear" with the Column Resamer node
- 5. Transform startNear and ID to string with the Number to String node
- 6. Concatenate the different project types in euroSciVoc with the GroupBy node
- 7. Rename this new column with the Column Renamer node







### Your BDTI journey starts here



Congrats! You are on the first step to data-driven innovation.





Brainstorm your data project



Apply for BDTI Pilotlight process



### **Course discussion board**





Here, you can ask questions and discuss topics related to the BDTI Essentials Online Course. The board is moderated by the BDTI team, so your questions will be answered quickly. We also encourage members to discuss and help each other where possible to build a collaborative space and rewarding community.







#### Helpful links

# Get in touch and follow the BDTI activities

Are you working for a public administration in need of infrastructure for data analytics?





Visit BDTI's website



Subscribe to BDTI's newsletter



Subscribe to BDTI's Joinup



Register for BDTI Essentials online course

ommission

https://big-data-test-infrastructure.ec.europa.eu/