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SCUOLA DI INGEGNERIA INDUSTRIALE
E DELL'INFORMAZIONE

Systems and Methods for Big and Unstructured Data Project

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Contents

Contents	i
I Delivery	3
1 Import	3
1.1 Libraries used	3
1.2 Publications	3
1.3 Venues	5
1.4 Fields of study and relation N-M	5
1.5 Authors	6
2 Queries with Spark	7
2.1 Data creation/update	7
2.1.1 Query 1	7
2.1.2 Query 2	7
2.1.3 Query 3	8
2.1.4 Query 4	9
2.1.5 Query 5	10
2.2 Queries	11
2.2.1 WHERE, JOIN	11
2.2.2 WHERE, LIMIT, LIKE	11
2.2.3 WHERE, IN, Nested Query	12
2.2.4 GROUP BY, 1 JOIN, AS	13
2.2.5 WHERE, GROUP BY	14
2.2.6 GROUP BY, HAVING, AS	14
2.2.7 WHERE, GROUP BY, HAVING, AS	15
2.2.8 WHERE, Nested Query (i.e., 2-step Queries), GROUP BY	16
2.2.9 WHERE, GROUP BY, HAVING, 1 JOIN	17
2.2.10 WHERE, GROUP BY, HAVING, 2 JOINS	18

Part I

Delivery 3

1 | Import

For the import phase, we created five different json files using the same principle present in the MongoDB import. Through the json library, we opened the raw dataset file, we extracted the relevant information and finally we wrote them on the interested file.

The created files are:

- **publications.json** containing all the publications;
- **venues.json** with all the venues;
- **fos.json** for the fields of study;
- **authors.json** containing all the attributes of the authors;
- **rel_dw.json** which represents the N-M relationship between the fields of study and the publications;

1.1. Libraries used

To work on the raw dataset, in addition to those described in the MongoDB Import section, we added these libraries:

- **hashlib** to generate an id for the fields of study;
- **re** which permits to modify the output files in order to have a well-formatted json file.

1.2. Publications

For the publications, we take the information already present in the raw dataset, like the id or the publisher, and we created the ones which are not.

One of the tricky cases we had, was for the venue id. As a matter of fact, some venues have not the id field, so the script provides one, using a random number between 4.000.000 and 9.999.999. We opted for this range, because plotting all the ids, they are all below the

lower threshold we fixed (Figure 1.1). In this way, the probability of having two identical ids created through the *random()* method is very low.

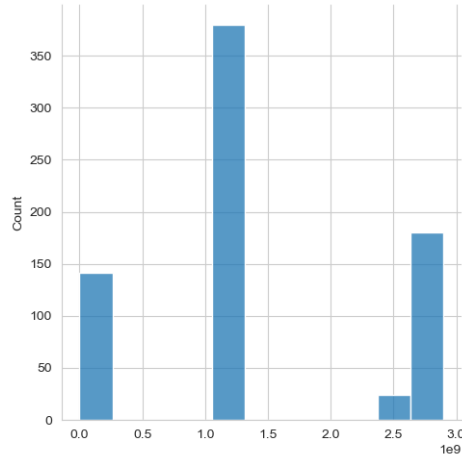


Figure 1.1: Plot of the ids

```

1 # Publisher part (authors[] filled in author part)
2 pub['id'] = data[i]['id']
3 pub['pages'] = random.randint(1, 20)
4 pub['abstract'] = "Abstract of " + \
5     data[i]['title'] + ": " + lorem.words(5)
6 pub['title'] = data[i]['title']
7 if data[i]['publisher'] == "":
8     data[i]['publisher'] = "PoliPrint, Milano"
9 pub['publisher'] = data[i]['publisher']
10 pub['year'] = data[i]['year']
11 if 'id' in data[i]['venue']:
12     pub['venue'] = data[i]['venue']['id']
13 else:
14     data[i]['venue']['id'] = random.randint(4_000_000_000, 9_999_999_999
15 )
16     pub['venue'] = data[i]['venue']['id']
17 pub['references'] = []
18 for j in range(random.randint(1, 6)):
19     number = vect_id[random.randint(0, n_doc-1)]
20     while number in pub['references']:
21         number = vect_id[random.randint(0, n_doc-1)]
22     pub["references"].append(number)

```


1.3. Venues

The venue part has the same methods viewed before. Notice that before performing the operations, we check if the venue was already present in the one inserted before.

This problem may arise because some publications can derive from the same venue, and if we did not check this, the venue would be inserted many times, creating replicates. This way of solving this issue will be used also for the *authors* and for the *fields of study*.

```

1 # Venue part
2 if data[i]['venue']['id'] not in vect_ven_ids:
3     vect_ven_ids.append(data[i]['venue']['id'])
4     ven['id'] = data[i]['venue']['id']
5     ven['raw'] = data[i]['venue']['raw']
6     ven['volume'] = random.randint(1, 3)
7     ven['number'] = random.randint(1, 20)
8     if 'type' in data[i]['venue']:
9         ven['type'] = data[i]['venue']['type']
10    else:
11        data[i]['venue']['type'] = 'J'
12        ven['type'] = data[i]['venue']['type']
13    ven['date'] = date.fromordinal(random.randint(date(day=1, month=1,
14        year=data[i]['year']).toordinal(
15        ), date(day=31, month=12, year=data[i]['year']).toordinal())).
16    isoformat()

```

1.4. Fields of study and relation N-M

In this part of the script we cope with the issue of the *id* field. To avoid that two fields of study have the same id, we used the *md5* hash method, which starting from a seed, which is the name of the fos (that is unique), generates an alphanumeric string.

```

1 # Relation Deals_With + Field Of Study (FOS) part
2 for f_o_s in data[i]['fos']:
3     obj = {}
4     obj['pub_id'] = data[i]['id']
5     obj['fos_id'] = hashlib.md5(f_o_s['name'].encode()).hexdigest()
6     obj['weight'] = f_o_s['w']
7     rel_dw.append(obj)
8     if f_o_s['name'] not in vect_fos_ids:
9         obj = {}
10        vect_fos_ids.append(f_o_s['name'])
11        obj['id'] = hashlib.md5(f_o_s['name'].encode()).hexdigest()
12        obj['name'] = f_o_s['name']

```

```
13         fos.append(obj)
```

1.5. Authors

This part has the same structure of the previous part and the same methods presented in the import of MongoDB.

Notice that the last line of the following code presents also the creation of the array of authors of the publication

```
1 # Author part
2 pub['authors'] = []
3 for author in data[i]['authors']:
4     if author['id'] not in vect_aut_ids:
5         obj = {}
6         vect_aut_ids.append(author['id'])
7         obj['id'] = author['id']
8         obj['name'] = author['name']
9         obj['email'] = author['name'].split()[0] + '.' + \
10             author['name'].split()[len(
11                 author['name'].split())-1] + '@mail.com'
12         obj['bio'] = 'Bio of ' + author['name'] + ': ' + lorem.words(5)
13         if 'org' in author.keys():
14             obj['affiliation'] = author['org']
15         else:
16             obj['affiliation'] = vect_org[random.randint(
17                 0, len(vect_org) - 1)]
18         aut.append(obj)
19     pub['authors'].append(author['id'])
```

2 | Queries with Spark

2.1. Data creation/update

Here are the five data creation/update commands.

2.1.1. Query 1

```
1 columns = ["date", "id", "number", "raw", "type", "volume"]
2 newRow = spark.createDataFrame([("2022-12-05", 9999999, 99, "New venue
   raw", "C", 3)], columns)
3 venue_df = venue_df.union(newRow)
4 venue_df.show(truncate=False)
```

Listing 2.1: Creation of a new row

2.1.2. Query 2

```
1 publication_df.withColumn("references number", size(col("references
   "))).show(truncate=False)
```

Listing 2.2: Creation of a new column that counts the number of references

abstract	authors	id	pages	publisher	references	title	venue	year	references number
Abstract of Preli...	[2312688602, 2482...	1091	9	Springer, Berlin,...	[1304089, 819435,...	Preliminary Desig...	1127419992	2013	6
Abstract of Furth...	[2718958994]	1388	2	PoliPrint, Milano	[316117]	Further Results o...	73158690	2000	1
Abstract of A met...	[2103626414, 2117...	1674	15	Eurographics Asso...	[1099225, 1838952...	A methodology for...	2754954274	2011	5
Abstract of Compa...	[2300589394, 2308...	1688	17	Springer, Berlin,...	[618979, 122273, ...	Comparison of GAR...	1136274694	2009	5
Abstract of COMPA...	[2125293936, 2101...	5411	6	PoliPrint, Milano	[1588621, 2228599...	COMPARING GNG3D A...	1136212596	2009	5
Abstract of Vecto...	[1237859792, 2208...	5781	17	PoliPrint, Milano	[2122821, 348325, ...	Vectorial fast co...	2764847869	2004	6
Abstract of Impro...	[2022192081, 2023...	6522	7	Springer, London	[870555, 217368, ...	Improved Secret I...	1125967516	2011	5
Abstract of A Sel...	[2142249029, 2113...	6762	12	PoliPrint, Milano	[1482794, 2176252...	A Self-Stabilizin...	1196153040	2003	6
Abstract of Forma...	[2611851107]	8373	13	Springer, Berlin,...	[496663, 15883, 1...	Formal agent-orie...	1123338449	2012	4
Abstract of Fur V...	[2156900172, 2281...	8763	9	Springer, Cham	[1487729, 359688, ...	Fur Visualisation...	1196868077	2014	5
Abstract of Ident...	[2563642081, 2561...	9415	20	Springer, Cham	[450079, 762797]	Identifying Psych...	2755612976	2013	2
Abstract of Multi...	[2307482452, 2832...	11068	18	Springer, Berlin,...	[305562, 137649, ...	Multisymplectic S...	2706111989	2002	3
Abstract of The R...	[1251725090]	11796	15	PoliPrint, Milano	[1792135, 1055523]	The Role of the B...	1171805742	2006	2
Abstract of Speec...	[2163873308, 1971...	11895	8	Morgan Kaufmann P...	[778541]	Speech training s...	1203999783	1979	1
Abstract of Softw...	[1978340988, 1986...	12993	6	PoliPrint, Milano	[359688, 1364731, ...	Software Evolutio...	50368787	2003	6
Abstract of Knowl...	[218416969, 81737...	13070	14	IOS Press	[344196, 746710, ...	Knowledge Enginee...	1153467564	2008	3
Abstract of Desig...	[2404438944, 2656...	13205	3	PoliPrint, Milano	[274954, 1538985, ...	Design of an audi...	1177287137	2002	4
Abstract of A Pla...	[2051773316, 2506...	13407	17	Springer, Berlin,...	[601863, 986644, ...	A Platform for Di...	2755952065	2013	6
Abstract of A COM...	[192576500, 20750...	14870	4	PoliPrint, Milano	[2192057]	A COMPUTATIONAL S...	1198225011	2009	1
Abstract of Clean...	[2318310288, 2778...	15548	13	PoliPrint, Milano	[929837]	Cleaneval: a Comp...	1164963593	2008	1

only showing top 20 rows

Figure 2.1: Result of query 9.1.2

2.1.3. Query 3

Creation of a new column that represents the region of the university based on the vectors vectASIA, vectEU, vectAMERICA that we manually initialized in the python file

```

1 author_df = author_df.withColumn('Continent',
2     when(author_df.affiliation.isin(vectASIA), lit("Asia"))\
3     .when(author_df.affiliation.isin(vectAMERICA), lit("America"
4     ))\
5     .when(author_df.affiliation.isin(vectEU), lit("Europa"))\
6     .otherwise(lit("Rest of the world")))\
7     .show(truncate=False)
8

```

affiliation	bio	email	id	name	Continent
Shinshu University	Bio of Makoto Sat...	Makoto.Satoh@mail...	2312688602	Makoto Satoh	Asia
Shinshu University	Bio of Ryo Murama...	Ryo.Muramatsu@mail...	2482909946	Ryo Muramatsu	Asia
Shinshu University	Bio of Mizue Kaya...	Mizue.Kayama@mail...	2128134587	Mizue Kayama	Asia
Shinshu University	Bio of Kazunori I...	Kazunori.Itoh@mail...	2101782692	Kazunori Itoh	Asia
Shinshu University	Bio of Masami Has...	Masami.Hashimoto@...	2114054191	Masami Hashimoto	Asia
Shinshu University	Bio of Makoto Ota...	Makoto.Otani@mail...	1989208940	Makoto Otani	Asia
Nagano Prefectura...	Bio of Michio Shi...	Michio.Shimizu@ma...	2134989941	Michio Shimizu	Asia
Takushoku Univers...	Bio of Masahiko S...	Masahiko.Sugimoto...	2307479915	Masahiko Sugimoto	Rest of the world
Politecnico di Mi...	Bio of Pranava K...	Pranava.Jha@mail.com	2718958994	Pranava K. Jha	Europa
Archaeological Co...	Bio of G. Beale: ...	G..Beale@mail.com	2103626414	G. Beale	Rest of the world
Archaeological Co...	Bio of G. Earl: m...	G..Earl@mail.com	2117665592	G. Earl	Rest of the world
Department of Sta...	Bio of Altaf Hoss...	Altaf.Hossain@mail...	2300589394	Altaf Hossain	Rest of the world
Department of Sys...	Bio of Faisal Zam...	Faisal.Zaman@mail...	2308774408	Faisal Zaman	Asia
Department of Sta...	Bio of M. Nasser:...	M..Nasser@mail.com	2126056503	M. Nasser	Rest of the world
Department of Com...	Bio of M. Mufakhk...	M..Islam@mail.com	2425818370	M. Mufakhkharul I...	Asia
The University of...	Bio of Rafael Álv...	Rafael.Álvarez@ma...	2125293936	Rafael Álvarez	America
Department of Com...	Bio of Leandro To...	Leandro.Tortosa@m...	2101693188	Leandro Tortosa	Asia
Department of Sta...	Bio of José-Franc...	José-Francisco.Vi...	2159120860	José-Francisco Vi...	Asia
Department of Sys...	Bio of Antonio Za...	Antonio.Zamora@ma...	2146570697	Antonio Zamora	Asia
Nagano Prefectura...	Bio of Jovan Dj. ...	Jovan.Golic@mail.com	1237859792	Jovan Dj. Golic	Asia

only showing top 20 rows

Figure 2.2: Result of query 9.1.3

2.1.4. Query 4

```

1 fos_df = fos_df.withColumn('name',
2                             when(fos_df.name.contains("mathematics"),
3                                 regexp_replace(fos_df.name, 'Discrete
4                                 mathematics', 'Discrete Math'))
5                                 .when(fos_df.name.contains("Mathematics"),
6                                     regexp_replace(fos_df.name, '
7                                     Mathematics', 'Math'))
8                                     .when(fos_df.name.contains("Artificial
9                                     intelligence"),
10                                         regexp_replace(fos_df.name, 'Artificial
11                                         intelligence', 'AI'))
12                                         .otherwise(fos_df.name)).show()

```

Listing 2.3: Update of fos names with replacing

id	name
ac6663816c9635e15de8053dbf92ec41	Telecommunications network
284fcfb183d1919532b3c7a6dba33873	Computer science
d17475f16d76e40529473c3afeff8fd1	Mind map
c2a5462d06dd702e2e6a87693479a635	Human-computer interaction
2f56b4f336dc97edf739bf79523fb9a6	Multimedia
ff369ad079366681e0d102c1bdfe8f34	Empirical research
28e169980e17fc27c452e7580e186068	Comprehension
2e74da7ce756356a026dadfc11039ae4	Communications protocol
4cdbd2bafa8193091ba09509cedf94fd	Graph
27ce971356df02c63cc695dffce88863	Discrete Math
6c2f06ae9649fffd101787ec6e3859e1	Combinatorics
05df30932021c337626edb064998c7ac	Direct product
540b21ecdb276f5087ee585cedd6d5d0	Math
f34b29e2dd11d27c2d3725ffc221c3aa	Statue
e3df226c8bed8843867f4adb9b7eb7dc	Engineering drawing
7c0d914a5aa9dc8f2162f3ef93824c79	Virtual reconstruction
ce09e3d68182639402e8fd2f50368167	Visualization
1e1b9006b2ad5f189dcbdd0599d29895	Polychrome
9d0996a44c6d51cf223e833dceecb286	AI
b2a57f84041a796df2d1ff776a32db92	Autoregressive-moving-average model

only showing top 20 rows

Figure 2.3: Result of query 9.1.4

2.1.5. Query 5

```
1 publication_df= publication_df.where(publication_df.pages < 10).show()
```

abstract	authors	id	pages	publisher	references	title	venue	year
[Abstract of Preli...	[2312688602, 2482...	1091	9	Springer, Berlin,...	[1304089, 819435,...	Preliminary Desig...	1127419992	2013
[Abstract of Furth...	[2718958994]	1388	2	PoliPrint, Milano	[316117]	Further Results o...	73158690	2000
[Abstract of COMPA...	[2125293936, 2101...	5411	6	PoliPrint, Milano	[1588621, 2228599...	COMPARING GNG3D A...	1136212596	2009
[Abstract of Impro...	[2022192081, 2023...	6522	7	Springer, London	[870555, 217368, ...	Improved Secret I...	1125967516	2011
[Abstract of Fur V...	[2156900172, 2281...	8763	9	Springer, Cham	[1487729, 359688,...	Fur Visualisation...	1196868077	2014
[Abstract of Speec...	[2163873308, 1971...	11895	8	Morgan Kaufmann P...	[778541]	Speech training s...	1203999783	1979
[Abstract of Softw...	[1978340988, 1986...	12993	6	PoliPrint, Milano	[359688, 1364731,...	Software Evolutio...	50368787	2003
[Abstract of Desig...	[2404438944, 2656...	13205	3	PoliPrint, Milano	[274954, 1538985,...	Design of an audi...	1177287137	2002
[Abstract of A COM...	[192576500, 20750...	14870	4	PoliPrint, Milano	[2192057]	A COMPUTATIONAL S...	1198225011	2009
[Abstract of Lever...	[135218249, 21208...	15883	4	USENIX Association	[1506917, 899173,...	Leveraging legacy...	1185109434	2008
[Abstract of A ped...	[2789599552, 2935...	15901	3	Springer, Berlin,...	[1638427, 316117,...	A pedestrian navi...	1127419992	2013
[Abstract of Extra...	[2182498006, 2298...	21951	4	PoliPrint, Milano	[1557874, 953868,...	Extracted knowled...	1130566378	2007
[Abstract of Conte...	[2064022781, 1576...	24270	7	Fuji Technology P...	[1448152]	Context Dependent...	4511983	2007
[Abstract of FTP M...	[2581588131, 2712...	27301	8	USENIX Association	[440308, 937946, ...	FTP Mirror Tracke...	1161835747	2000
[Abstract of A Cla...	[2110538291]	29332	8	Springer, Vienna	[1446963, 2235786...	A Clausal Genetic...	1131576334	1995
[Abstract of Using...	[1220847850, 2147...	29521	5	IASTED	[188688, 1770742,...	Using Classpects ...	2755873345	2006
[Abstract of Autom...	[2123350797]	29841	4	Springer, Berlin,...	[2104602]	Automatic Detecti...	1140961231	2013
[Abstract of On th...	[2687023189, 2650...	37090	8	PoliPrint, Milano	[1309515, 2231236...	On the Design of ...	2755927266	1977
[Abstract of On th...	[1968885353]	38130	2	PoliPrint, Milano	[309696, 295139, ...	On the Universali...	1155899826	1986
[Abstract of Autom...	[2068146743, 2252...	38917	4	Springer, Berlin,...	[2165112, 942590,...	Automated Object ...	1164975091	2008

only showing top 20 rows

Figure 2.4: Result of query 9.1.5

2.2. Queries

2.2.1. WHERE, JOIN

Return the papers that have as a field of study "Artificial Intelligence"

```

1 fosDF
2   .filter(col("name") == "Artificial intelligence")
3   .join(rel_dwDF, fosDF.id == rel_dwDF.fos_id, "inner")
4   .join(publicationsDF, rel_dwDF.pub_id == publicationsDF.id, "inner")
5   .select("title")
6   .show()

```

Listing 2.4: Query 9.2.1

```

+-----+
|title|
+-----+
|A methodology for the physically accurate visualisation of roman polychrome statuary|
|Comparison of GARCH, Neural Network and Support Vector Machine in Financial Time Series Prediction|
|COMPARING GNG3D AND QUADRIC ERROR METRICS METHODS TO SIMPLIFY 3D MESHES|
|Vectorial fast correlation attacks.|
|Improved Secret Image Sharing Method By Encoding Shared Values With Authentication Bits|
|Identifying Psychological Theme Words from Emotion Annotated Interviews|
|A COMPUTATIONAL SALIENCY MODEL INTEGRATING SACCADE PROGRAMMING|
|Extracted knowledge interpretation in mining biological data: A survey|
|Automated Object Identification and Position Estimation for Airport Lighting Quality Assessment|
|Face Detection, Recognition in an Image Sequence Using Eigenedginess.|
|Qualitative Spatial and Temporal Reasoning in Cardiac Electrophysiology|
|Speech recognition based on the integration of FSVQ and neural network.|
|Auditory-based formant estimation in noise using a probabilistic framework.|
|A Study on the Development of High Precision Cam Profile CNC Grinding Machine with CAD/CAM System.|
|Simple Synchrony Networks: Learning Generalisations across Syntactic Constituents.|
|Multi-layer topology preserving mapping for K-means clustering|
|A general semantic analyser for data base access|
|Kernel PLS variants for regression|
|Two notes from experimental study on image steganalysis|
|Near-synonym choice in natural language generation|
+-----+
only showing top 20 rows

```

Figure 2.5: Result of query 9.2.1

2.2.2. WHERE, LIMIT, LIKE

Return the first 5 authors name of our database that have affiliation with a "Politecnico" and return also the titles of their publications.

```

1 # import expr
2 from pyspark.sql.functions import expr
3
4 authorsDF
5   # Rename the column of the author id for the join
6   .withColumnRenamed("id", "authorId")

```

```

7      # Affiliation with Politecnico in it .filter(col("affiliation").
like("%Politecnico%"))
8      # Limit the df got at 5 authors
9      .limit(5)
10     # Join the publicationsDF to get these 5 authors publications
11     .join(publicationsDF, expr("array_contains(authors, authorId)"))
12     # Select what we want
13     .select(col("title").alias("publicationTitle"), col("name").alias("
authorName"), "affiliation")
14     .show(truncate=False)

```

Listing 2.5: Query 9.2.2

publicationTitle	authorName	affiliation
Further Results on Independence in Direct-Product Graphs.	Pranava K. Jha	Politecnico di Milano
Vectorial fast correlation attacks.	Guglielmo Morgari	Politecnico di Bari
Software Evolution through Transformations.	Reiko Heckel	Politecnico di Bari
Design of an audio-visual speech corpus for the czech audio-visual speech synthesis.	Petr Cisar	Politecnico di Milano
Logical Derivation of a Prolog Interpreter.	Kazuhiro Fuchi	Politecnico di Milano

Figure 2.6: Result of query 9.2.2

Notice that the result contains only 5 elements because the authors to which we've limited the query published only one paper each (or, at least, we have one paper for each of them in our database).

2.2.3. WHERE, IN, Nested Query

This query return the percentage of papers that have "Computer" in their fos, e.g. paper with "Computer Science" or "Computer Vision".

```

1      computerScience_fos = fos_df.filter(col("name").contains("Computer")
).select(col("id")).collect()
2      computerScience_fos = [csf[0] for csf in computerScience_fos]
3
4      count_cs_publications = rel_df.filter(col("fos_id").isin(
computerScience_fos)).select(col("pub_id")).distinct().count()
5      print("percentage of publications that have fos about Computer:" +
str(count_cs_publications/2500*100) + "%")

```

Listing 2.6: Query 9.2.3

Percentage of publications that have fos with Computer: 80.4%

Figure 2.7: Result of query 9.2.3

2.2.4. GROUP BY, 1 JOIN, AS

The query counts for each name of the venue (that has different ID since it can have different editions or volumes) the number of papers that were presented there

```

1 venue_df.join(publication_df, venue_df.id == publication_df.venue, "
  inner")\
2   .groupby("raw")\
3   .count()\
4   .select(venue_df.raw, col("count").alias("Number of
  papers for every raw")).orderBy(col("Number of papers for every raw")
  .desc()).show()

```

Listing 2.7: Query 9.2.4

```

+-----+-----+
|          raw|Number of papers for every raw|
+-----+-----+
|Conference of the...|111|
|Applied Reconfigu...|57|
| Int. CMG Conference|41|
|International Con...|40|
|International Con...|37|
|International Con...|36|
|Software Engineer...|36|
|Medical Image Com...|35|
|          Computing|33|
|Journal of Object...|32|
|International Con...|30|
|Parallel and Dist...|30|
|International Con...|30|
|International Con...|28|
|Americas Conferen...|27|
|International Con...|27|
|International Joi...|25|
|Developments in L...|24|
|Database and Expe...|24|
|Annales Des Téléc...|24|
+-----+-----+
only showing top 20 rows

```

Figure 2.8: Result of query 9.2.4

2.2.5. WHERE, GROUP BY

Filter the papers that have at least 3 authors, then it shows for every publisher the max number of pages between the papers he published, shown in descending order

```

1    publication_df.filter(size('authors') >= 3)\
2        .groupBy('publisher').max('pages')\
3        .select(publication_df.publisher, col("max(pages)").alias("
    Maxpages"))\
4        .orderBy(col("Maxpages").desc())\
5        .show(truncate=false)

```

Listing 2.8: Query 9.2.5

publisher	Maxpages
Digital Government Society of North America	20
AAAI Press	20
International Foundation for Autonomous Agents and Multiagent Systems	20
Springer	20
Stud Health Technol Inform	20
L. Erlbaum Associates Inc.	20
Springer, London	20
Springer Berlin Heidelberg	20
Springer, Cham	20
NIST	20
Fuji Technology Press Ltd.	20
PoliPrint, Milano	20
Springer, Berlin, Heidelberg	20
Kluwer Academic Publishers	19
Springer, Dordrecht	19
Elsevier	19
IOS Press	19
Society for Computer Simulation International	19
Centre for Discrete Mathematics & Computing	19
International Speech and Communication Association	19

only showing top 20 rows

Figure 2.9: Result of query 9.2.5

2.2.6. GROUP BY, HAVING, AS

The query groups the venues by affiliation, for each of them it counts the number of authors by id and collects the names of the authors in a list, then it filters (HAVING) the affiliations by the number of authors in the list that should be between 5 and 15. We

order the table by descending order for the number of authors, and in case of tie they show the affiliation name in alphabetic order

```

1  author_df.groupBy("affiliation")\
2    .agg(
3      countDistinct("id").alias("Number of Authors"),
4      collect_list(author_df.name).alias("Authors list")
5    )\
6    .filter((col("Number of Authors") < 15) & (col("Number
of Authors") > 5))\
7    .orderBy(col("Number of Authors").desc(), col("
affiliation").asc())\
8    .show(truncate=False)

```

Listing 2.9: Query 9.2.6

affiliation	Number of Authors	Authors list
CHINESE ACADEMY O...	12	[Dengguo Feng, Ch...
Stanford, University	12	[Edward H. Shortl...
Harbin Institute ...	11	[Wangmeng Zuo, Ho...
RWTH Aachen Unive...	11	[Nicolas R. Gauge...
Microsoft Researc...	9	[John R. Douceur, ...
Northeastern, Uni...	8	[Alireza Khalafi, ...
Regenstrief Insti...	8	[Gunther Schadow, ...
Faculty of System...	7	[Hirokazu Taki, F...
Humboldt-Universi...	7	[Mathias Nitzsche...
Nanyang Technolog...	7	[Yin-Leng Theng, ...
National Universi...	7	[Qaiser Mehmood, ...
The University Of...	7	[Michele Turitto, ...
UNIVERSITY OF AVEIRO	7	[Iouliia Skliarov...
University of Hei...	7	[Karl Rohr, Reinh...
University of Okl...	7	[Matthew L. Jense...
University of Sal...	7	[Manfred Tschelig...
University of Wat...	7	[Therese C. Biedl...
VŠB-Technical uni...	7	[Václav Snášel, J...
#N##TAB##TAB##TAB...	6	[Hiroki Arimura, ...
, Aalborg University	6	[Niels Nørgaard S...
Aoyama Gakuin Univ.	6	[Takashi Kawashim...
Centro Nacional d...	6	[Núria Malats, Da...
ETH Zürich	6	[Roger Gassert, O...
Email: contact@sk...	6	[Daniel Rodriguez...
Hebei United Univ...	6	[Huaiyong Nie, Mi...
Hunan University	6	[Li Shutao, Yu Xi...
Nippon Hoso Kyokai	6	[Kazuo Onoe, Shin...
Otto von Guericke...	6	[Gunter Saake, Cl...
Polish Academy of...	6	[Mieczysław A. Kł...
School of Compute...	6	[Zhanhuai Li, Zho...

only showing top 30 rows

Figure 2.10: Result of query 9.2.6

2.2.7. WHERE, GROUP BY, HAVING, AS

This query lists the Publisher that have published at least 2 papers presented in a Conference, and shows the number of published paper, as well as the average number of pages of a paper published by them.

```

1  publication_df.filter(publication_df.type == "Conference").groupBy("
publisher").agg(
2  count("id").alias("number of published paper"),
3  avg("pages").alias("Pages average"),
4  ).filter(
5      col("number of published paper") > 1
6  ).orderBy(
7      col("number of published paper").desc(),
8      col("pages average").desc()
9  ).show(truncate = False)

```

Listing 2.10: Query 9.2.7

publisher	number of published paper	Pages average
PoliPrint, Milano	1247	10.40176423416199
Springer, Berlin, Heidelberg	580	10.572413793103449
Springer, Cham	115	10.634782608695652
Morgan Kaufmann Publishers Inc.	32	8.15625
Springer	27	11.703703703703704
IOS Press	27	10.444444444444445
Springer Berlin Heidelberg	23	11.130434782608695
American Medical Informatics Association	22	10.954545454545455
AAAI Press	21	9.666666666666666
USENIX Association	18	9.666666666666666
Springer-Verlag	15	12.066666666666666
Springer, Boston, MA	15	8.066666666666666
Springer, London	12	7.75
Springer, Dordrecht	11	10.636363636363637

Figure 2.11: Result of query 9.2.7

2.2.8. WHERE, Nested Query (i.e., 2-step Queries), GROUP BY

Count the publications grouping by the type of the venue they are taken from ('C' for Conference, 'J' for Journal). The publications counted are only the ones with at least one author from the 'Politecnico di Bari'.

```

1  authors_from_bari = author_df.filter(col('affiliation') == 'Politecnico
di Bari').select('id').collect()
2  authors_from_bari = [barese[0] for barese in authors_from_bari]
3
4  exploded_pub = publication_df.select(publication_df.id, publication_df.
venue, explode(publication_df.authors))

```

```

5 exploded_pub = exploded_pub.withColumnRenamed("col", "author")
6 exploded_pub = exploded_pub.withColumnRenamed("id", "pub_id")
7
8 exploded_pub.filter(col('author').isin(authors_from_bari))\
9     .join(venue_df, exploded_pub.venue == venue_df.id, "inner")\
10    .groupBy('type').agg(countDistinct('pub_id').alias('n_doc')).show()

```

Listing 2.11: Query 9.2.8

```

+-----+-----+
|type|n_doc|
+-----+-----+
|  C | 127 |
|  J | 113 |
+-----+-----+

```

Figure 2.12: Result of Query 9.2.8

2.2.9. WHERE, GROUP BY, HAVING, 1 JOIN

From all the publications wrote before the 2000s, compute the average weight of the weights of its Field of Studies and display the top five of them, having the average above 0.5.

```

1 publication_df.filter(col('year')<2000)\
2     .join(rel_df, publication_df.id == rel_df.pub_id, "inner")\
3     .groupBy('id').agg(avg('weight').alias('avg_weight')).filter(col('
4     avg_weight')>0.5)\
5     .orderBy(col('avg_weight').desc())\
6     .limit(5).show()

```

Listing 2.12: Query 9.2.9

id	avg_weight
1840116	0.7599
683650	0.744615
1787282	0.735108
1645932	0.7261075
1546711	0.6944836363636363

Figure 2.13: Result of Query 9.2.9

2.2.10. WHERE, GROUP BY, HAVING, 2 JOINS

Count the number of publications, grouping them by publisher, having the Field of Study dealing with Computers (its name must contain the word 'Computer').

```

1 fos_df.filter(col('name').contains('Computer'))\
2   .join(rel_df, fos_df.id == rel_df.fos_id, "inner")\
3   .join(publication_df, rel_df.pub_id == publication_df.id, "inner")\
4   .groupBy('publisher').agg(countDistinct('pub_id').alias('n_pub'))\
5   filter(col('n_pub')>25)\
6   .orderBy(col('n_pub').desc())\
   .limit(5).show()

```

Listing 2.13: Query 9.2.10

publisher	n_pub
PoliPrint, Milano	1046
Springer, Berlin,...	459
Springer, Cham	95
Morgan Kaufmann P...	32

Figure 2.14: Result of Query 9.2.10