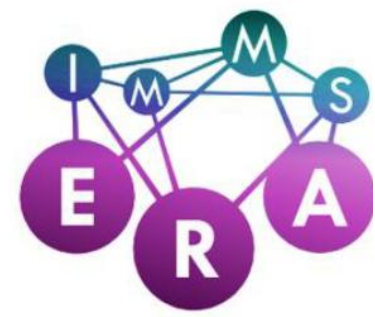


5th International Meeting on Materials Science for Energy Related Applications

September 25-26, 2025, Belgrade



Automatizing data storage, analysis, and sharing for the electrochemical CO₂ reduction – the FAIR case of SuPERCO₂

Dr. Federico Dattila

Department of Applied Science and Technology
Polytechnic of Turin



**Politecnico
di Torino**

SESSION 3 – (ELECTRO)CATALYSIS IN ENERGY
CONVERSION & STORAGE

25/09/25, Belgrade, Serbia

The polytechnic of Turin

Located in Turin

Technical university since 1859

Both experimental and computational groups working on electrocatalysis

1.5h flight Milan-Belgrade, then 2h bus

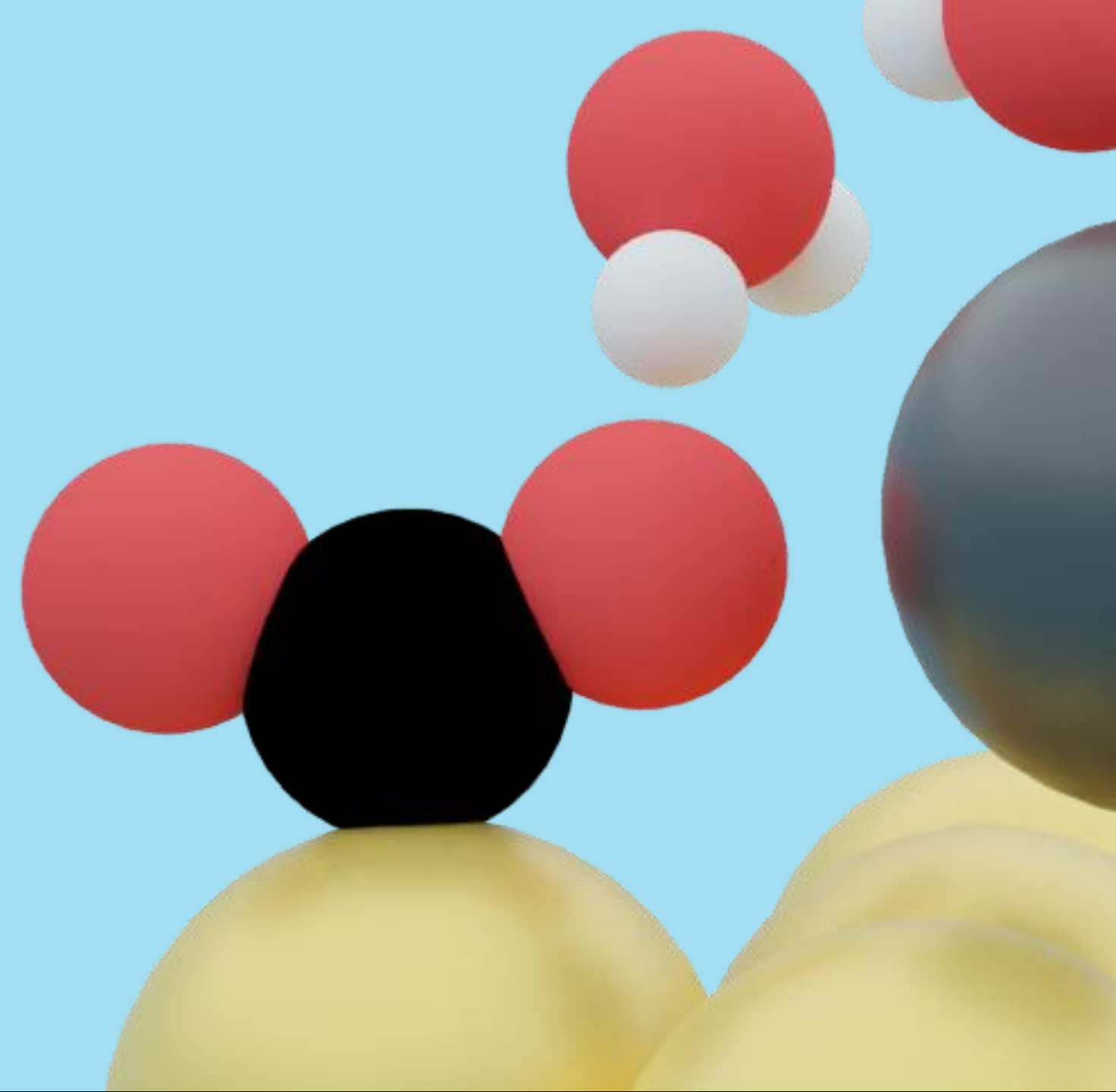


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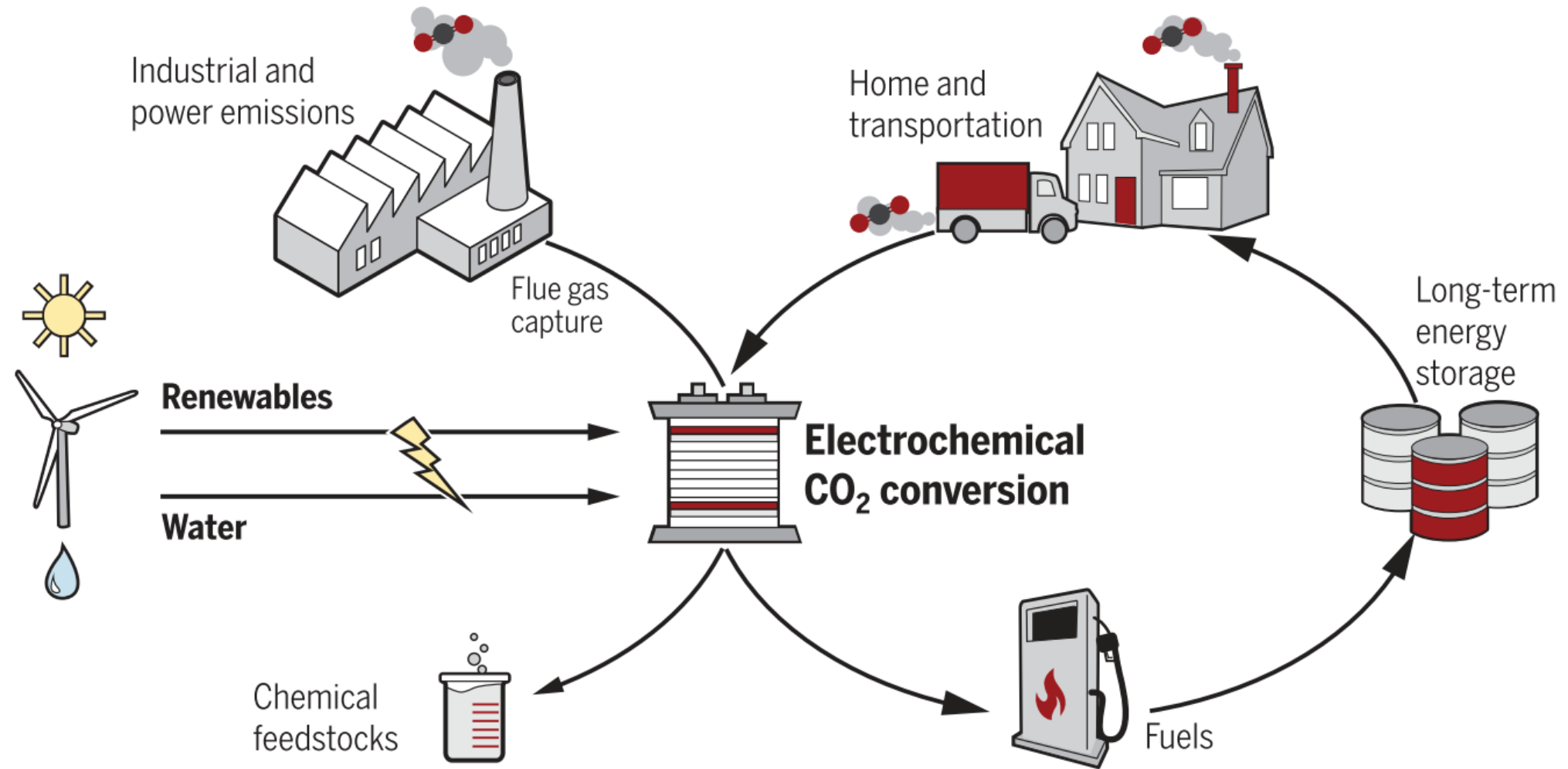


ECCELLENZA 2019 – 2027



The FAIR problem of CO₂R

Electrochemical CO₂ reduction



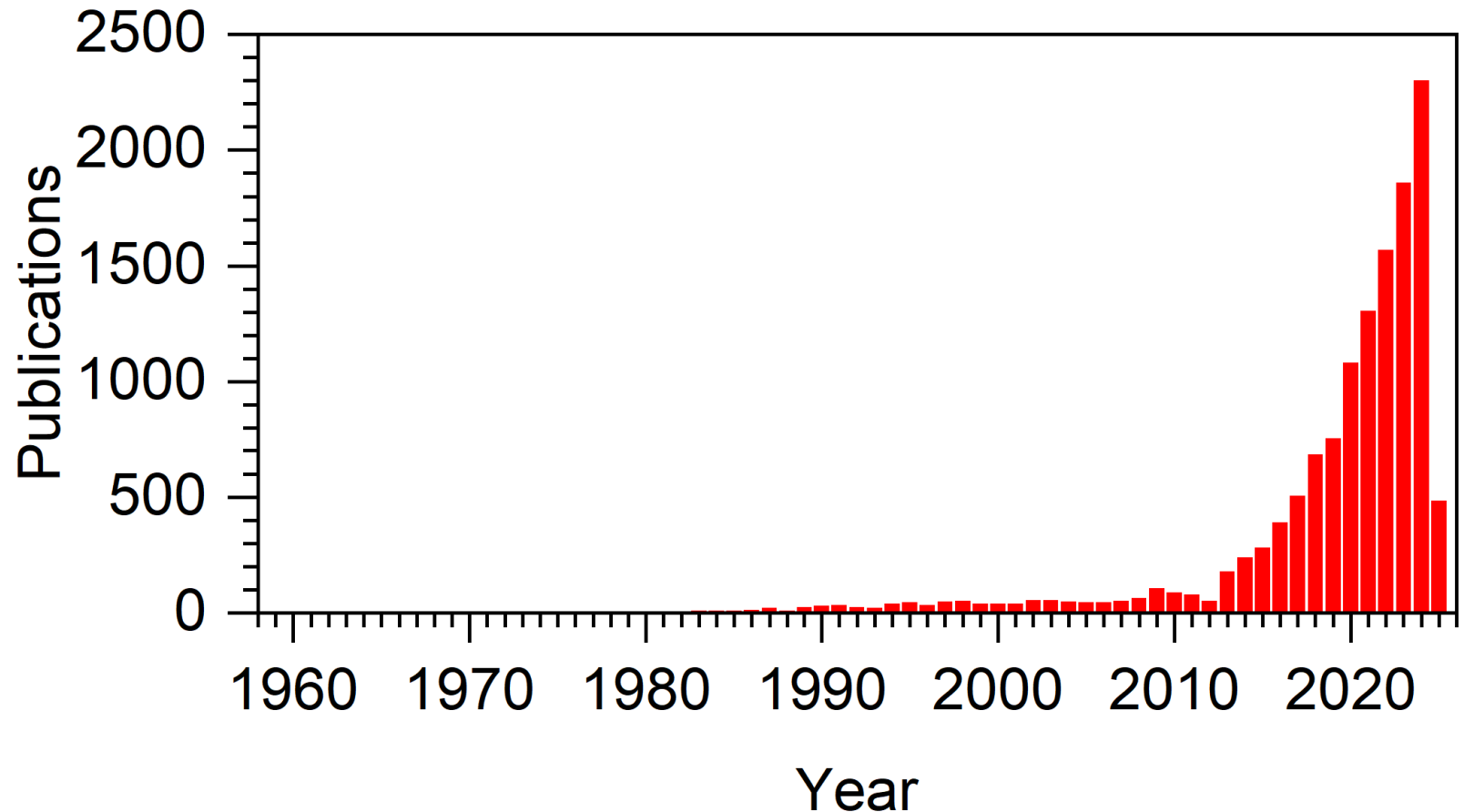
Publications on CO₂R

POLAROGRAPHIC REDUCTION OF CARBON DIOXIDE

Sir:

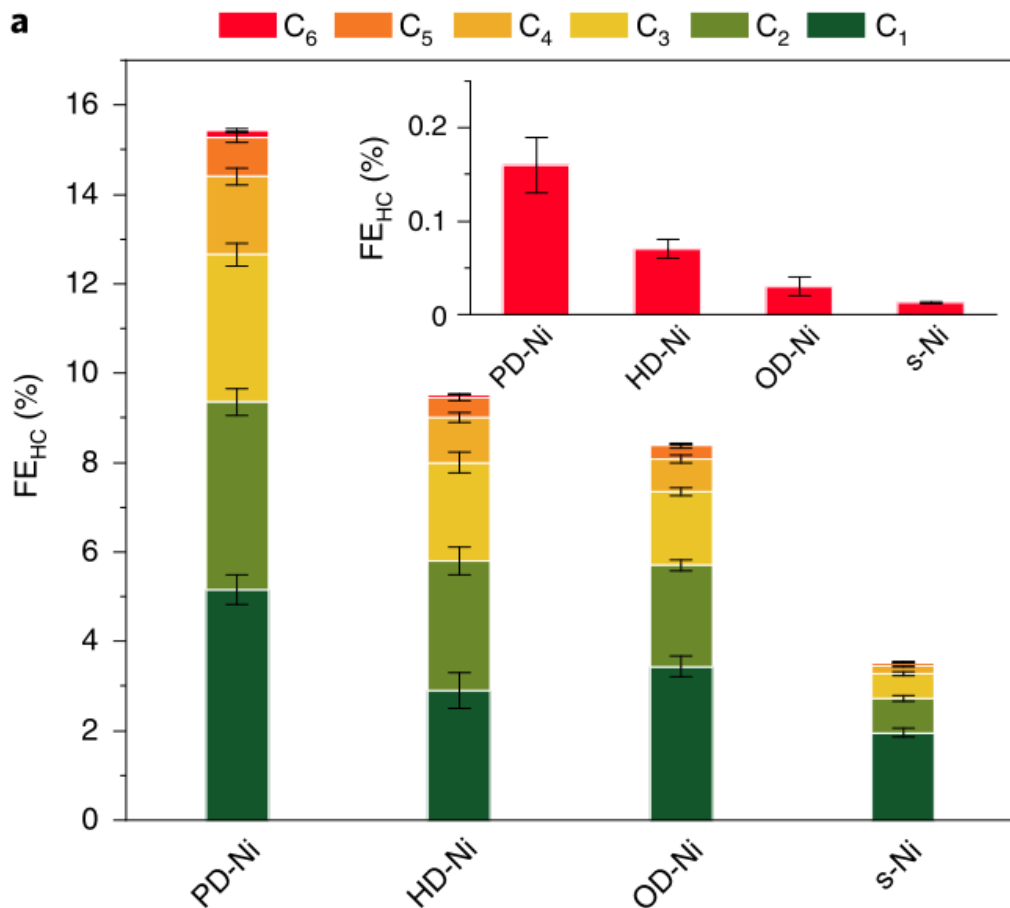
We have been engaged for some time in the study of the reduction of carbon dioxide at the dropping mercury cathode. On the basis of a large number of polarograms obtained by means of a Heyrovsky Polarograph of the Sargent Co. with solutions of carbon dioxide in 0.1 molar tetramethylammonium chloride, we have arrived at the following definite conclusions: carbon dioxide exhibits well-defined reduction waves with a half-wave potential of remarkable constancy: -2.24 ± 0.01 volt referred to the saturated calomel electrode. This is the average of ten independent determinations with varying amounts of carbon dioxide. When hydrogen is passed through the substituted ammonium salt solution before the dissolution of carbon dioxide, the waves are equally well defined, but the half-wave potential is somewhat less negative: -2.18 ± 0.02 volt referred to the saturated calomel electrode. This is the average of seven independent determinations, with varying amounts of hydrogen and carbon dioxide.

Van Rysselberghe, P.; Alkire, G. J.
J. Am. Chem. Soc. **1944**, 66, 1801.



The need for FAIR protocols

Findability / Accessibility / Interoperability / Reusability



Data to Zenodo

Figure 1 CO₂ electroreduction to hydrocarbons on Ni catalysts

Figure 1a FE of C1-C6 hydrocarbons formed on s-Ni, PD-Ni, HD-Ni and OD-Ni.xlsx

9.9 kB

Figure 1b Stability of the PD-Ni catalyst.xlsx

802.2 kB

Figure 1c Ratio of the FE of hydrocarbons formed on PD-Ni compared with that on s-Ni versus the carbon number.xlsx

8.6 kB

Figure 2 Hydrocarbon formation on Cu and PD-Ni

Figure 2 Hydrocarbon formation on Cu and PD-Ni.xlsx

8.8 kB

Figure 3 Redox response and operando X-ray absorption spectroscopy studies

Figure 3a Cyclic voltammograms of nickel phosphate under Ar and CO₂.xlsx

142.5 kB

Files (1.2 MB)

Name

Size

Download all

Data to Zenodo.zip

md5:50fc947751f1d5ce4c5255f83b55745c

1.2 MB

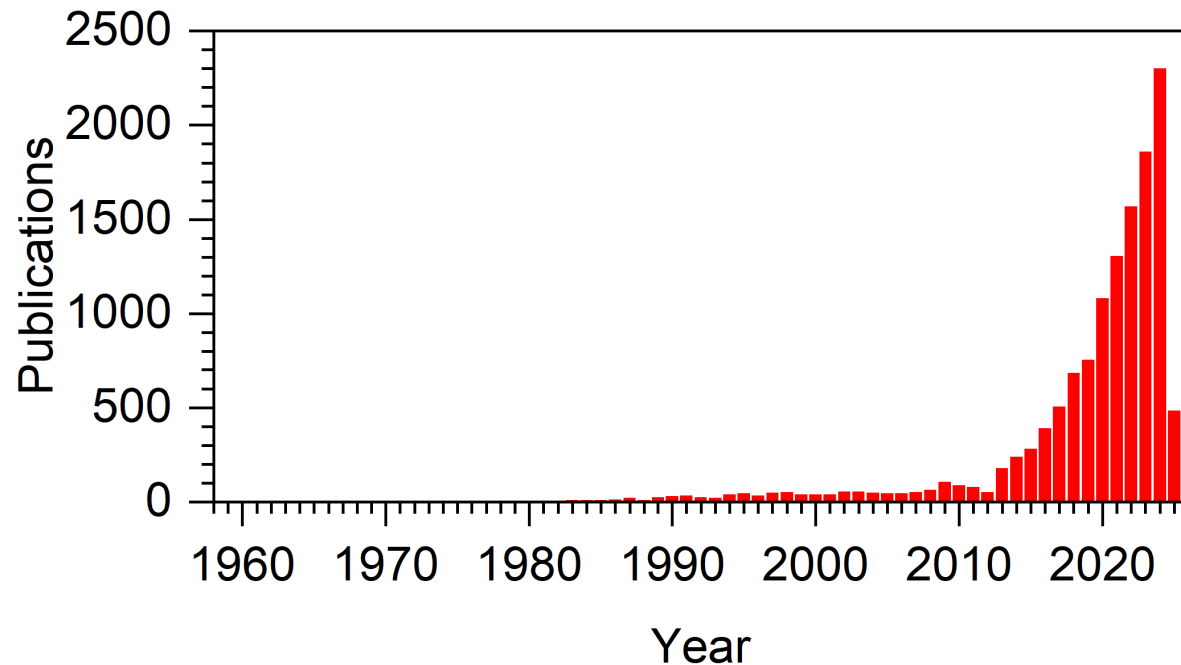
Preview

Download

The limit of current repositories

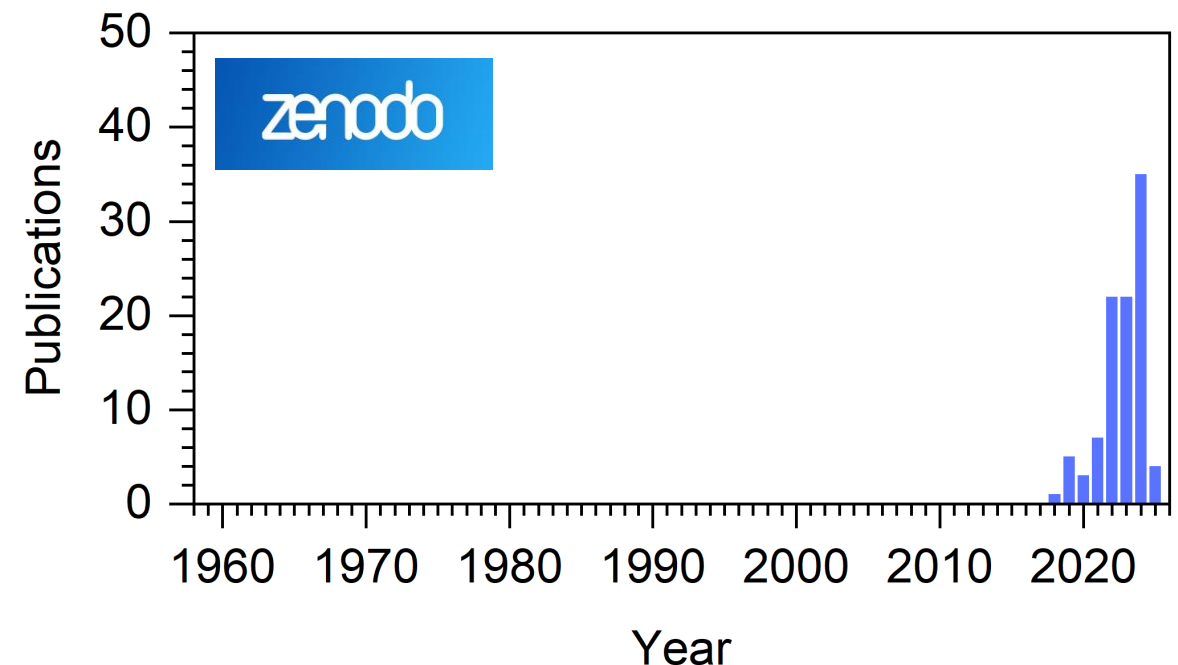
Keywords:

Electrochemical CO₂ reduction

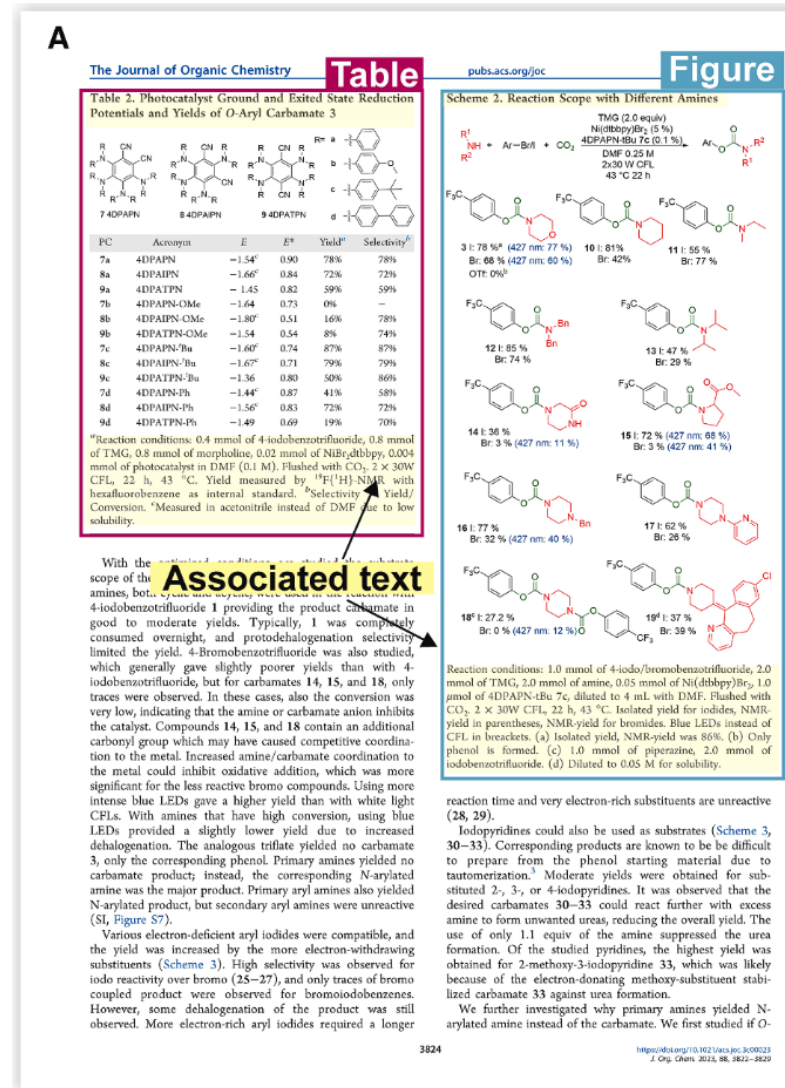
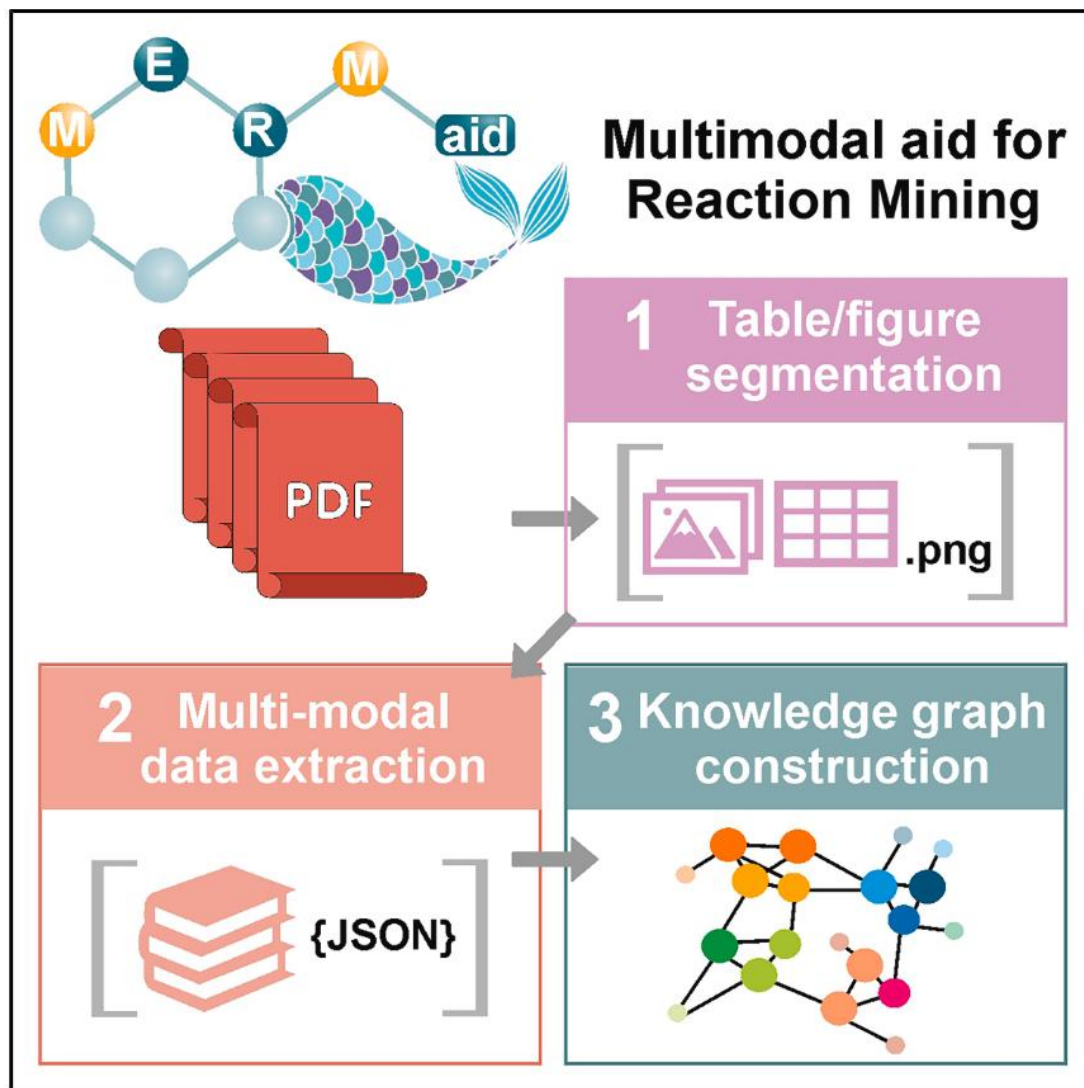


Keywords:

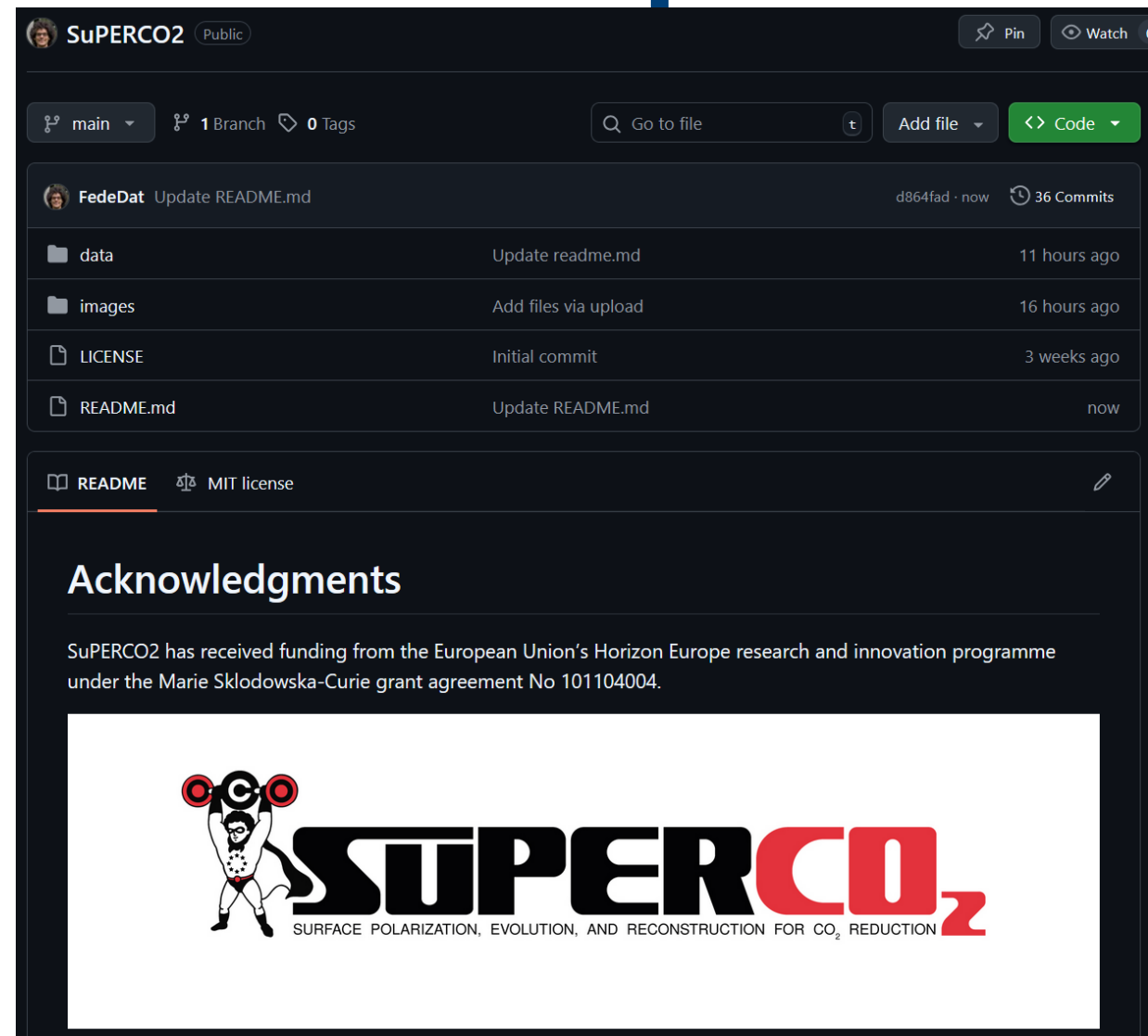
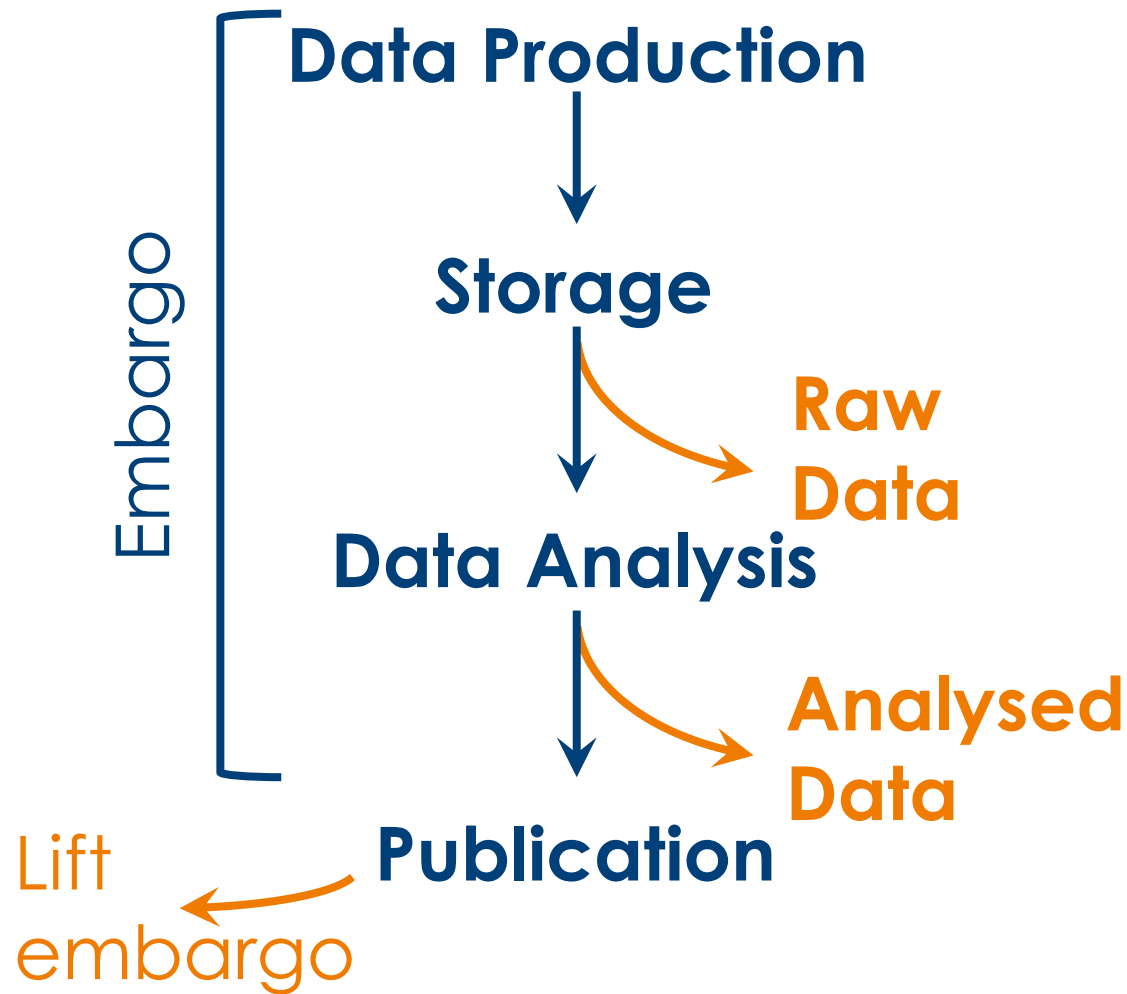
Electrochemical CO₂ reduction,
Zenodo

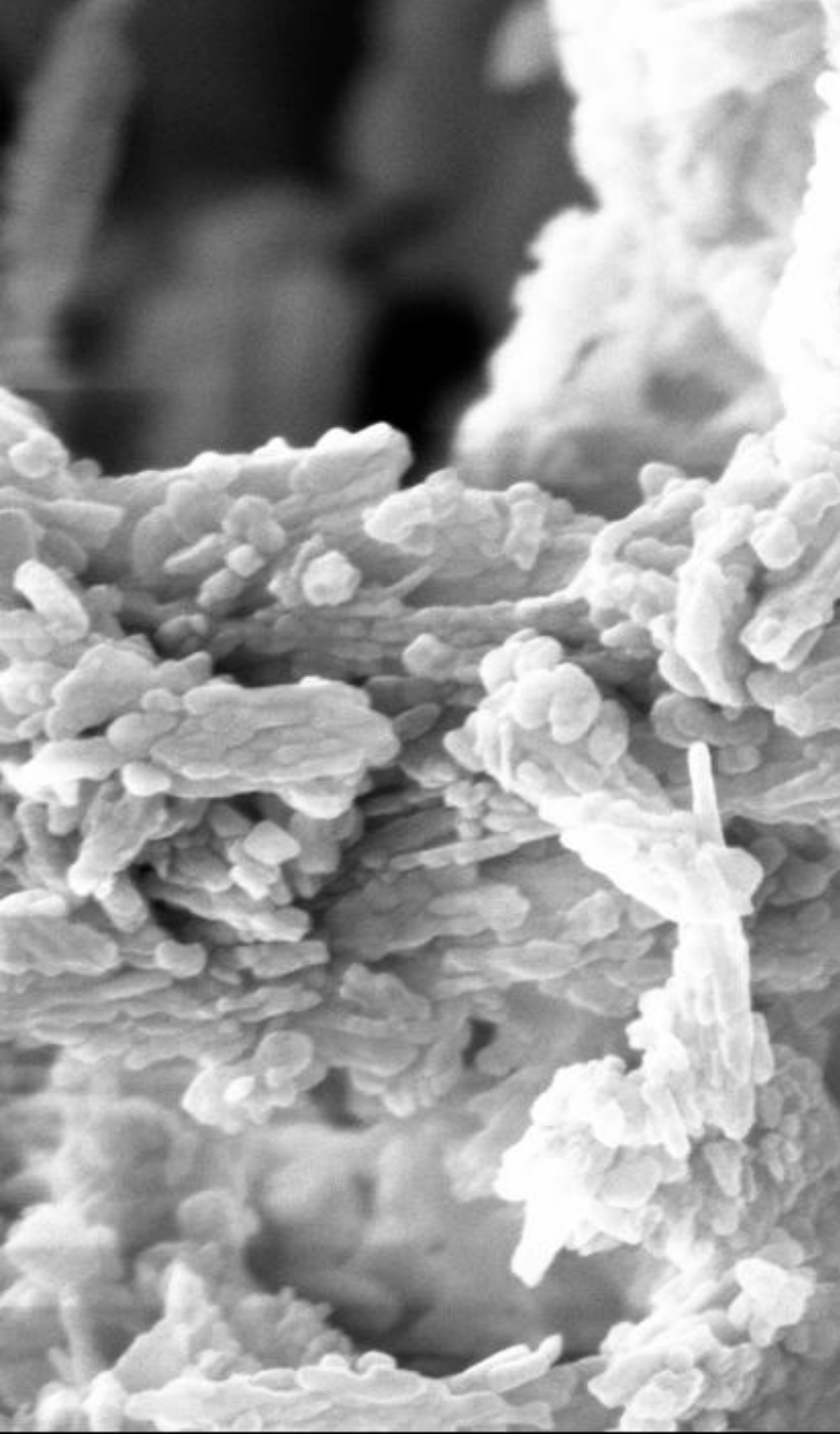


AI-based solution: MERmaid



The SuPERCO2 concept





Characterization

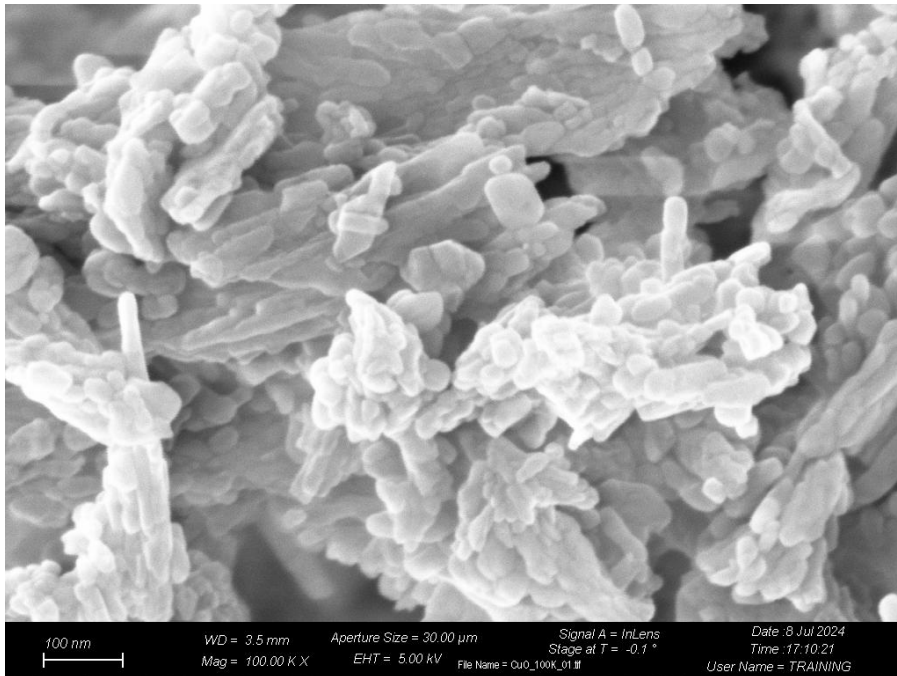
Repository: github.com/FedeDat/SuPERCO2

item: data/24-10-30_CuO/characterization

Scanning Electron Microscopy

Post-processing

Data analysis & Visualization



SuPERCO2 / data / 24-10-30_CuO / characterization / fesem / [↑ Top](#)

100K.png	Add files via upload	11 hours ago
150K.png	Add files via upload	11 hours ago
25K.png	Add files via upload	11 hours ago
50K.png	Add files via upload	11 hours ago
5K.png	Add files via upload	11 hours ago
readme.md	Update readme.md	11 minutes ago

readme.md [✎](#) [☰](#)

Methods

A ZEISS Supra 40 Field Emission Scanning Electron Microscopy (FESEM) was employed to characterize the morphology of the CuO powder electrode.


5K magnification

A high-magnification scanning electron microscopy (SEM) image showing the morphology of CuO powder at 5K magnification. The image displays a dense collection of elongated, needle-like or rod-like structures.

Energy Dispersive X-ray

Post-processing

Data analysis & Visualization

 **FedeDat** Add files via upload

Preview Code Blame 4 lines

Search this file

	Cu	Fe	O	C	Na	Cl
1						
2	42.59		57.41			
3	42.77		57.23			
4	41.37		58.63			

```
def cm_to_inch(cm):
    inch=float(cm/2.54)
    return inch

def AgCl_to_RHE(V_AgCl,pH,ref):
    if ref == "Ag/AgCl":
        V_RHE=+0.197+V_AgCl+0.059*pH
    return V_RHE

# Read the content of the file
url_EDX = "https://raw.githubusercontent.com/FedeDat/SuPERCO2/main/data/24-10-30_CuO/characterization/EDX.csv"

data_EDX = pd.read_csv(url_EDX, usecols=lambda column: pd.notnull(column))
data_EDX.dropna(axis=1, how='all', inplace=True)

#Calculate ratio between metal and oxygen

data_EDX[''+str(data_EDX.columns[0])+'/'+str(data_EDX.columns[1])+'']=data_EDX[data_EDX.columns[0]]/data_EDX[data_EDX.columns[1]]

data_EDX.index=list(range(1, len(data_EDX)+1))

ax1 = data_EDX.iloc[:, :-1].plot(kind='bar', stacked=True, figsize=(cm_to_inch(10), cm_to_inch(8)))
plt.legend(loc='upper center')

ax2 = ax1.twinx() # Create a second y-axis
ax2.plot(data_EDX.index-1, data_EDX[data_EDX.columns[-1]], linestyle='None', marker='o', markersize=8, markerfacecolor='red')

# Adding titles and labels
ax1.set_xticklabels(list(range(1, len(data_EDX)+1)), rotation=0, ha='right')
```

X-ray Diffraction

Post-processing

Data analysis & Visualization

SuPERCO2 / data / 24-10-30_CuO / characterization / XRD.csv

	Preview	Code	Blame	2694 lines (2694 loc) · 46 KB
28	Angle	Intensity		
29	20.01513028	862		
30	20.04139085	848		
31	20.06765141	881		
32	20.09391198	856		
33	20.12017254	853		
34	20.14643311	891		
35	20.17269367	876		

```
def AgCl_to_RHE(V_AgCl,pH,ref):  
    if ref == "Ag/AgCl":  
        V_RHE=+0.197+V_AgCl+0.059*pH  
        return V_RHE  
  
url_XRD = "https://raw.githubusercontent.com/FedeDat/SuPERCO2/main/data/24-10-30_CuO/characterization/XRD.csv"  
  
data_XRD = pd.read_csv(url_XRD, header=None, usecols=[0, 1], names=['Angle', 'Intensity'])  
  
k_XRD = float(data_XRD[data_XRD.apply(lambda row: row.astype(str).str.contains('K-Alpha1 wavelength')).any(axis=1)]['Intensity'])  
n_XRD = int(data_XRD[data_XRD.apply(lambda row: row.astype(str).str.contains('No. of points')).any(axis=1)]['Intensity'])  
  
data_XRD_s1 = data_XRD.iloc[(data_XRD[data_XRD.apply(lambda row: row.astype(str).str.contains('Angle')).any(axis=1)].index[0])  
data_XRD_s1 = data_XRD_s1.astype(float)  
  
data_XRD_s1['Intensity'] = data_XRD_s1['Intensity'] - np.min(data_XRD_s1['Intensity'])  
data_XRD_s1['Intensity']=data_XRD_s1['Intensity']/np.max(data_XRD_s1['Intensity'])  
  
data_XRD_s1['Intensity_savgol'] = savgol_filter(data_XRD_s1['Intensity'], window_length=10, polyorder=1)  
  
k_XRD = float(data_XRD[data_XRD.apply(lambda row: row.astype(str).str.contains('K-Alpha1 wavelength')).any(axis=1)]['Intensity'])  
n_XRD = int(data_XRD[data_XRD.apply(lambda row: row.astype(str).str.contains('No. of points')).any(axis=1)]['Intensity'])
```




Electrochemical testing

Repository: github.com/FedeDat/SuPERCO2

item: data/24-10-30_CuO/EC

Open Circuit Potential

Post-processing

Data analysis & Visualization

SuPERCO2 / data / 24-10-30_CuO / EC / OCV.csv



FedeDat Add files via upload

Preview

Code

Blame

548 lines (548 loc)

Search this file

1	pH	6.8
2	Ref	Ag/AgCl
3		
4	t (min)	Ewe (V)
5	0.00E+00	-2.14E-01

```
# Read the content of the file

url_OCV = "https://raw.githubusercontent.com/FedeDat/SuPERCO2/main/data/24-10-30_CuO/EC/OCV.csv"

data_OCV = pd.read_csv(url_OCV, header=None, names=['time', 'V'])

pH_OCV=float(data_OCV['V'][0])
ref_OCV=str(data_OCV['V'][1])

data_OCV=data_OCV[4:]

data_OCV=data_OCV.astype(float)

data_OCV['V']=AgCl_to_RHE(data_OCV['V'],6.8,ref_OCV)

plt.figure(figsize=(cm_to_inch(12), cm_to_inch(8)))

plt.plot(data_OCV['time'],data_OCV['V'], )

plt.title('Open circuit potential')

plt.xlabel(r"$t$ (min)")
plt.ylabel(r"$V$ (vs RHE)")

plt.ylim(min(np.min(data_OCV['V']),0.5*np.mean(data_OCV['V'])),max(np.max(data_OCV['V']),1.5*np.mean(data_OCV['V'])))

plt.savefig('OCV.png', format='png', dpi=300, transparent=True, bbox_inches='tight')


plt.close()
```

Cyclic voltammetry

Post-processing

Data analysis & Visualization

SuPERCO2 / data / 24-10-30_CuO / EC / CV-CO2.csv

 **FedeDat** Add files via upload

Preview Code Blame 1810 lines (1810 loc)

Search this file

1	pH	6.8
2	Ref	Ag/AgCl
3		
4	Ewe (V)	j (mA/cm2)
5	-1.45E+00	-8.35E+00

```
data_CV_CO2['V']=AgCl_to_RHE(data_CV_CO2['V'],6.8,ref_CV_CO2)

url_CV_N2 = "https://raw.githubusercontent.com/FedeDat/SuPERCO2/main/data/24-10-30_CuO/EC/CV-N2.csv"
data_CV_N2 = pd.read_csv(url_CV_N2, header=None, names=['V', 'j'])

pH_CV_N2=float(data_CV_N2['j'][0])
ref_CV_N2=str(data_CV_N2['j'][1])

data_CV_N2=data_CV_N2[4:]

data_CV_N2=data_CV_N2.astype(float)

data_CV_N2['V']=AgCl_to_RHE(data_CV_N2['V'],6.8,ref_CV_N2)

plt.figure(figsize=(cm_to_inch(12), cm_to_inch(12)))

plt.plot(data_CV_CO2['V'],data_CV_CO2['j'], label=r"CO2 saturated",color='red')
plt.plot(data_CV_N2['V'],data_CV_N2['j'], label=r"N2 saturated",color='gray')

plt.title('Cyclic voltammetry')
plt.legend(loc='best')

plt.xlabel(r"$V$ (vs RHE)")
plt.ylabel(r"$j$ (mA/cm2)")

plt.savefig('CV.png', format='png', dpi=300, transparent=True, bbox_inches='tight')

plt.close()
```

alculation of CO₂ reduction activity and selectivity

```
# Analyse Faradaic Efficiency
```

Chronoamperometry

Post-processing (CA)

SuPERCO2 / data / 24-10-30_CuO / EC / CA1.csv

FedeDat Add files via upload

Preview Code Blame 21125 lines (21125 loc) · 392 Bytes

Search this file

1	pH	6.8
2	Ref	Ag/AgCl
3	V	-1.199
4	Flow Rate (NmL/min)	20
5	Type	CA
6		
7	t (s)	j (mA/cm2)
8	2.63E+00	-2.43E-05

Post-processing (GC)

SuPERCO2 / data / 24-10-30_CuO / EC / GC1.csv

FedeDat Add files via upload

Preview Code Blame 11 lines (11 loc) · 392 Bytes

Search this file

1	H2	O2	N2	CH4	CO	CO2	C2H4
2	0	0	0.938709	0.00071317	0	0	0
3	0.08128675	0	0.94220475	0	0	0	0
4	0.15007389	0	0.93674202	0	0	0	0
5	0.14496378	0	0.93663998	0	0	0	0
6	0.14065497	0	0.92155004	0	0	0	0
7	0.14463469	0	0.91517259	0	0	0	0
8	0.14796505	0	0.91493298	0	0	0	0

Activity & Selectivity

Data analysis & Visualization

Calculation of CO₂ reduction activity and selectivity

```
▶ # Analyse Faradaic Efficiency

# GitHub API URL for the folder
api_url = "https://api.github.com/repos/FedeDat/SuPERCO2/contents/data/24-10-30_CuO/EC"

# Get the file listing
response = requests.get(api_url)
files = response.json()

metadata_EC = pd.DataFrame(columns=['pH', 'ref', 'V'])
anl_EC = pd.DataFrame(columns=['V_RHE', 'Flow Rate', 'Mode', 'Average j', 'Q (mC)'])

data = pd.DataFrame()

n1=0

plt.figure(figsize=(cm_to_inch(18), cm_to_inch(12)))
```

The FAIR CO₂R problem

Open data is limited in CO₂R

Zenodo limits interoperability

AI mining on published data

The SuPERCO2 concept

Early data sharing

Characterization: EDX XRD

Electrochemistry: OCV CV CA

Next steps

Polish & upload Python scripts

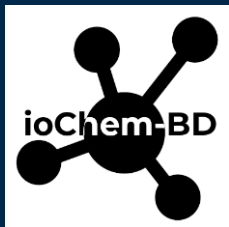
Provide a tutorial

Automatize upload & analysis

Create new modules: CP, etc.

Acknowledgements

Inspiration



M.Eng. Moisés Álvarez
(ICIQ)

Exp collaboration

M.Sc. Ali Zarei
(catalyst synthesis, electrochemical testing)

Prof. Simelys Hernández
(lab Access)



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Thank you!
Questions or feedbacks?

Here or via email
(federico.dattila@polito.it)



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