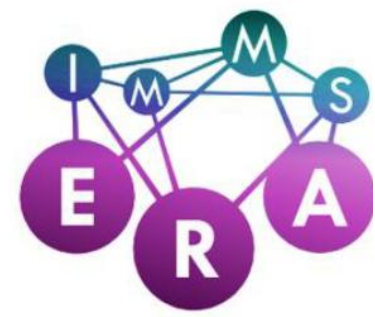


5th International Meeting on Materials Science for Energy Related Applications

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

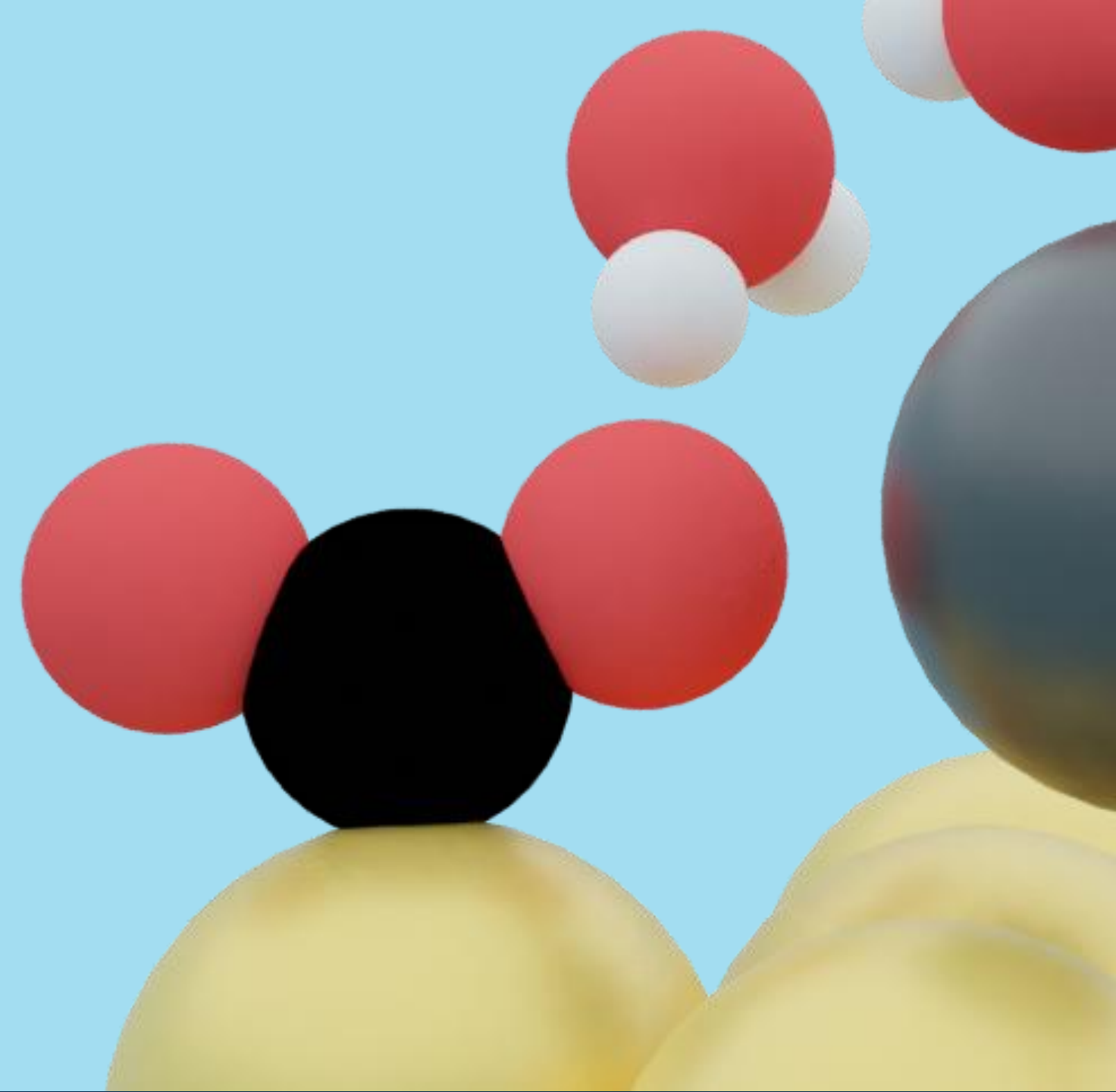
Polytechnic of Turin



**Politecnico
di Torino**

General meeting – Prof. Simelys Hernández's group

17/10/25, Turin



The FAIR problem of CO₂R

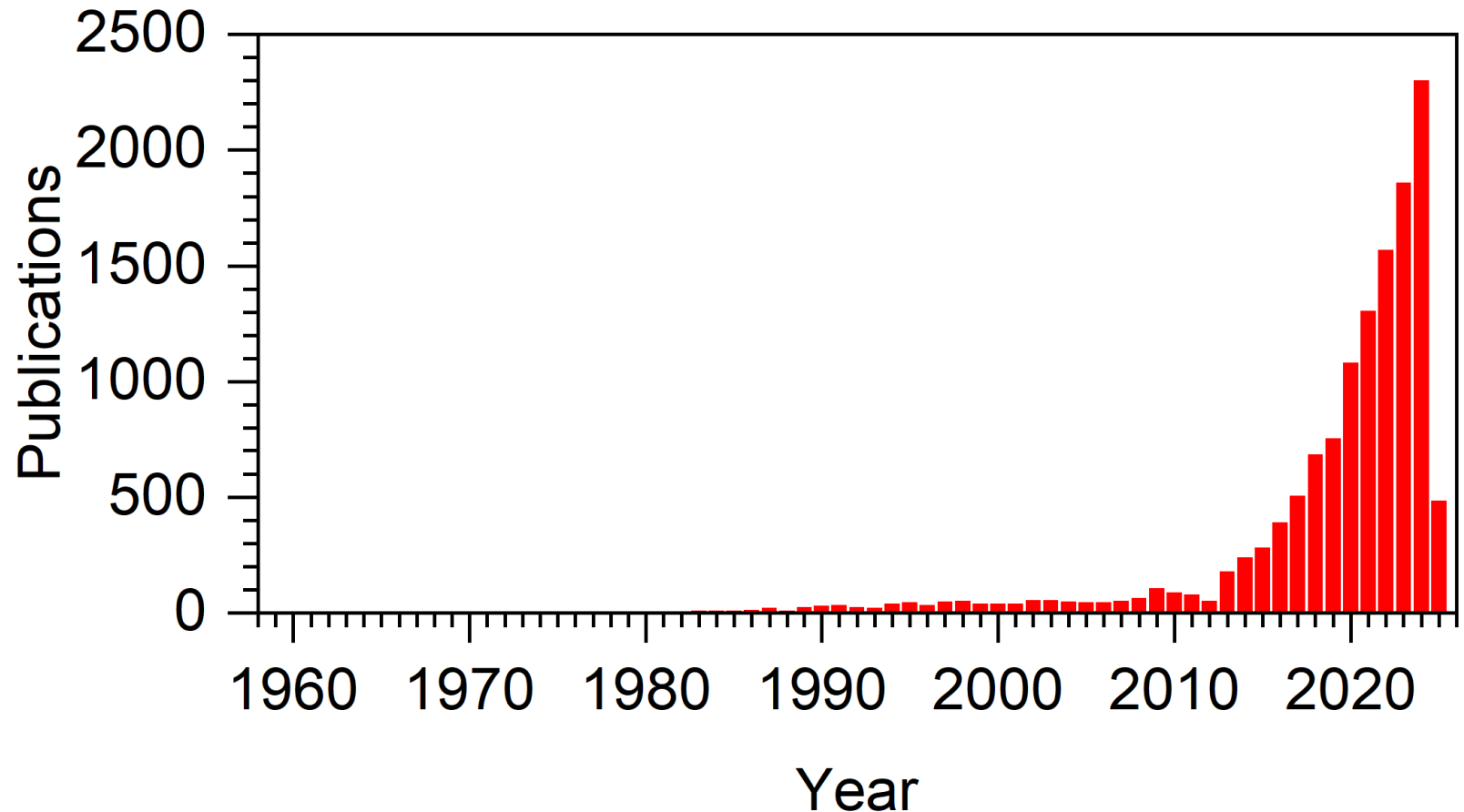
Publications on CO₂R

POLAROGRAPHIC REDUCTION OF CARBON DIOXIDE

Sir:

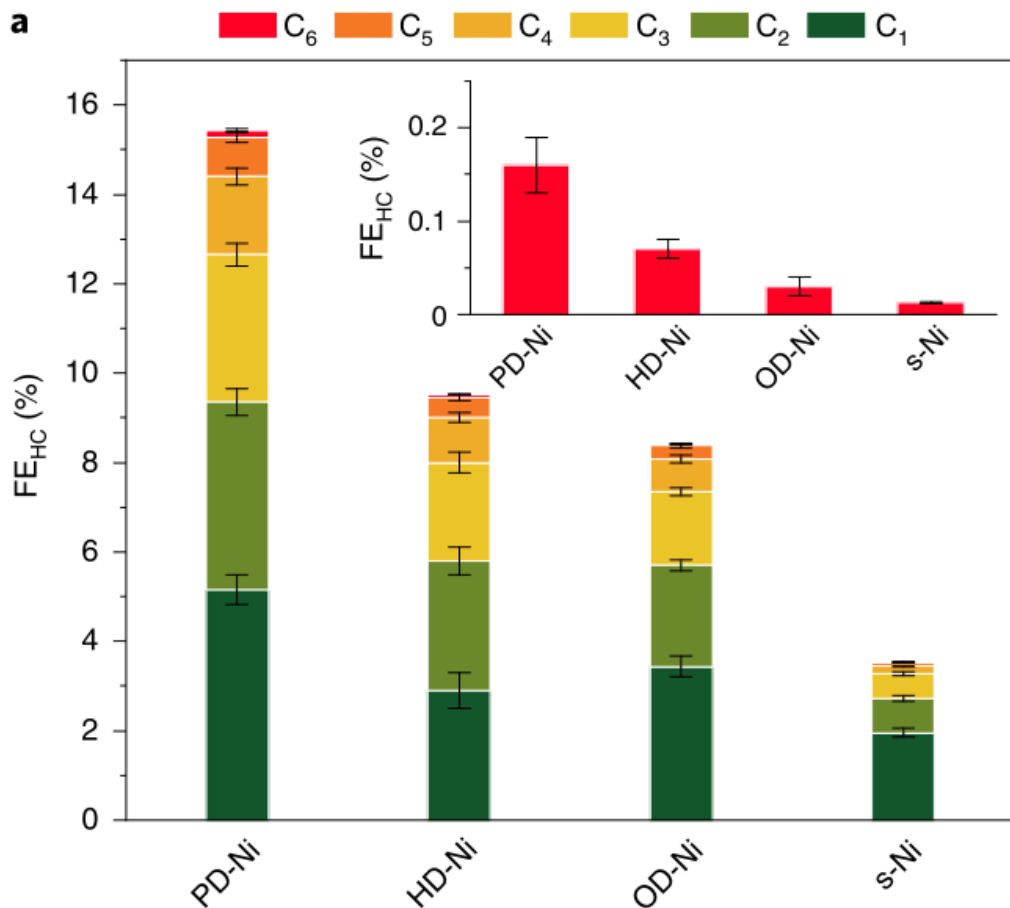
We have been engaged for some time in the study of the reduction of carbon dioxide at the dropping mercury cathode. On the basis of a large number of polarograms obtained by means of a Heyrovsky Polarograph of the Sargent Co. with solutions of carbon dioxide in 0.1 molar tetramethylammonium chloride, 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

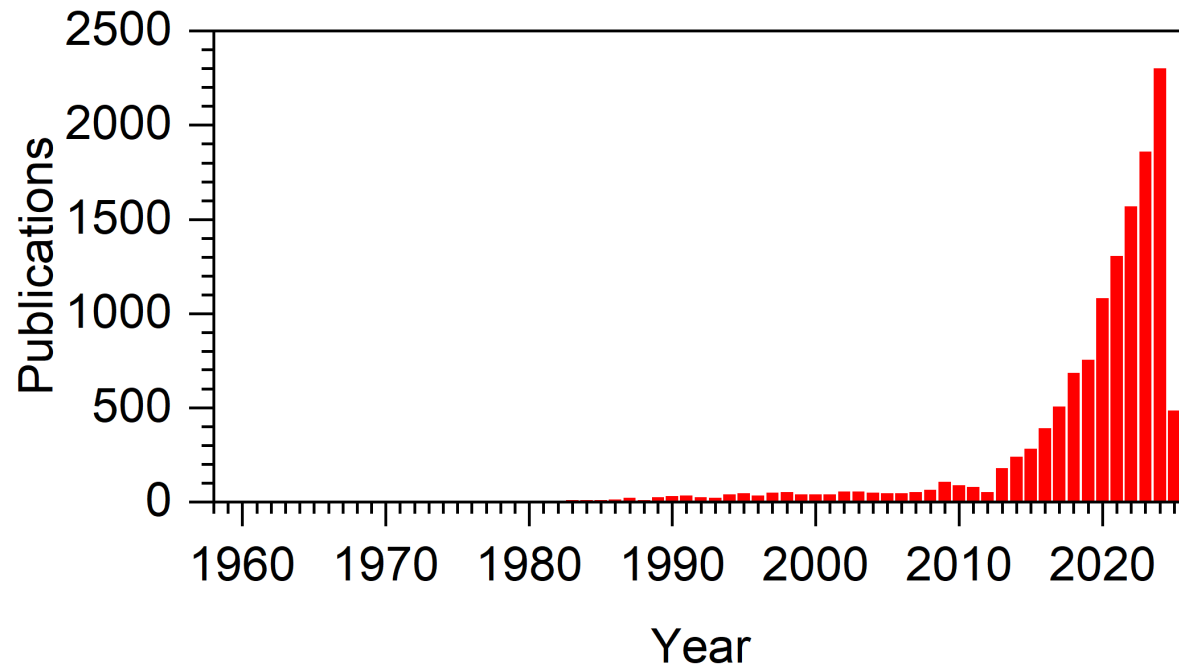
Preview

Download

The limit of current repositories

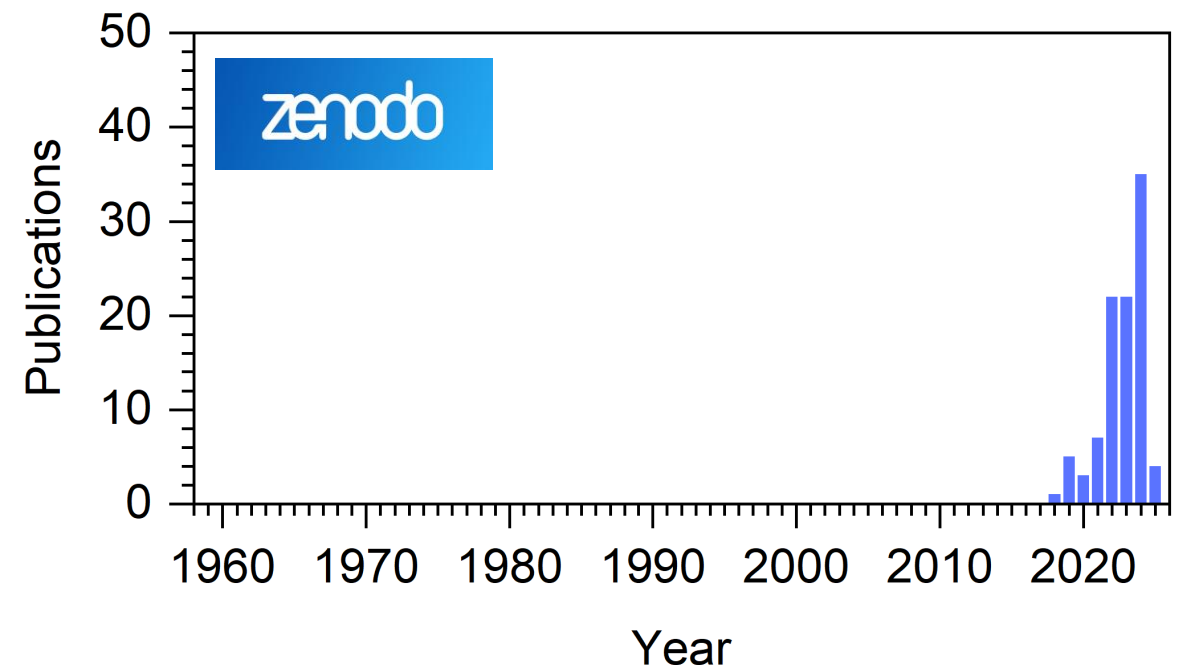
Keywords:

Electrochemical CO₂ reduction

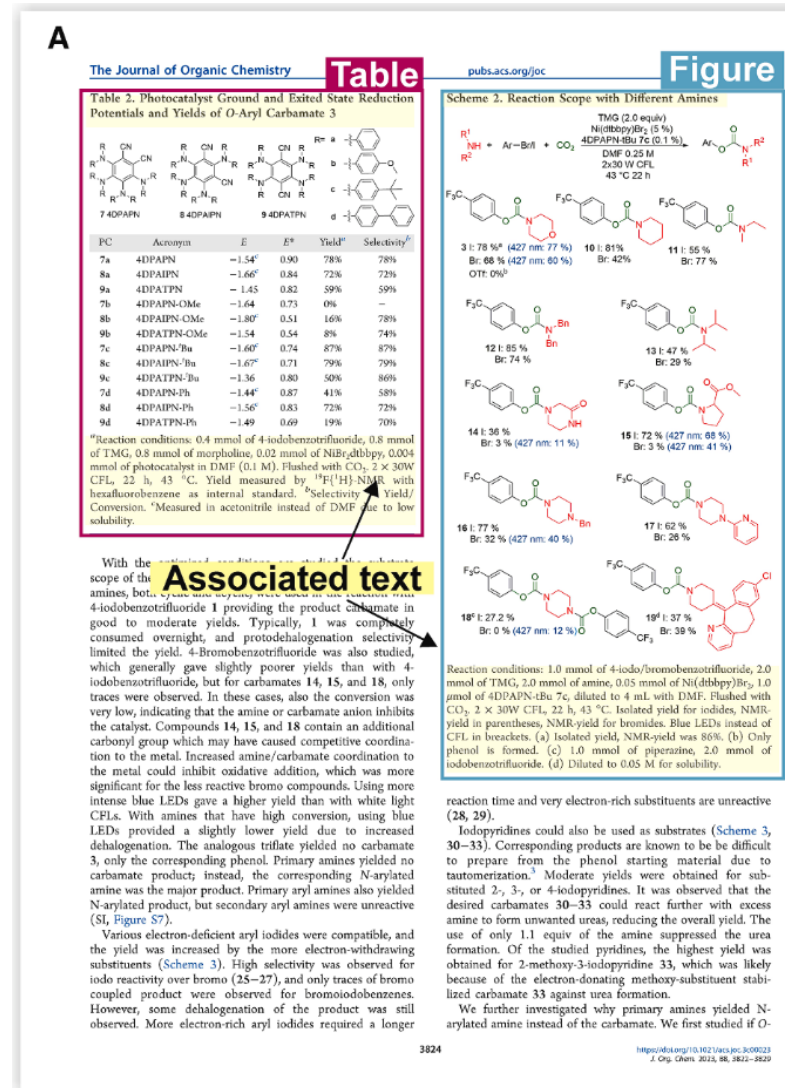
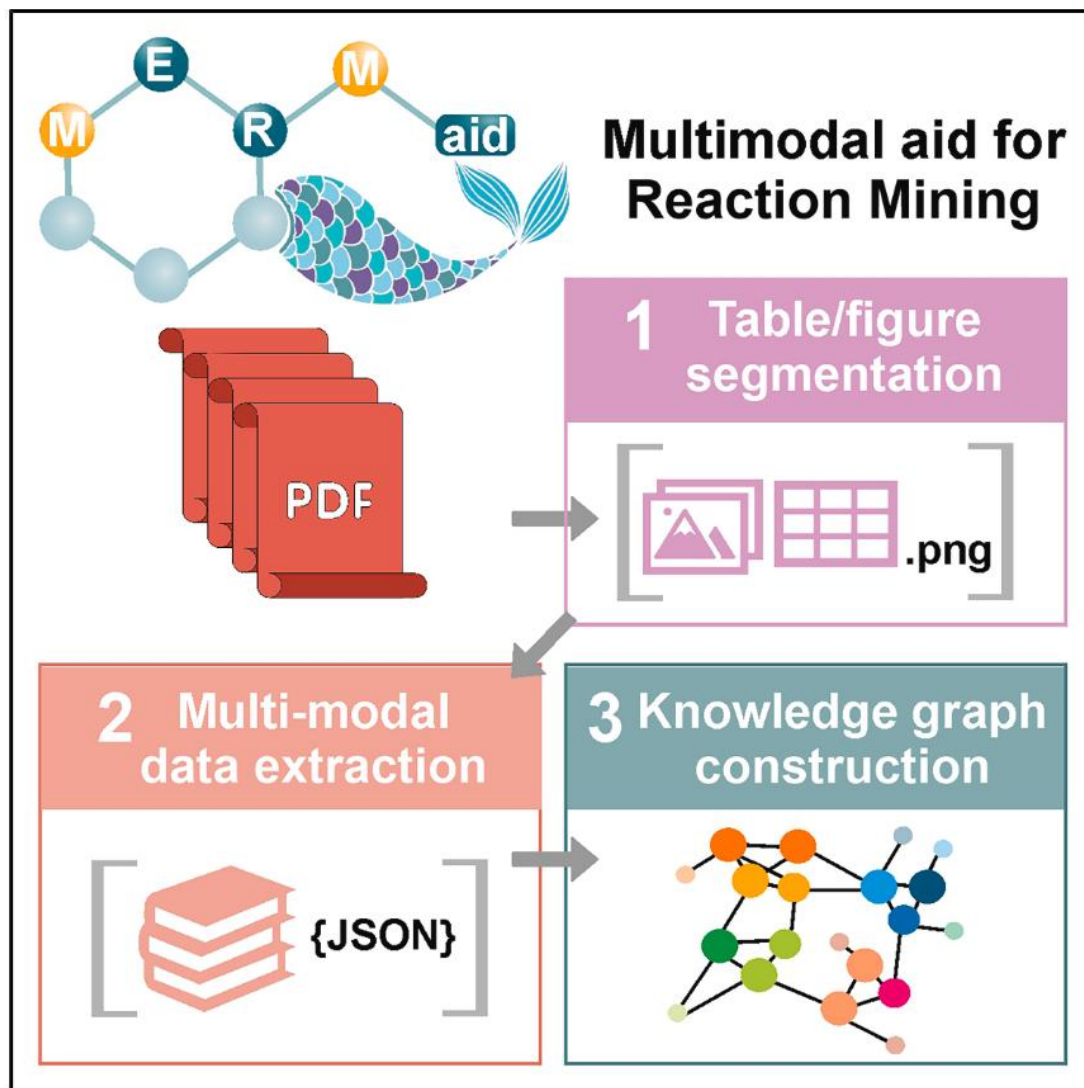


Keywords:

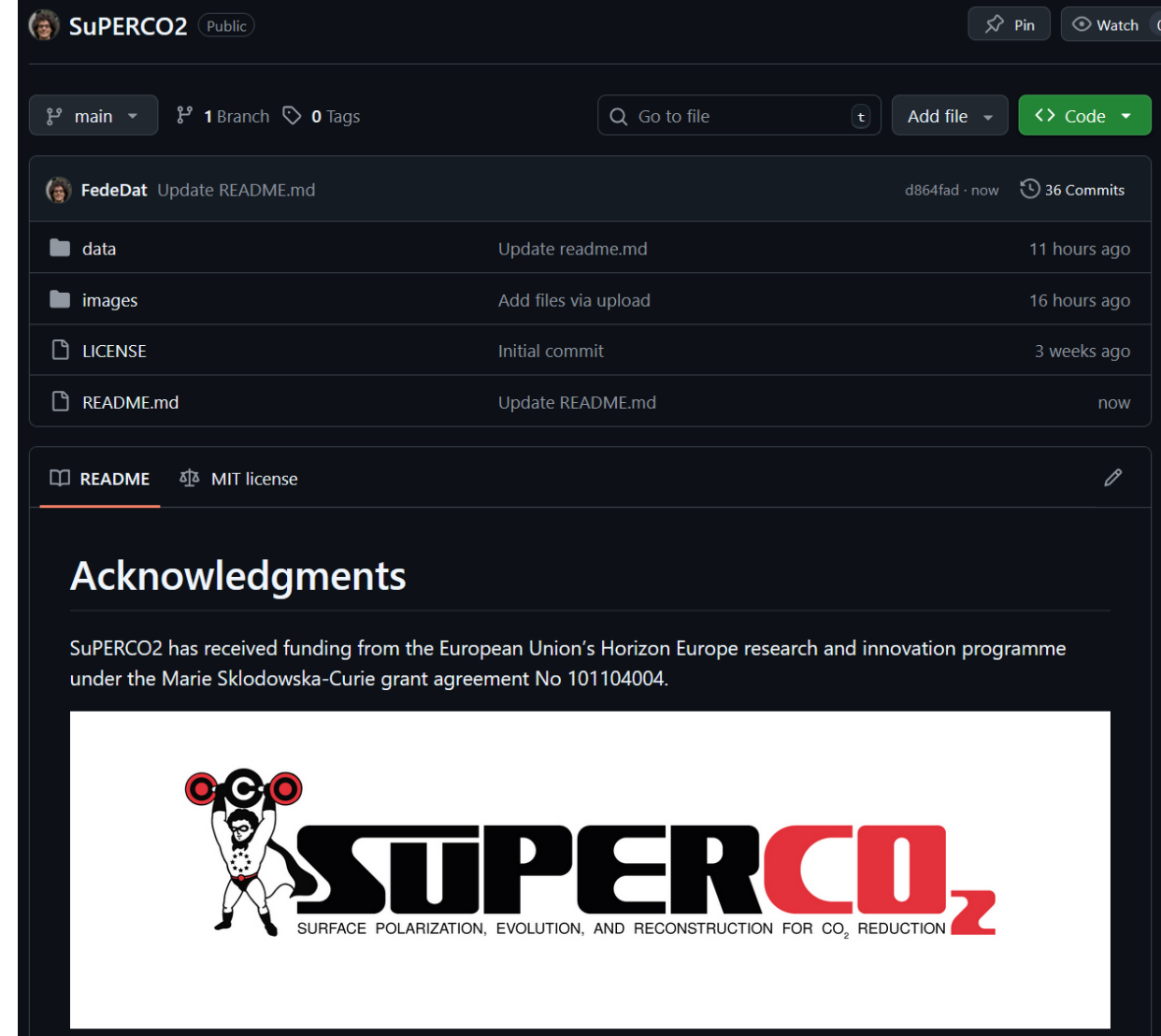
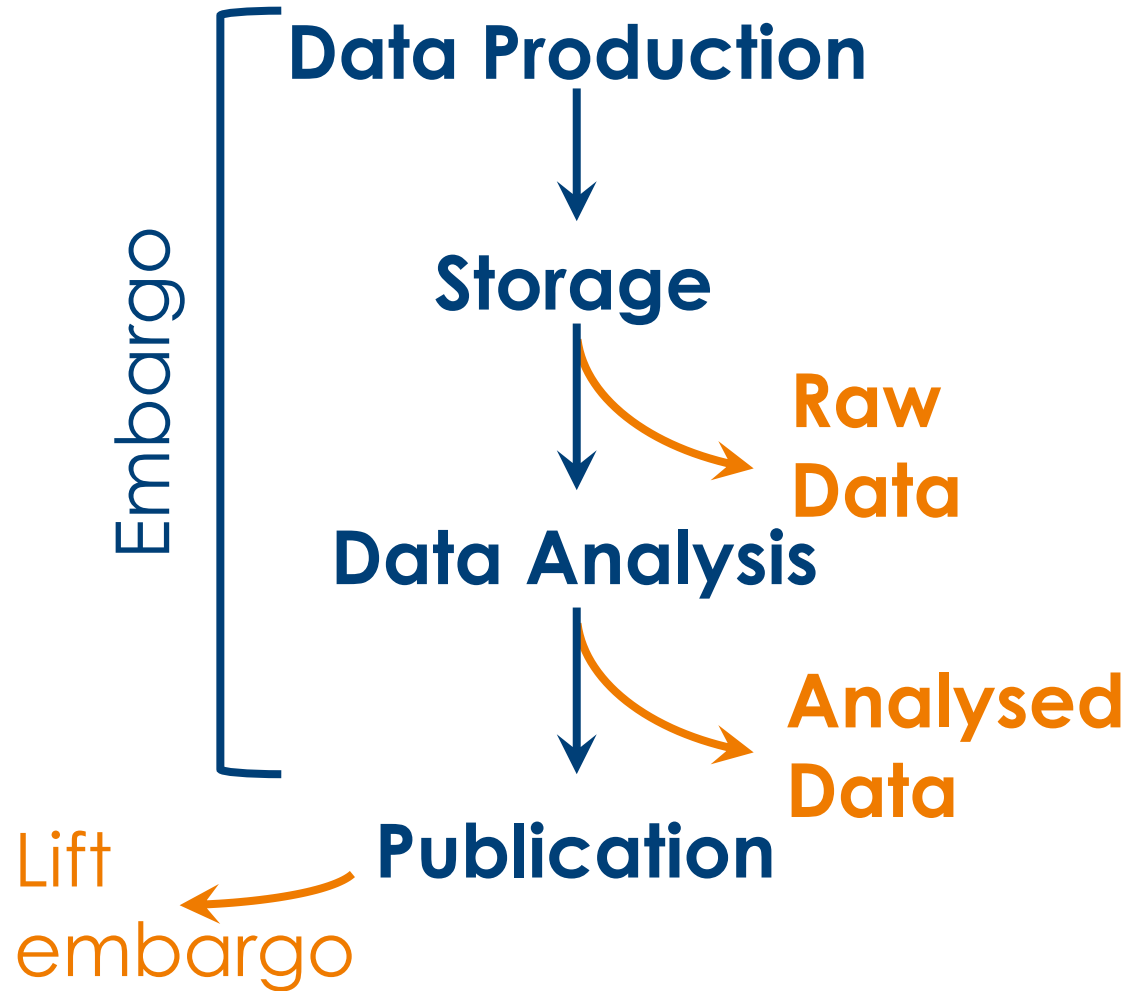
Electrochemical CO₂ reduction,
Zenodo



AI-based solution: MERmaid



The SuPERCO2 concept





Synthesis

Repository: github.com/FedeDat/SuPERCO2

item: data/CuO/synthesis

Internal logbook

Precursors

1	Precursor	CAS number	C (M)	C (mol)	V (mL)	m (g)	Molar mass (g/mol)
2	CuCl ₂ 2H ₂ O	10125-13-0	0.02	0.00	250.00	0.60	134.45
3	NaOH	1310-73-2	6.08	0.30	50.00	12.17	40.00
4	milliQ		55.50	13.88	250.00	249.96	18.02

Protocol

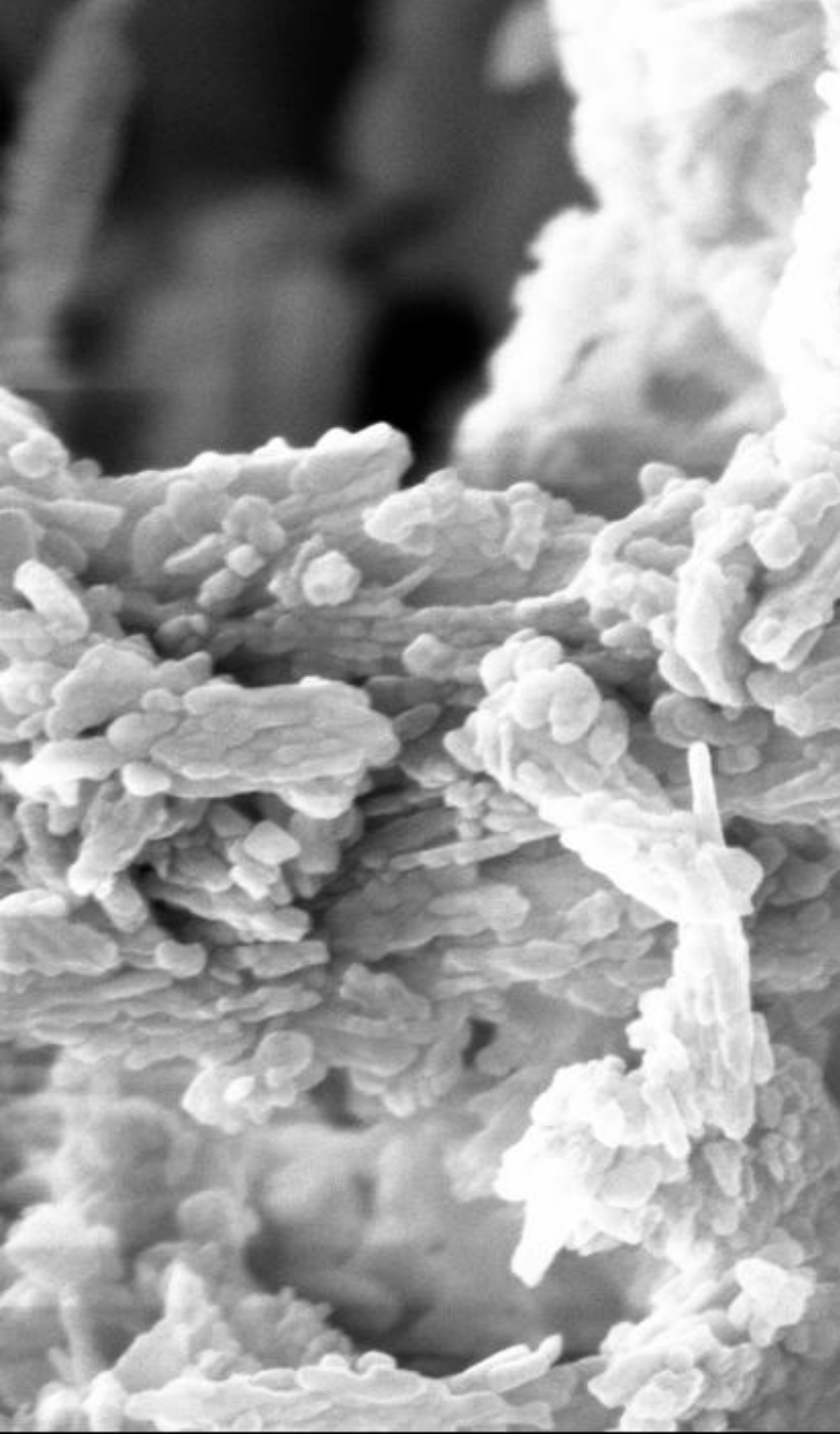
Equipment

- A Peristaltic Pump was used to control the flow rate during addition of the sodium hydroxide solutions.
- Magnetic stirrers were essential for ensuring uniform mixing of the reactants, thus they were employed whenever a solution was involved.
- A centrifuge was necessary for separating the precipitated particles from the reaction mixture

Synthesis procedure

Synthesis carried out on 03/07/2024.

The synthesis of CuO was a simple water-based precipitation, followed by calcination (Table 1). 250 mL of milliQ water were added to 600 mg of copper chloride (CuCl₂·2H₂O) under vigorous magnetic stirring to achieve a 20 mM solution. Then a solution of 6 M NaOH was formed by mixing 12.1674 g of NaOH into a 50 mL volumetric flask. After this, half the NaOH solution (25 mL) was added to the copper chloride solution via the peristaltic pump with a fixed flow rate of 25 mL per minute. As soon as NaOH was inserted, a light blue copper hydroxide precipitate appeared (Figure 7). The precipitate was centrifuged at 5000 rpm and cleaned with milliQ water. Four additional cleaning/centrifugation cycles were carried out, where the solvent was promptly removed. Then the resulting precipitate was transferred to a standard oven at 200 °C for 2 hours where the hydroxide decomposed to CuO.



Characterization

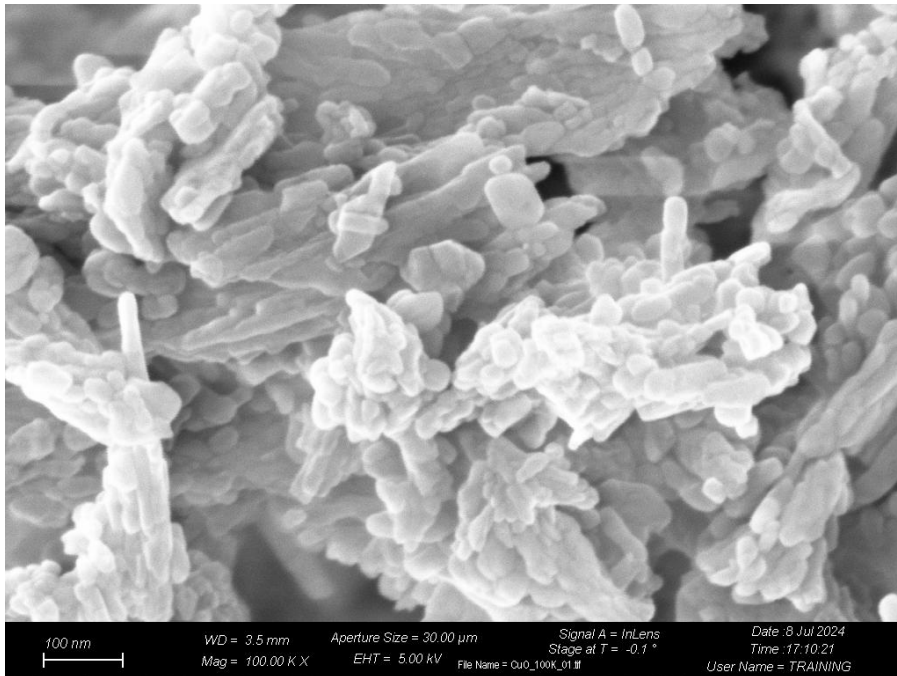
Repository: github.com/FedeDat/SuPERCO2

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 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

1	Cu	Fe	O	C	Na	Cl
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/CuO/electrochemistry

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

Work in progress
for liquid products

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)))
```

Chronopotentiometry

Post-processing (CP)

Preview Code Blame 918 lines (918 loc) · 38.8 KB				
Q Search this file				
1	Area (cm2)	1.33		
2	Electrolyte	KOH		
3	Conc (M)	1		
4	Loading (mg/cm2)	0.6		
5	pH initial vs final	14	13.73	
6	ZIR (Ohm)	4.945	4.801	
7	Ref	Ag/AgCl		
8	V			
9	CO2 Flow Rate input vs output (mL/min)	10	8.7	
10	Electrolyte Flow Rate (mL/min)	1		
11	Type	CP		
12				
13	t (s)	i (mA)	Ewe (V)	Ecell (V)
14	29.99979924	0	-0.35689005	-0.12803969

Post-processing (GC, HPLC)

1	H2	O2	N2	CH4
2	33107.7722			131.4632
3	43991.6529			141.6516
4	52639.5361			148.6601
5	60080.1305			143.4467
6	66500.9825			141.7904
7	71986.6802			139.0178
8	71553.3711			0

1	HCOO-	AcO-
2	3.6034	0.0714
3		
4		
5		
6		
7		

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
CP

Next steps

Provide a tutorial

Find sparring partner

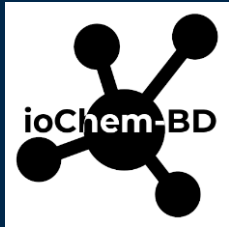
Let's try the power
of SUPERCO₂!



Politecnico
di Torino

Acknowledgements

Inspiration



M.Eng. Moisés Álvarez
(ICIQ)

Exp collaboration

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

Prof. Simelys Hernández
(lab Access)



**Politecnico
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Thank you!
Questions or feedbacks?

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



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