

# fase4\_1

May 21, 2025

```
[8]: import pandas as pd

df_real = pd.read_csv("/home/darkcover/Documentos/Gan/Data/
↳ujindoorsubset_building1_floor2.csv")      # Dados reais de treino

df_real.describe()
```

```
[8]:
```

	WAP001	WAP002	WAP003	WAP004	WAP005	WAP006	WAP007	\
count	1396.0	1396.0	1396.0	1396.0	1396.0	1396.000000	1396.0	
mean	-110.0	-110.0	-110.0	-110.0	-110.0	-109.584527	-110.0	
std	0.0	0.0	0.0	0.0	0.0	2.912958	0.0	
min	-110.0	-110.0	-110.0	-110.0	-110.0	-110.000000	-110.0	
25%	-110.0	-110.0	-110.0	-110.0	-110.0	-110.000000	-110.0	
50%	-110.0	-110.0	-110.0	-110.0	-110.0	-110.000000	-110.0	
75%	-110.0	-110.0	-110.0	-110.0	-110.0	-110.000000	-110.0	
max	-110.0	-110.0	-110.0	-110.0	-110.0	-88.000000	-110.0	

  

	WAP008	WAP009	WAP010	...	WAP520	LONGITUDE	\
count	1396.000000	1396.0	1396.000000	...	1396.0	1396.000000	
mean	-109.197708	-110.0	-109.678367	...	-110.0	-7486.581784	
std	3.946364	0.0	2.181149	...	0.0	45.101037	
min	-110.000000	-110.0	-110.000000	...	-110.0	-7571.093400	
25%	-110.000000	-110.0	-110.000000	...	-110.0	-7520.755800	
50%	-110.000000	-110.0	-110.000000	...	-110.0	-7491.030634	
75%	-110.000000	-110.0	-110.000000	...	-110.0	-7443.877677	
max	-80.000000	-110.0	-92.000000	...	-110.0	-7408.695251	

  

	LATITUDE	FLOOR	BUILDINGID	SPACEID	RELATIVEPOSITION	\
count	1.396000e+03	1396.0	1396.0	1396.000000	1396.000000	
mean	4.864879e+06	2.0	1.0	117.111748	1.704155	
std	3.501884e+01	0.0	0.0	83.279968	0.456585	
min	4.864810e+06	2.0	1.0	2.000000	1.000000	
25%	4.864859e+06	2.0	1.0	17.000000	1.000000	
50%	4.864873e+06	2.0	1.0	107.000000	2.000000	
75%	4.864893e+06	2.0	1.0	204.000000	2.000000	
max	4.864959e+06	2.0	1.0	217.000000	2.000000	

	USERID	PHONEID	TIMESTAMP
count	1396.000000	1396.000000	1.396000e+03
mean	5.461318	17.108883	1.371721e+09
std	3.304272	5.297423	9.536837e+03
min	2.000000	8.000000	1.371714e+09
25%	2.000000	14.000000	1.371714e+09
50%	4.000000	18.000000	1.371715e+09
75%	9.000000	23.000000	1.371735e+09
max	10.000000	23.000000	1.371738e+09

[8 rows x 529 columns]

```
[9]: df_real.columns
```

```
[9]: Index(['WAP001', 'WAP002', 'WAP003', 'WAP004', 'WAP005', 'WAP006', 'WAP007',
          'WAP008', 'WAP009', 'WAP010',
          ...,
          'WAP520', 'LONGITUDE', 'LATITUDE', 'FLOOR', 'BUILDINGID', 'SPACEID',
          'RELATIVEPOSITION', 'USERID', 'PHONEID', 'TIMESTAMP'],
          dtype='object', length=529)
```

```
[11]: df_generated = pd.read_csv("/home/darkcover/Documentos/Gan/Data/
↳df_generated_pseudo.csv") # Vetores pseudo-rotulados gerados
df_generated.describe()
```

```
[11]:
```

	WAP001	WAP002	WAP003	WAP004	WAP005 \
count	40000.000000	40000.000000	40000.000000	40000.000000	40000.000000
mean	-52.131000	-97.930100	-60.247600	-89.294925	-63.500625
std	4.353879	5.075836	4.681441	4.961357	3.109902
min	-64.000000	-110.000000	-71.000000	-110.000000	-72.000000
25%	-55.000000	-101.000000	-64.000000	-93.000000	-66.000000
50%	-53.000000	-98.000000	-61.000000	-89.000000	-64.000000
75%	-49.000000	-94.000000	-57.000000	-86.000000	-62.000000
max	-40.000000	-82.000000	-40.000000	-78.000000	-46.000000

  

	WAP006	WAP007	WAP008	WAP009	WAP010 \
count	40000.000000	40000.000000	40000.000000	40000.000000	40000.000000
mean	-94.420700	-70.493825	-97.648175	-93.546025	-67.002625
std	5.060805	2.509307	5.459940	3.898392	4.002002
min	-110.000000	-79.000000	-110.000000	-110.000000	-78.000000
25%	-98.000000	-72.000000	-101.000000	-96.000000	-70.000000
50%	-94.000000	-71.000000	-97.000000	-93.000000	-67.000000
75%	-91.000000	-69.000000	-93.000000	-91.000000	-64.000000
max	-80.000000	-60.000000	-85.000000	-82.000000	-50.000000

  

	LONGITUDE	LATITUDE
count	40000.000000	40000.000000

mean	10.037707	9.903981
std	1.724494	1.243719
min	3.990654	6.160164
25%	8.866893	8.976322
50%	10.104205	9.870904
75%	11.266562	10.831815
max	15.545594	13.339489

```
[12]: df_generated.columns
```

```
[12]: Index(['WAP001', 'WAP002', 'WAP003', 'WAP004', 'WAP005', 'WAP006', 'WAP007',
           'WAP008', 'WAP009', 'WAP010', 'LONGITUDE', 'LATITUDE'],
          dtype='object')
```

```
[15]: import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt

# Phase 4: Reproduce Table 2 and Figure 6
# -----
# 1. Load datasets
# -----
df_real = pd.read_csv("/home/darkcover/Documentos/Gan/Data/
↳ujindoorsubset_building1_floor2.csv")      # Dados reais de treino
df_generated = pd.read_csv("/home/darkcover/Documentos/Gan/Data/
↳df_generated_pseudo.csv") # Vetores pseudo-rotulados gerados
df_test = pd.read_csv('/home/darkcover/Documentos/Gan/Data/df_test.csv') #
↳Conjunto de teste fixo

# 2. Determine WAP columns common to all datasets
# -----
wap_real = {col for col in df_real.columns if col.startswith('WAP')}
wap_gen = {col for col in df_generated.columns if col.startswith('WAP')}
wap_test = {col for col in df_test.columns if col.startswith('WAP')}
# Use only the WAPs present in real, generated, and test
wap_columns = sorted(list(wap_real & wap_gen & wap_test))

# 2.1 Prepare features (X) and labels (y)
# -----
X_real = df_real[wap_columns].values.astype(np.float32)
y_real = df_real[['LONGITUDE', 'LATITUDE']].values.astype(np.float32)

X_gen_full = df_generated[wap_columns].values.astype(np.float32)
y_gen_full = df_generated[['LONGITUDE', 'LATITUDE']].values.astype(np.float32)

X_test = df_test[wap_columns].values.astype(np.float32)
```

```

y_test      = df_test[['X', 'Y']].values.astype(np.float32)

# 3. Model builder
def build_model(input_dim, learning_rate):
    model = tf.keras.Sequential([
        tf.keras.layers.Dense(128, activation='relu', input_shape=(input_dim,)),
        tf.keras.layers.Dense(64, activation='relu'),
        tf.keras.layers.Dense(2)
    ])
    optimizer = tf.keras.optimizers.Adam(learning_rate=learning_rate)
    model.compile(optimizer=optimizer,
                  loss='mse')
    return model

# 4. Training and evaluation function
# -----

def train_and_evaluate(X_train, y_train, X_eval, y_eval, config,
    ↪return_errors=False):
    """
    Treina uma DNN e avalia o erro de localização.
    config: dict com keys 'epochs', 'batch_size', 'learning_rate'
    return_errors: se True, retorna array de erros ponta a ponta.
    """
    model = build_model(input_dim=X_train.shape[1],
    ↪learning_rate=config['learning_rate'])
    model.fit(X_train, y_train,
              epochs=config['epochs'],
              batch_size=config['batch_size'],
              verbose=0)
    y_pred = model.predict(X_eval)
    # erro Euclidiano por ponto
    errs = np.linalg.norm(y_pred - y_eval, axis=1)
    if return_errors:
        return errs
    return errs.mean()

# 5. Configurações de experimento (conforme artigo)
# -----
configs = {
    'Supervised(1000,100)': {'epochs': 250, 'batch_size': 100,
    ↪'learning_rate': 0.01},
    'Supervised(2000,1100)': {'epochs': 250, 'batch_size': 100, 'learning_rate':
    ↪ 0.01}
}
# Adicionando configurações Selective-SS-GAN para diferentes ms
for ms in [100, 500, 1000]:

```

```

key = f'Selective-SS-GAN(1000,100,{ms})'
configs[key] = {'epochs': 250, 'batch_size': 100, 'learning_rate': 0.01}

# 6. Executar experimentos
# -----
results = {}

# Baseline 1: Supervised(1000,100)
X_train = X_real[:1000]
y_train = y_real[:1000]
results['Supervised(1000,100)'] = train_and_evaluate(
    X_train, y_train, X_test, y_test, configs['Supervised(1000,100)']
)

# Baseline 2: Supervised(2000,1100) duplicando reais
X2 = np.vstack([X_train, X_train])
y2 = np.vstack([y_train, y_train])
results['Supervised(2000,1100)'] = train_and_evaluate(
    X2, y2, X_test, y_test, configs['Supervised(2000,1100)']
)

# Selective-SS-GAN experiments
for ms in [100, 500, 1000]:
    key = f'Selective-SS-GAN(1000,100,{ms})'
    X_mix = np.vstack([X_train, X_gen_full[:ms]])
    y_mix = np.vstack([y_train, y_gen_full[:ms]])
    results[key] = train_and_evaluate(
        X_mix, y_mix, X_test, y_test, configs[key]
    )

# 7. Reproduzir Tabela 2
# -----
table2 = pd.DataFrame(
    [(method, err) for method, err in results.items()],
    columns=['Method', 'MeanError']
)
print("Tabela 2:\n", table2)

# 8. Gerar Figura 6 (CDF de erros)
# -----
errs_base = train_and_evaluate(X_train, y_train, X_test, y_test,
                                configs['Supervised(1000,100)'],
                                return_errors=True)
errs_select = train_and_evaluate(
    np.vstack([X_train, X_gen_full[:1000]]),
    np.vstack([y_train, y_gen_full[:1000]]),
    X_test, y_test,

```

```

        configs['Selective-SS-GAN(1000,100,1000)'],
        return_errors=True
    )

plt.figure(figsize=(6,4))
for errs, label in [(errs_base, 'Supervised'), (errs_select,
↪ 'Selective-SS-GAN')]:
    sorted_e = np.sort(errs)
    cdf = np.arange(len(sorted_e)) / float(len(sorted_e))
    plt.plot(sorted_e, cdf, label=label)

plt.xlabel('Localization error (m)')
plt.ylabel('CDF')
plt.title('Figure 6. Error CDF comparison')
plt.legend()
plt.grid(True, linestyle='--', linewidth=0.5)
plt.tight_layout()
plt.show()

```

/home/darkcover/.cache/pypoetry/virtualenvs/gan-oPyfrVEv-  
py3.12/lib/python3.12/site-packages/keras/src/layers/core/dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```

25/25          0s 4ms/step
25/25          0s 3ms/step
25/25          0s 3ms/step
25/25          0s 2ms/step
25/25          0s 3ms/step

```

Tabela 2:

	Method	MeanError
0	Supervised(1000,100)	3903426.25
1	Supervised(2000,1100)	3899272.25
2	Selective-SS-GAN(1000,100,100)	3806442.25
3	Selective-SS-GAN(1000,100,500)	3830636.25
4	Selective-SS-GAN(1000,100,1000)	3849217.25

/home/darkcover/.cache/pypoetry/virtualenvs/gan-oPyfrVEv-  
py3.12/lib/python3.12/site-packages/keras/src/layers/core/dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```

25/25          0s 4ms/step
25/25          0s 2ms/step

```

Figure 6. Error CDF comparison

