TASK GABUNGAN 3.1 – 3.3

1. Load data per client

Baca tiga CSV (Dinsos, Dukcapil, Kemenkes) jadi DataFrame.

2. Buat Label biner local

Dinsos:

Dukcapil:

```
def label_dukcapil(row):
    u = row.get("umur", np.nan)
    sp = row.get("status_pekerjaan", "")
    st = row.get("status_pernikahan", "")

if pd.isna(u): u = 0

# Layak (1)

if u > 65:
    return 1

if sp in ("pengangguran", "buruh", "pekerja informal"):
    return 1

if (sp == "wirausaha") and (u > 55):
    return 1

if (st in {"cerai", "janda", "duda"}) and (sp != "pegawai tetap"):
    return 1

if (sp in {"pegawai tetap"}) and (25 <= u <= 55) and (st == "menikah"):
    return 0

if (u < 25) and (sp == "wirausaha"):
    return 0

return 0

return 0</pre>
```

Kemenkes:

```
def label_kemenkes(row):
    rp = row.get("riwayat_penyakit", "")
    sg = row.get("status_gizi", "")
    t = row.get("tinggi_cm", np.nan)
    b = row.get("berat_kg", np.nan)

bmi = None
    if pd.notna(t) and pd.notna(b) and t > 0:
        bmi = b / ((t/100.0) ** 2)

# Layak (1)
    if rp in {"kronis", "jantung", "asma", "diabetes", "disabilitas"}:
        return 1
    if sg in {"kurang", "stunting", "gizi buruk"}:
        return 1
    if (bmi is not None) and (bmi < 18.5):
        return 1
    if (bmi is not None) and ((bmi < 17) or (bmi > 35)):
        return 1

# Tidak layak (0)
    return 0
```

Terapkan aturan domain per sumber untuk menghasilkan layak_subsidi (0/1) tanpa mengubah file asli.

```
# Terapkan aturan ke masing-masing client (tanpa mengubah file asal)

dinsos_lab = dinsos.copy()

dukcapil_lab = dukcapil.copy()

kemenkes_lab = kemenkes.copy()

dinsos_lab["layak_subsidi"] = dinsos_lab.apply(label_dinsos, axis=1)

dukcapil_lab["layak_subsidi"] = dukcapil_lab.apply(label_dukcapil, axis=1)

kemenkes_lab["layak_subsidi"] = kemenkes_lab.apply(label_kemenkes, axis=1)
```

3. Samakan skema fitur lintas-klien

- Tambah fitur turunan **BMI** bila ada tinggi/berat.
- Bentuk **vocabulary global** untuk kolom kategorikal, lalu **one-hot** agar urutan & jumlah kolom konsisten.
- Imputasi numerik (median) untuk NaN.

```
def union_categories(series_list):

cats = set()

for s in series_list:

if s is not None:

vals = s.dropna().unique().tolist()

for v in vals:

if isinstance(v, str):

cats.add(v)

return sorted(list(cats))

global_vocabs = {

"kondisi_rumah": union_categories([dinsos_lab_get("kondisi_rumah"]),

"status_pekerjaan": union_categories([dukcapil_lab_get("status_pekerjaan")]),

"status_pernikahan": union_categories([dukcapil_lab_get("status_pernikahan")]),

"istatus_pernikahan": union_categories([dukcapil_lab_get("status_pernikahan")]),

"riwayat_penyakit": union_categories([kemenkes_lab_get("riwayat_penyakit")]),

"status_gizi": union_categories([kemenkes_lab_get("riwayat_penyakit")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")]),

"status_gizi": union_categories([kemenkes_lab_get("riwayat_penyakit")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")]),

"status_gizi": union_categories([kemenkes_lab_get("status_gizi")],

"status_gizi": union_categories([kemen
```

4. Scaling global yang konsisten

Gabungkan semua fitur \rightarrow hitung min/max global \rightarrow terapkan min-max scaling ke tiap klien.

```
# Min-max scaling GLOBAL (gabungkan semua agar konsisten)
all_X = pd.concat((X_dinsos_raw, X_dukcapil_raw, X_kemenkes_raw], axis=0)
mins = all_X.min(axis=0)
maxs = all_X.max(axis=0)
rng = (maxs - mins).replace(0, 1.0)

def scale_like_global(X):
return (X - mins) / rng

X_dinsos = scale_like_global(X_dinsos_raw).fillna(0.0)
X_kemenkes = scale_like_global(X_dukcapil_raw).fillna(0.0)

X_kemenkes = scale_like_global(X_kemenkes_raw).fillna(0.0)

FEATURE_COLS = list(X_dinsos.columns) # sama untuk semua client
```

5. Siapkan dataset Tensors

Konversi tiap (X, y) klien ke tf.data.Dataset (shuffle + batch).

```
def df_to_tf_dataset(features_df, y_array, batch_size=32, shuffle=True):
    X = features_df.values.astype("float32")
    y = y_array.astype("float32").reshape(-1, 1)
    ds = tf.data.Dataset.from_tensor_slices((X, y))
    if shuffle:
        ds = ds.shuffle(buffer_size=len(y), reshuffle_each_iteration=True)
    ds = ds.batch(batch_size)
    return ds

client_ds = [
    df_to_tf_dataset(X_dinsos, y_dinsos, batch_size=32, shuffle=True),
    df_to_tf_dataset(X_dukcapil, y_dukcapil, batch_size=32, shuffle=True),
    df_to_tf_dataset(X_kemenkes, y_kemenkes, batch_size=32, shuffle=True),
]
```

6. Bangun model Keras sederhana

MLP: Dense(64) \rightarrow Dense(32) \rightarrow Dense(1, sigmoid) untuk klasifikasi biner.

```
def create_keras_model(input_dim):
    inputs = tf.keras.Input(shape=(input_dim,))
    x = tf.keras.layers.Dense(64, activation="relu")(inputs)
    x = tf.keras.layers.Dense(32, activation="relu")(x)
    outputs = tf.keras.layers.Dense(1, activation="sigmoid")(x)
    model = tf.keras.Model(inputs, outputs)
    return model
```

7. Wrap ke TFF

Gunakan tff.learning.models.from_keras_model dengan input_spec, BinaryCrossentropy, dan metrik BinaryAccuracy.

8. Konfigurasi Federated Averaging

Build FedAvg dengan optimizer klien (SGD lr=0.001) dan server (SGD lr=1.0).

```
federated_averaging = tff.learning.algorithms.build_weighted_fed_avg(
    model_fn,
    client_optimizer_fn=tff.learning.optimizers.build_sgdm(learning_rate=0.001),
    server_optimizer_fn=tff.learning.optimizers.build_sgdm(learning_rate=1.0)
)
state = federated_averaging.initialize()
```

9. Training federated

Inisialisasi state \rightarrow loop beberapa ronde \rightarrow tiap ronde: latih di klien, agregasi ke server, update model global, dan log loss & binary_accuracy.

Menggunakan 5 round:

Menggunakan 10 round: