

Задание 1

Запустить модель Stable Diffusion (в Jupyter Notebook от Anaconda или в Colab), придумать свой вариант промпта, связанный с вашим знаком зодиака, сгенерировать изображения, соответствующие промпту, прислать результат через ТУИС.

Задание 2

Загрузите из keras.datasets набор данных California Housing price regression dataset (https://keras.io/api/datasets/california_housing/), обучите нейронную сеть прогнозировать медианную цену домов в зависимости от количества комнат в доме, визуализируйте процесс обучения.

I'll put all my labs here https://github.com/hallame/deep_learning

Task01

```
!pip -q install -U keras-cv keras silence-tensorflow matplotlib
```

```
Preparing metadata (setup.py) ... done
```

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```
Building wheel for silence-tensorflow (setup.py) ... done
```

```
!pip install ipywidgets
```

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Downloading jedi-0.19.2-py2.py3-none-any.whl (1.6 MB)
```

1.6/1.6 MB 19.9 MB/s eta 0:00:00

Installing collected packages: jedi
Successfully installed jedi-0.19.2

```
#import
import os
os.environ["KERAS_BACKEND"] = "tensorflow"

from silence_tensorflow import silence_tensorflow
silence_tensorflow()

import time, random, numpy as np, matplotlib.pyplot as plt

import keras          # Keras 3
import keras_cv        # KerasCV (Stable Diffusion)
import tensorflow as tf # ok après silence

#for ui
import ipywidgets as w
from IPython.display import display, clear_output

# Repro & outings
random.seed(42); np.random.seed(42)
os.makedirs("outputs/sd", exist_ok=True)

print("TF:", tf.__version__, "| Keras:", keras.__version__, "| KerasCV:", keras_cv.__version__)
```

TF: 2.19.0 | Keras: 3.12.0 | KerasCV: 0.9.0

```
#helpers utilities
def plot_images(images, title=None, figsize=(18, 6)):
    plt.figure(figsize=figsize)
    for i in range(len(images)):
        ax = plt.subplot(1, len(images), i + 1)
        plt.imshow(images[i])
        plt.axis("off")
    if title:
        plt.suptitle(title, fontsize=14)
    plt.show()
```

```
#model

import gc
import tensorflow as tf
import keras
import keras_cv

def reset_tf():
    """Session cleaning TF/Keras ."""
    keras.backend.clear_session()
    tf.keras.backend.clear_session()
    gc.collect()

_SD = {} # cache by size

def get_sd(size: int):
    """Retourn pipeline StableDiffusion for 'size'. """
    if size not in _SD:
        reset_tf()
        _SD[size] = keras_cv.models.StableDiffusion(
            img_width=size, img_height=size, jit_compile=False
        )
    return _SD[size]

# sd = keras_cv.models.StableDiffusion(
#     img_width=256,
#     img_height=256,
#     jit_compile=False # robust on CPU/Metal
# )
```

```
#zodiac signs
zodiac_style = {
    "aries": "ram made of stardust, fiery aura",
    "taurus": "majestic bull, emerald forest, grounded strength",
    "gemini": "twin spirits, mirrored neon, duality and motion",
    "cancer": "lunar crab, silver tides, protective shell",
    "leo": "golden lion, royal crown, sunlit glow",
    "virgo": "celestial maiden, wheat field, precise elegance",
    "libra": "ethereal scales, balance of light and shadow",
    "scorpio": "obsidian scorpion, midnight desert, mystic sting",
    "sagittarius": "celestial archer, comet trail, boundless horizon",
    "capricorn": "sea-goat on marble cliff, ancient wisdom",
    "aquarius": "water bearer of stars, electric waves, futurism",
    "pisces": "twin koi in nebula waters, dreamy currents"
}
```

```
def build_prompt(sign: str) -> str:
    return f"""{zodiac_style.get(sign, 'zodiac spirit')}, fantasy art,
high quality, highly detailed, elegant, sharp focus,
concept art, dramatic lighting, volumetric light, 8k render""".strip()

# Widgets
sign_dd = w.Dropdown(
    options=[
        ("♈ Aries", "aries"), ("♉ Taurus", "taurus"), ("♊ Gemini", "gemini"),
        ("♋ Cancer", "cancer"), ("♌ Leo", "leo"), ("♍ Virgo", "virgo"),
        ("♎ Libra", "libra"), ("♏ Scorpio", "scorpio"), ("♐ Sagittarius", "sagittarius"),
        ("♑ Capricorn", "capricorn"), ("♒ Aquarius", "aquarius"), ("♓ Pisces", "pisces"),
    ],
    value="aquarius",
    description="Signe:",
)
```

```
size_dd = w.Dropdown(options=[256, 512], value=256, description="Size:")
batch_slider = w.IntSlider(min=1, max=4, step=1, value=3, description="Batch:")
gen_btn = w.Button(description="Generate", button_style="success")
out = w.Output()
```

```
def on_generate(_):
    with out:
        clear_output()
        sign = sign_dd.value
        size = int(size_dd.value)
        batch = int(batch_slider.value)

        prompt = build_prompt(sign)
        print("Prompt:", prompt)
        print(f"Size: {size}x{size} | Batch: {batch}")

        # re-use un pipeline from cache
        sd = get_sd(size)

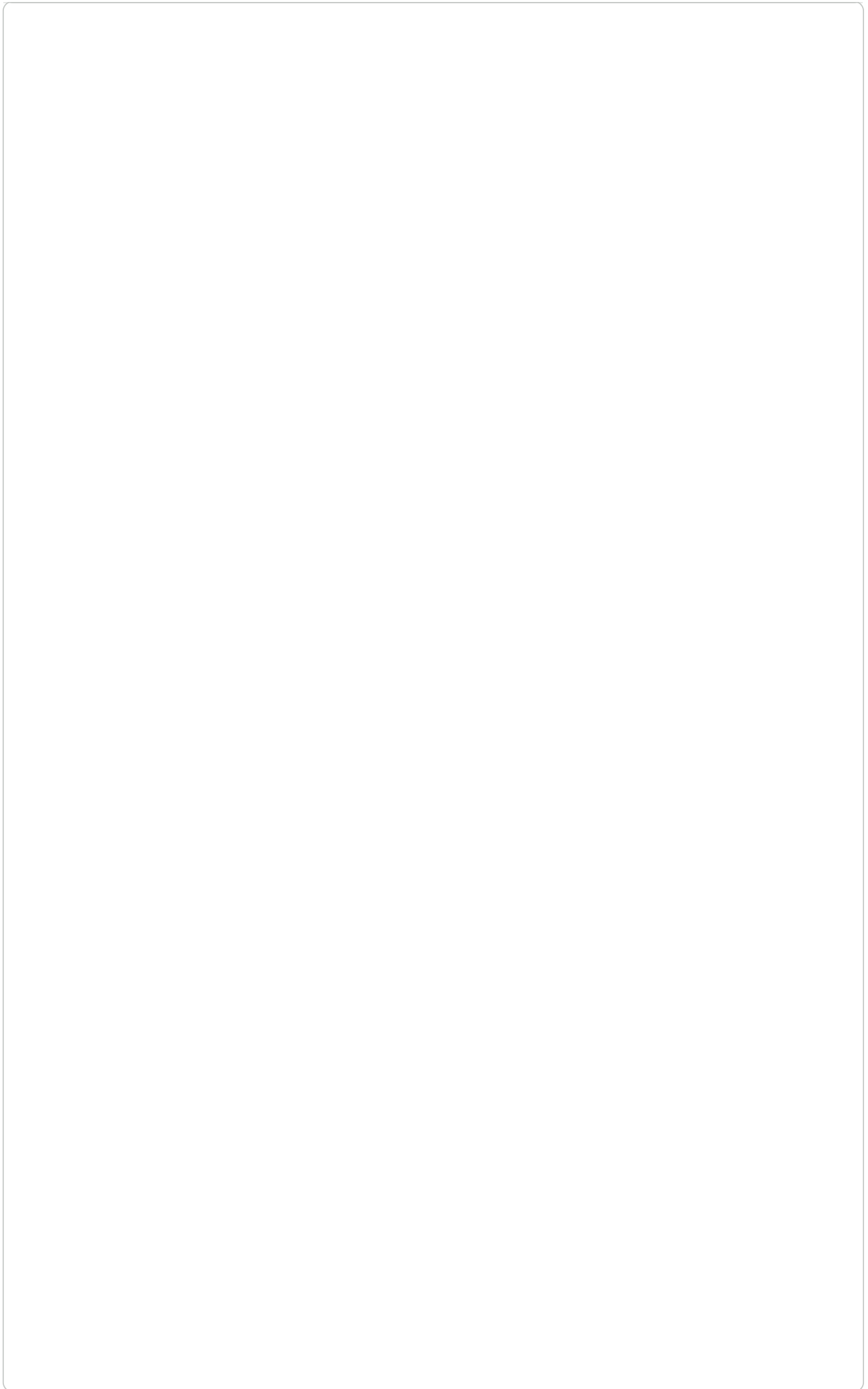
        t0 = time.time()
        images = sd.text_to_image(prompt, batch_size=batch)
        print(f"Time: {time.time()-t0:.1f}s")

        plot_images(images, title=f"Stable Diffusion - {sign.capitalize()}")

        saved = []
        for i, im in enumerate(images):
            path = f"outputs/sd/{sign}_{i+1}_{size}.png"
            plt.imsave(path, im)
            saved.append(path)
        print("Files saved:", saved)

gen_btn.on_click(on_generate)

ui = w.VBox([
    w.HBox([sign_dd, size_dd, batch_slider]),
    gen_btn,
    out
])
display(ui)
```



Signe:  Aquarius

Size: 256

Batch: 

3

Generate

Prompt: water bearer of stars, electric waves, futurism, fantasy art,
high quality, highly detailed, elegant, sharp focus,
concept art, dramatic lighting, volumetric light, 8k render

Size: 256x256 | Batch: 3

By using this model checkpoint, you acknowledge that its usage is subject to the terms of
the CreativeML Open RAIL-M license at [https://raw.githubusercontent.com/CompVis/stable-](https://raw.githubusercontent.com/CompVis/stable-diffusion/main/LICENSE)
[diffusion/main/LICENSE](https://raw.githubusercontent.com/CompVis/stable-diffusion/main/LICENSE)

Downloading data from

https://github.com/openai/CLIP/blob/main/clip/bpe_simple_vocab_16e6.txt.gz?raw=true

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[diffusion/resolve/main/kcv_encoder.h5](https://huggingface.co/fchollet/stable-diffusion/resolve/main/kcv_encoder.h5)








































































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1599651840/3439090152		12s	0us/step
1612144640/3439090152		12s	0us/step
1624088576/3439090152		12s	0us/step
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1655439360/3439090152		11s	0us/step
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1727283200/3439090152		11s	0us/step
1737981952/3439090152		11s	0us/step
1749385216/3439090152		10s	0us/step
1758478336/3439090152		10s	0us/step
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1818345472/3439090152		10s	0us/step
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1984299008/3439090152		9s	0us/step
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2068160512/3439090152		8s	0us/step
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2111045632/3439090152		8s	0us/step

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2236964864/3439090152	7s	0us/step
2242248704/3439090152	7s	0us/step
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Task 03

45/50 ————— 2s 434ms/step

46/50 ————— 1s 434ms/step

47/50 ————— 1s 434ms/step

48/50 ————— 0s 434ms/step

```
#import
import os
os.environ["KERAS_BACKEND"] = "tensorflow"

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

import tensorflow as tf
import keras
from keras import layers

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import mean_absolute_error, r2_score

# cut the logs TF
try:
    from silence_tensorflow import silence_tensorflow
    silence_tensorflow()
except Exception:
    pass

# Reproducibility
tf.keras.backend.clear_session()
np.random.seed(42)
tf.random.set_seed(42)

print("TF:", tf.__version__, " | Keras:", keras.__version__)
```

TF: 2.19.0 | Keras: 3.12.0

```
#uploading data
USE_SKLEARN_FALLBACK = False
try:
    from keras.datasets import california_housing
    _ = california_housing.load_data
except Exception:
    USE_SKLEARN_FALLBACK = True

FEATURE_NAMES = ["MedInc", "HouseAge", "AveRooms", "AveBedrms",
                  "Population", "AveOccup", "Latitude", "Longitude"]

if not USE_SKLEARN_FALLBACK:
    (x_train, y_train), (x_test, y_test) = california_housing.load_data()
    X = np.vstack([x_train, x_test]).astype(np.float32)
    y = np.concatenate([y_train, y_test]).astype(np.float32) # unit
    from sklearn.datasets import fetch_california_housing
    data = fetch_california_housing()
    X = data.data.astype(np.float32)
    y = data.target.astype(np.float32) # unit = 100k$

df = pd.DataFrame(X, columns=FEATURE_NAMES)
df["MedHouseVal"] = y

print(df.shape, " | y min/med/max:", float(y.min()), float(np.median(y)), float(y.max()))
df.head()
```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/california_housing_train.tfrecord 0s 0us/step

(20640, 9) | y min/med/max: 0.1499900072813034 1.7970000505447388 5.000010013580322

	MedInc	HouseAge	AveRooms	AveBedrms	Population	AveOccup	Latitude	Longitude	MedHouseVal
0	8.3252	41.0	6.984127	1.023810	322.0	2.555556	37.880001	-122.230003	8.3252
1	8.3014	21.0	6.238137	0.971880	2401.0	2.109842	37.860001	-122.220001	8.3014
2	7.2574	52.0	8.288136	1.073446	496.0	2.802260	37.849998	-122.239998	7.2574
3	5.6431	52.0	5.817352	1.073059	558.0	2.547945	37.849998	-122.250000	5.6431
4	3.8462	52.0	6.281853	1.081081	565.0	2.181467	37.849998	-122.250000	3.8462

Étapes suivantes : [New interactive sheet](#)

```
#select only AveRooms +split (baseline 1D)
X_one = df[["AveRooms"]].values.astype(np.float32)
y_all = df["MedHouseVal"].values.astype(np.float32) # 100k$
```

```
# Split 80/10/10
X_train, X_temp, y_train, y_temp = train_test_split(
    X_one, y_all, test_size=0.2, random_state=42
)
X_val, X_test, y_val, y_test = train_test_split(
    X_temp, y_temp, test_size=0.5, random_state=42
)
```

```
# Standardisation X
x_scaler = StandardScaler().fit(X_train)
X_train_s = x_scaler.transform(X_train)
X_val_s = x_scaler.transform(X_val)
X_test_s = x_scaler.transform(X_test)

# Standardisation y (stability ++)
y_scaler = StandardScaler().fit(y_train.reshape(-1,1))
y_train_s = y_scaler.transform(y_train.reshape(-1,1))
y_val_s = y_scaler.transform(y_val.reshape(-1,1))
y_test_s = y_scaler.transform(y_test.reshape(-1,1))

X_train_s[:3], y_train_s[:3].ravel()[:3]
```

```
(array([[ -0.17491654],
        [ -0.4028354 ],
        [  0.08821608]], dtype=float32),
 array([ -0.90118915,  1.5127715 , -0.2992126 ], dtype=float32))
```

```
#Keras model (1D regression) + compilation
def build_reg():
    m = keras.Sequential([
        layers.Input(shape=(1,)),
        layers.Dense(64, activation="relu"),
        layers.Dense(64, activation="relu"),
        layers.Dense(1)
    ])
    m.compile(
        optimizer=keras.optimizers.Adam(learning_rate=1e-3, clipnorm=1.0),
        loss="mse",
        metrics=[keras.metrics.MeanAbsoluteError(name="mae")]
    )
    return m
```



```
reg = build_reg()
reg.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 64)	128
dense_1 (Dense)	(None, 64)	4,160
dense_2 (Dense)	(None, 1)	65

Total params: 4,353 (17.00 KB)

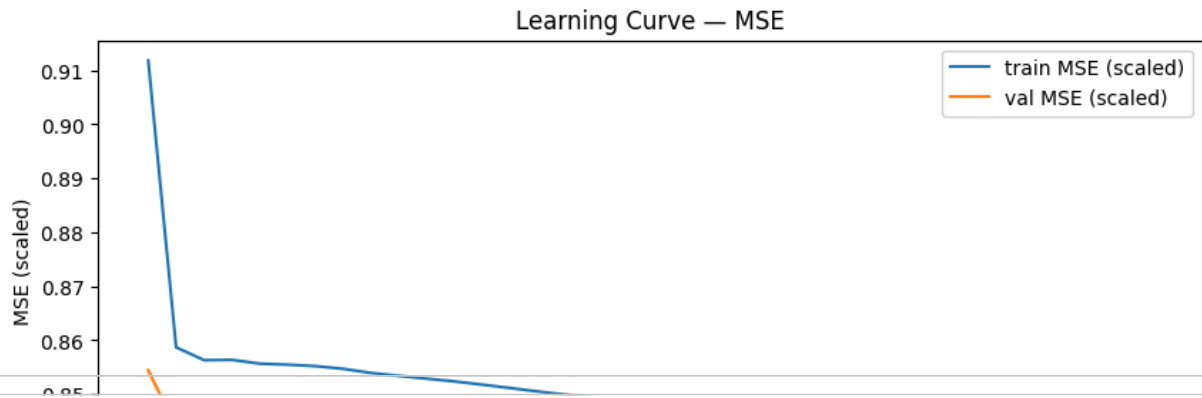
Trainable params: 4,353 (17.00 KB)

```
#train & curves
early = keras.callbacks.EarlyStopping(
    monitor="val_loss", patience=10, restore_best_weights=True
)

hist = reg.fit(
    X_train_s, y_train_s,
    validation_data=(X_val_s, y_val_s),
    epochs=200, batch_size=128, verbose=0, callbacks=[early]
)

# Courbes (loss/mae in normal standardized scale)
plt.figure(figsize=(10,4))
plt.plot(hist.history["loss"], label="train MSE (scaled)")
plt.plot(hist.history["val_loss"], label="val MSE (scaled)")
plt.xlabel("Epoch"); plt.ylabel("MSE (scaled)"); plt.legend(); plt.title("Learning Curve – MS

plt.figure(figsize=(10,4))
plt.plot(hist.history["mae"], label="train MAE (scaled)")
plt.plot(hist.history["val_mae"], label="val MAE (scaled)")
plt.xlabel("Epoch"); plt.ylabel("MAE (scaled)"); plt.legend(); plt.title("Learning Curve – MA
```



```
# Evaluation test & visualization (+ inverse transform)
y_pred_s = reg.predict(X_test_s, verbose=0)
y_pred = y_scaler.inverse_transform(y_pred_s).ravel() # 100k$

# Metrics in origin unit (100k$)
mse = np.mean((y_pred - y_test)**2)
mae = np.mean(np.abs(y_pred - y_test))
print(f"Test MSE: {mse:.4f} | Test MAE: {mae:.4f} (unit = 100k$)")

# Scatter reel vs predict (100k$)
plt.figure(figsize=(5,5))
plt.scatter(y_test, y_pred, s=10)
mn, mx = float(min(y_test.min(), y_pred.min())), float(max(y_test.max(), y_pred.max()))
plt.plot([mn, mx], [mn, mx], lw=2)
plt.xlabel("Reel price (100k$)"); plt.ylabel("Prediction (100k$)")
plt.title("Vrai vs Prédit — AveRooms → MedHouseVal")
plt.show()

print("MAE (100k$):", mean_absolute_error(y_test, y_pred))
print("R²:", r2_score(y_test, y_pred))
```

Test MSE: 1.0879 | Test MAE: 0.8180 (unit = 100k\$)

