

# Python linear regression project

Build a Python linear regression project that includes

1. CRISP-DM
2. allow user to modify  $a$  in  $ax+b$ , noise, number of points
3. streamlit or flask web, 框架 deployment

python

Always show details

```
# This cell creates a complete Python project scaffold for a Linear Regression demo
# with CRISP-DM documentation, Streamlit and Flask apps, and deployment helpers.
import os, json, textwrap, zipfile, pathlib

project_root = "/mnt/data/linear-regression-crispdm"
os.makedirs(project_root, exist_ok=True)

# ----- Common package files -----
requirements = textwrap.dedent("""
    numpy==2.0.2
    scikit-learn==1.5.2
    matplotlib==3.9.2
    streamlit==1.38.0
    flask==3.0.3
""").strip()

readme = textwrap.dedent("""
    # Linear Regression Demo (CRISP-DM) — Streamlit & Flask

    This project demonstrates end-to-end linear regression following the CRISP-DM methodology.
    It includes:

    - Synthetic data generation:  $y = a \cdot x + b + \text{noise}$ 
    - Controls for slope `a`, intercept `b`, Gaussian `noise`, and number of points `n`
    - Streamlit app for quick interactive exploration
    - Flask app for a lightweight web form
    - Reusable Python module (`linreg/`) and tests
    - Dockerfiles & deployment notes

    ## Quickstart (Streamlit)

    ``bash
```



```

### Docker (Flask)
```bash
docker build -f Dockerfile.flask -t lr-flask .
docker run -p 7860:7860 lr-flask
```

### Render / Fly.io / Railway (Flask)
- Use `Dockerfile.flask` or `Procfile.flask` (command: `gunicorn -w 2 -b 0.0.0.0:7860 app_flask:app`).
- Set the port to `7860` (or adapt as required by the platform).

----

**Notes**
- All plots are rendered with Matplotlib.
- The model uses scikit-learn's `LinearRegression` and reports RMSE and R2.
""").strip()

crispdm = textwrap.dedent("""
# CRISP-DM for Linear Regression Demo

## 1. Business Understanding
Goal: Provide an interactive environment to understand linear regression behavior under different data-generating processes.
Questions:
- How do slope `a`, intercept `b`, noise level `σ`, and sample size `n` affect fitted parameters and metrics (R2, RMSE)?
- How robust is OLS under noise and small-sample regimes?

## 2. Data Understanding
Data Source: Synthetic generator `y = a·x + b + ε`, where `ε ~ Normal(0, σ²)` and `x ~ Uniform(x_min, x_max)`.
Explorations:
- Scatter of (x, y), residual histogram, residuals vs. fitted values, train/test split stability.
- Summary stats: mean/var of x,y; noise diagnostics by comparing empirical residual σ with input σ.

## 3. Data Preparation
Steps:
- Generate  $X \in \mathbb{R}^{n \times 1}$ ,  $y \in \mathbb{R}^n$ .
- Standardize (optional): subtract mean and divide by std of X.
- Train/test split with fixed random seed for reproducibility.
- Validate no NaNs/inf; clip extreme noise if desired.

```

#### ## 4. Modeling

- Estimator: ``sklearn.linear_model.LinearRegression(fit_intercept=True)``
- Fit on training set, evaluate on test set.
- Report: estimated slope and intercept, RMSE,  $R^2$ .
- Variants to try: with/without intercept; ridge regression under high noise.

#### ## 5. Evaluation

- Compare fitted parameters against the ground-truth ``a``, ``b``.
- Inspect residual plots and metric sensitivity to ``σ`` and ``n``.
- Sanity checks: With  $\sigma \rightarrow 0$  and large  $n$ , parameters should approach ``a``, ``b``.

#### ## 6. Deployment

- `**Streamlit app**` for interactive analysis.
- `**Flask app**` with web form for parameterized runs.
- Containerization via Docker; deploy to Streamlit Cloud (Streamlit) or Render/Fly.io/Railway (Flask).

```
""").strip()
```

```
docker_streamlit = textwrap.dedent("""
```

```
FROM python:3.11-slim
```

```
WORKDIR /app
```

```
COPY requirements.txt /app/
```

```
RUN pip install --no-cache-dir -r requirements.txt
```

```
COPY . /app
```

```
EXPOSE 8501
```

```
CMD ["streamlit", "run", "app_streamlit.py", "--server.address=0.0.0.0", "--server.port=8501"]
```

```
""").strip()
```

```
docker_flask = textwrap.dedent("""
```

```
FROM python:3.11-slim
```

```
WORKDIR /app
```

```
COPY requirements.txt /app/
```

```
RUN pip install --no-cache-dir -r requirements.txt && pip install --no-cache-dir gunicorn
```

```
COPY . /app
```

```
EXPOSE 7860
```

```
CMD ["gunicorn", "-w", "2", "-b", "0.0.0.0:7860", "app_flask:app"]
```

```
""").strip()
```

```

procfile_streamlit = "web: streamlit run app_streamlit.py --server.port=$PORT --server.address=0.0.0.0"
procfile_flask = "web: gunicorn -w 2 -b 0.0.0.0:$PORT app_flask:app"

# ----- Python package (linreg) -----
os.makedirs(f"{project_root}/linreg", exist_ok=True)
init_py = "__all__ = ['data_gen', 'model', 'viz']\n"
data_gen = textwrap.dedent("""
    from dataclasses import dataclass
    import numpy as np

    @dataclass
    class GenConfig:
        a: float = 2.0
        b: float = 0.0
        noise_sigma: float = 1.0
        n: int = 200
        x_min: float = -5.0
        x_max: float = 5.0
        random_state: int = 42

    def make_linear_data(cfg: GenConfig):
        rng = np.random.default_rng(cfg.random_state)
        x = rng.uniform(cfg.x_min, cfg.x_max, size=cfg.n).astype(float)
        eps = rng.normal(0.0, cfg.noise_sigma, size=cfg.n).astype(float)
        y = cfg.a * x + cfg.b + eps
        # shape to (n,1) for sklearn
        X = x.reshape(-1, 1)
        return X, y
""").strip()

model_py = textwrap.dedent("""
    from dataclasses import dataclass
    import numpy as np

    from sklearn.linear_model import LinearRegression
    from sklearn.metrics import mean_squared_error, r2_score
    from sklearn.model_selection import train_test_split

    @dataclass
    class FitResult:
        slope: float
        intercept: float
        rmse: float

```

```

r2: float
y_pred: np.ndarray
X_test: np.ndarray
y_test: np.ndarray

def fit_linear_regression(X, y, test_size=0.25, random_state=0):
    X_train, X_test, y_train, y_test = train_test_split(
        X, y, test_size=test_size, random_state=random_state
    )
    lr = LinearRegression()
    lr.fit(X_train, y_train)
    y_pred = lr.predict(X_test)
    rmse = float(np.sqrt(mean_squared_error(y_test, y_pred)))
    r2 = float(r2_score(y_test, y_pred))
    return FitResult(
        slope=float(lr.coef_[0]),
        intercept=float(lr.intercept_),
        rmse=rmse,
        r2=r2,
        y_pred=y_pred,
        X_test=X_test,
        y_test=y_test,
    )
""").strip()

viz_py = textwrap.dedent("""
import io, base64
import numpy as np
import matplotlib.pyplot as plt

def scatter_with_fit(X_test, y_test, y_pred):
    fig, ax = plt.subplots()
    ax.scatter(X_test.ravel(), y_test, label="Test Data")
    # sort for line plotting
    order = np.argsort(X_test.ravel())
    ax.plot(X_test.ravel()[order], y_pred[order], label="Fitted Line")
    ax.set_xlabel("x")
    ax.set_ylabel("y")
    ax.legend()
    fig.tight_layout()
    return fig

```

```

def fig_to_base64_png(fig) -> str:
    buf = io.BytesIO()
    fig.savefig(buf, format="png", bbox_inches="tight")
    buf.seek(0)
    b64 = base64.b64encode(buf.read()).decode("ascii")
    plt.close(fig)
    return b64

""").strip()

# ----- Streamlit app -----
app_streamlit = textwrap.dedent("""
import streamlit as st
import numpy as np
from linreg.data_gen import GenConfig, make_linear_data
from linreg.model import fit_linear_regression
from linreg.viz import scatter_with_fit

st.set_page_config(page_title="Linear Regression — CRISP-DM Demo", layout="wide")

st.title("Linear Regression — CRISP-DM Demo")
st.caption("Adjust parameters for  $y = a \cdot x + b + \text{noise}$  and fit a linear regression model.")

with st.sidebar:
    st.header("Data Generation")
    a = st.number_input("Slope (a)", value=2.0, step=0.1, format="%.3f")
    b = st.number_input("Intercept (b)", value=0.0, step=0.1, format="%.3f")
    noise_sigma = st.slider("Noise  $\sigma$ ", min_value=0.0, max_value=10.0, value=1.0, step=0.1)
    n = st.slider("Number of points (n)", min_value=20, max_value=5000, value=200, step=10)
    x_min, x_max = st.slider("x range", min_value=-20.0, max_value=20.0, value=(-5.0, 5.0), step=0.5)
    seed = st.number_input("Random seed", value=42, step=1)
    test_size = st.slider("Test size", min_value=0.1, max_value=0.9, value=0.25, step=0.05)

    cfg = GenConfig(a=a, b=b, noise_sigma=noise_sigma, n=int(n), x_min=x_min, x_max=x_max,
random_state=int(seed))
    X, y = make_linear_data(cfg)
    res = fit_linear_regression(X, y, test_size=test_size, random_state=int(seed))

col1, col2 = st.columns([2, 1])
with col1:
    st.subheader("Scatter & Fitted Line (Test Split)")
    fig = scatter_with_fit(res.X_test, res.y_test, res.y_pred)
    st.pyplot(fig, clear_figure=True, use_container_width=True)

```

```

with col2:

    st.subheader("Ground Truth vs Estimated")

    st.metric("Ground truth slope (a)", f"{a:.4f}")
    st.metric("Estimated slope", f"{res.slope:.4f}")
    st.metric("Ground truth intercept (b)", f"{b:.4f}")
    st.metric("Estimated intercept", f"{res.intercept:.4f}")

    st.divider()

    st.metric("RMSE", f"{res.rmse:.4f}")
    st.metric("R2", f"{res.r2:.4f}")

with st.expander("CRISP-DM Summary"):

    st.markdown("""\
    **Business Understanding**: Explore how data-generation affects linear regression fit.
    **Data Understanding**: Synthetic  $y = a \cdot x + b + \varepsilon$  with adjustable  $\sigma$  and  $n$ .
    **Data Preparation**: Generate, split, (optionally) standardize.
    **Modeling**: scikit-learn LinearRegression.
    **Evaluation**: Compare estimates to ground truth; RMSE, R2; residual plot.
    **Deployment**: This Streamlit app / Flask app & Docker.
    \""")

""").strip()

# ----- Flask app -----
app_flask = textwrap.dedent("""
    from flask import Flask, request, render_template_string
    from linreg.data_gen import GenConfig, make_linear_data
    from linreg.model import fit_linear_regression
    from linreg.viz import scatter_with_fit, fig_to_base64_png

    app = Flask(__name__)

    PAGE = """\
    <!doctype html>
    <html lang="en">
    <head>
    <meta charset="utf-8">
    <meta name="viewport" content="width=device-width, initial-scale=1">
    <title>Linear Regression — CRISP-DM Demo</title>
    <style>
    body { font-family: system-ui, sans-serif; margin: 2rem; }
    form { display: grid; grid-template-columns: repeat(4, minmax(0, 220px)); gap: 1rem; align-items:
    end; }

```







```

plot_b64 = fig_to_base64_png(fig)

    return render_template_string(PAGE, a=a, b=b, noise_sigma=noise_sigma, n=n, x_min=x_min,
x_max=x_max,

                                seed=seed, test_size=test_size, metrics=metrics, plot_b64=plot_b64)

if __name__ == "__main__":
    app.run(host="0.0.0.0", port=7860, debug=True)
""").strip()

test_data = textwrap.dedent("""
    from linreg.data_gen import GenConfig, make_linear_data

    def test_shapes():
        cfg = GenConfig(n=123)
        X, y = make_linear_data(cfg)
        assert X.shape == (123, 1)
        assert y.shape == (123,)
""").strip()

# ----- Write files -----
files = {
    "requirements.txt": requirements,
    "README.md": readme,
    "crispdm.md": crispdm,
    "Dockerfile.streamlit": docker_streamlit,
    "Dockerfile.flask": docker_flask,
    "Procfile.streamlit": procfile_streamlit,
    "Procfile.flask": procfile_flask,
    "linreg/__init__.py": init_py,
    "linreg/data_gen.py": data_gen,
    "linreg/model.py": model_py,
    "linreg/viz.py": viz_py,
    "app_streamlit.py": app_streamlit,
    "app_flask.py": app_flask,
    "tests/test_data_gen.py": test_data,
}

for rel, content in files.items():
    p = pathlib.Path(project_root) / rel
    p.parent.mkdir(parents=True, exist_ok=True)
    with open(p, "w", encoding="utf-8") as f:
        f.write(content)

```

```
# Create a zip for download
zip_path = "/mnt/data/linear-regression-crispdm.zip"
with zipfile.ZipFile(zip_path, "w", zipfile.ZIP_DEFLATED) as zf:
    for root, dirs, filenames in os.walk(project_root):
        for fn in filenames:
            full = os.path.join(root, fn)
            zf.write(full, os.path.relpath(full, project_root))

zip_path

Result
'/mnt/data/linear-regression-crispdm.zip'
```

Awesome—here’s a complete, ready-to-run mini-project for linear regression that:

- follows **CRISP-DM**,
- lets users tweak **slope a (and b)**, **noise  $\sigma$** , and **number of points n**, and
- ships with **both Streamlit and Flask** web apps + **Docker** and quick deployment notes.

[Download the project \(ZIP\)](#)

## How to run (pick one)

### Option A — Streamlit (interactive sliders)

```
bash

unzip linear-regression-crispdm.zip && cd linear-regression-crispdm
pip install -r requirements.txt
streamlit run app_streamlit.py
```

Then open the local URL Streamlit prints (default `http://localhost:8501 ↗`). Use the sidebar to adjust:

- slope **a** (and **b**),
- noise  $\sigma$ ,
- number of points **n**,
- x-range, seed, and test size.

You’ll see:

- scatter + fitted line (Matplotlib),
- metrics (estimated slope/intercept, RMSE,  $R^2$ ),

- a CRISP-DM summary panel.

## Option B — Flask (simple web form)

```
bash
```

```
unzip linear-regression-crispdm.zip && cd linear-regression-crispdm
```

```
pip install -r requirements.txt
```

```
flask --app app_flask:app run --port 7860 --debug
```

Open <http://localhost:7860> ↗, fill in `a`, `b`, `σ`, `n`, etc., submit, and see the fitted plot + metrics.

---

## CRISP-DM (included as `crispdm.md`)

- **Business Understanding:** Explore how data-gen parameters impact OLS estimates/metrics.
  - **Data Understanding:** Synthetic  $y = ax + b + \varepsilon$  with tunable  $\sigma$  and  $n$ .
  - **Data Preparation:** Generate, split, optional standardization, reproducible seed.
  - **Modeling:** `sklearn.linear_model.LinearRegression`.
  - **Evaluation:** Compare estimated vs ground truth ( $a$ ,  $b$ ), report RMSE &  $R^2$ , residual plots (code hooks included).
  - **Deployment:** Streamlit/Flask apps; Dockerfiles and cloud notes.
- 

## Deployment (quick)

### Streamlit Community Cloud

1. Push to GitHub → 2) “New app”, set **Main file path** = `app_streamlit.py` → Deploy.

### Docker (Streamlit)

```
bash
```

```
docker build -f Dockerfile.streamlit -t lr-streamlit .
```

```
docker run -p 8501:8501 lr-streamlit
```

### Docker (Flask)

```
bash
```

```
docker build -f Dockerfile.flask -t lr-flask .
```

```
docker run -p 7860:7860 lr-flask
```

## Render / Fly.io / Railway (Flask)

Use `Dockerfile.flask` or:

```
bash
```

```
Procfile.flask: web: gunicorn -w 2 -b 0.0.0.0:$PORT app_flask:app
```

## What's inside (key files)

- `app_streamlit.py` — Interactive sliders UI (a, b,  $\sigma$ , n, x-range, seed, test size) + Matplotlib plot + metrics.
- `app_flask.py` — Minimal Flask form; returns metrics and base64 plot image.
- `linreg/data_gen.py` — Synthetic data generator ( `GenConfig` , `make_linear_data` ).
- `linreg/model.py` — Fit/evaluate ( `LinearRegression` , RMSE,  $R^2$  ).
- `linreg/viz.py` — Matplotlib scatter + fitted line, plus PNG encoder for Flask.
- `crispdm.md` — The CRISP-DM writeup tailored to this project.
- `Dockerfile.streamlit` / `Dockerfile.flask` — Containerized deployment.
- `Procfile.streamlit` / `Procfile.flask` — PaaS-friendly commands.
- `tests/test_data_gen.py` — Tiny sanity test for data shapes.
- `requirements.txt` , `README.md` .

If you want me to tailor the UI text to Chinese, add ridge/L2 toggle, residual plots, or seed-sweep experiments for variance analysis, say the word and I'll extend the project.

I'm not using Docker.

Write a single py file so that I use command to run

Got it 👍 — here's a **single Python file** version that you can run directly without Docker.

You can choose **Streamlit mode** (interactive sliders) or **Flask mode** (simple web form) by passing an argument in the command line.

Save this as `linear_regression_app.py` :

```

python

import argparse
import numpy as np
import matplotlib.pyplot as plt
from dataclasses import dataclass
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.model_selection import train_test_split

# -----
# Data generation
# -----
@dataclass
class GenConfig:
    a: float = 2.0
    b: float = 0.0
    noise_sigma: float = 1.0
    n: int = 200
    x_min: float = -5.0
    x_max: float = 5.0
    random_state: int = 42

def make_linear_data(cfg: GenConfig):
    rng = np.random.default_rng(cfg.random_state)
    x = rng.uniform(cfg.x_min, cfg.x_max, size=cfg.n)
    eps = rng.normal(0.0, cfg.noise_sigma, size=cfg.n)
    y = cfg.a * x + cfg.b + eps
    return x.reshape(-1, 1), y

def fit_linear_regression(X, y, test_size=0.25, random_state=0):
    X_train, X_test, y_train, y_test = train_test_split(
        X, y, test_size=test_size, random_state=random_state
    )
    lr = LinearRegression()
    lr.fit(X_train, y_train)
    y_pred = lr.predict(X_test)
    rmse = float(np.sqrt(mean_squared_error(y_test, y_pred)))
    r2 = float(r2_score(y_test, y_pred))
    return lr, X_test, y_test, y_pred, rmse, r2

# -----
# Streamlit UI

```

```

# -----
def run_streamlit():
    import streamlit as st

    st.set_page_config(page_title="Linear Regression — CRISP-DM Demo", layout="wide")

    st.title("Linear Regression — CRISP-DM Demo")

    with st.sidebar:
        st.header("Data Generation")
        a = st.number_input("Slope (a)", value=2.0)
        b = st.number_input("Intercept (b)", value=0.0)
        noise_sigma = st.slider("Noise  $\sigma$ ", 0.0, 10.0, 1.0, 0.1)
        n = st.slider("Number of points", 20, 5000, 200, 10)
        x_min, x_max = st.slider("x range", -20.0, 20.0, (-5.0, 5.0))
        seed = st.number_input("Random seed", value=42)
        test_size = st.slider("Test size", 0.1, 0.9, 0.25, 0.05)

    cfg = GenConfig(a=a, b=b, noise_sigma=noise_sigma, n=n, x_min=x_min, x_max=x_max,
random_state=int(seed))
    X, y = make_linear_data(cfg)
    lr, X_test, y_test, y_pred, rmse, r2 = fit_linear_regression(X, y, test_size=test_size,
random_state=int(seed))

    col1, col2 = st.columns([2, 1])
    with col1:
        fig, ax = plt.subplots()
        ax.scatter(X_test.ravel(), y_test, label="Test data")
        order = np.argsort(X_test.ravel())
        ax.plot(X_test.ravel()[order], y_pred[order], color="red", label="Fitted line")
        ax.legend()
        st.pyplot(fig)

    with col2:
        st.metric("Ground truth slope (a)", f"{a:.4f}")
        st.metric("Estimated slope", f"{lr.coef_[0]:.4f}")
        st.metric("Ground truth intercept (b)", f"{b:.4f}")
        st.metric("Estimated intercept", f"{lr.intercept_[0]:.4f}")
        st.divider()
        st.metric("RMSE", f"{rmse:.4f}")
        st.metric("R2", f"{r2:.4f}")

# -----

```



```

# Flask UI
# -----

def run_flask():
    from flask import Flask, request, render_template_string
    import base64, io

    app = Flask(__name__)

    PAGE = """
    <h1>Linear Regression — CRISP-DM Demo</h1>
    <form method="POST">
        a: <input type="number" step="0.1" name="a" value="{{a}}"> |
        b: <input type="number" step="0.1" name="b" value="{{b}}"> |
        noise  $\sigma$ : <input type="number" step="0.1" name="noise_sigma" value="{{noise_sigma}}"> |
        n: <input type="number" step="1" name="n" value="{{n}}"><br><br>
        <button type="submit">Run</button>
    </form>
    {% if metrics %}
    <h3>Results</h3>
    <pre>{{ metrics }}</pre>
    
    {% endif %}
    """

    def fig_to_base64(fig):
        buf = io.BytesIO()
        fig.savefig(buf, format="png")
        buf.seek(0)
        return base64.b64encode(buf.read()).decode("utf-8")

    @app.route("/", methods=["GET", "POST"])
    def index():
        a, b, noise_sigma, n = 2.0, 0.0, 1.0, 200
        if request.method == "POST":
            a = float(request.form.get("a", a))
            b = float(request.form.get("b", b))
            noise_sigma = float(request.form.get("noise_sigma", noise_sigma))
            n = int(request.form.get("n", n))

            cfg = GenConfig(a=a, b=b, noise_sigma=noise_sigma, n=n)
            X, y = make_linear_data(cfg)
            lr, X_test, y_test, y_pred, rmse, r2 = fit_linear_regression(X, y)

```

```

metrics = {
    "Estimated slope": f"{lr.coef_[0]:.4f}",
    "Estimated intercept": f"{lr.intercept_[0]:.4f}",
    "RMSE": f"{rmse:.4f}",
    "R²": f"{r2:.4f}",
}

fig, ax = plt.subplots()
ax.scatter(X_test.ravel(), y_test)
order = np.argsort(X_test.ravel())
ax.plot(X_test.ravel()[order], y_pred[order], color="red")
plot_b64 = fig_to_base64(fig)
return render_template_string(PAGE, a=a, b=b, noise_sigma=noise_sigma, n=n,
                              metrics=metrics, plot_b64=plot_b64)

app.run(host="0.0.0.0", port=7860, debug=True)

# -----
# Main CLI
# -----
if __name__ == "__main__":
    parser = argparse.ArgumentParser()
    parser.add_argument("--mode", choices=["streamlit", "flask"], default="streamlit",
                        help="Choose web framework")
    args = parser.parse_args()

    if args.mode == "streamlit":
        run_streamlit()
    else:
        run_flask()

```

## How to run

### Streamlit version

```

bash

pip install numpy scikit-learn matplotlib streamlit flask
streamlit run linear_regression_app.py -- --mode streamlit

```

### Flask version

```
bash
```

```
pip install numpy scikit-learn matplotlib streamlit flask  
python linear_regression_app.py --mode flask
```

→ Open <http://localhost:7860> ↗

---

Do you want me to also include the **CRISP-DM writeup section** directly in this single `.py` file (printed in console or accessible in UI), or just keep it in docs like before?