Hugging Face Hub

Module 4 of 7

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Introduction

What is Hugging Face Hub?

Hugging Face Hub = Centralized platform for ML datasets and models

Purpose

- Version control for large files Git LFS backend for datasets/models
- Optimized for ML workloads Efficient storage and transfer
- Community sharing and collaboration Open science principles
- Professional data management Reproducible research workflows

Why Essential for Research?

- Proper dataset versioning Track data changes systematically
- Model artifact storage Save and share trained models
- Collaboration enablement Team access to shared resources
- Reproducibility support Fixed dataset/model versions for papers

Comparison with Traditional Storage

BAI La	Storage Method	File Size Limit	Version Control	ML Optimization	Collaboration	Cost
	GitHub	100MB per file	Yes	No	Limited	Free/Paid
	Google Drive	5TB	No	No	Basic	Free/Paid
	Dropbox	No limit	No	No	Basic	Paid
	AWS S3	5TB per object	No	No	Programmatic	Paid
	HF Hub b · Hugging Face Hub ·	Very Large	Yes	Yes	Excellent	Free/Paid

Key Advantages

- Git LFS integration Seamless version control for large files
- Dataset viewer Browse data without downloading
- Model cards Standardized documentation
- **Direct integration** Works with transformers, datasets libraries
- Free hosting Generous limits for research use

Purpose-built for ML research workflows - Unlike general storage solutions

Essential Operations (80/20 Principle)

Authentication and Setup

```
# Install Hugging Face CLI
uv pip install huggingface_hub

# Login with your token (one-time setup)
huggingface-cli login
# Enter your token from: https://huggingface.co/settings/tokens
```

Core Upload/Download Operations

```
# Most common operations for research
     from huggingface_hub import login, upload_folder, download_folder, upload_file
     # Upload dataset folder
     upload folder(
         folder_path="dataset/processed/my_research_data",
         repo_id="bailab/research-dataset-v1",
         repo type="dataset"
     # Upload trained model
     upload_folder(
         folder_path="models/checkpoints/best_model",
         repo_id="bailab/transformer-model-v1",
         repo type="model"
     # Download for use in training
     download folder(
         repo id="bailab/research-dataset-v1",
         local dir="./data"
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```

Integration with Popular Libraries

```
# Direct integration with datasets library
from datasets import load_dataset, Dataset

# Load from HF Hub
dataset = load_dataset("bailab/research-dataset-v1")

# Save to HF Hub
dataset.push_to_hub("bailab/processed-dataset-v2")
```

Lab Demo with base-research-repo

Step 1: Setup HF Hub Authentication

```
cd base-research-repo

# Login to Hugging Face
huggingface-cli login
# Paste your token from: https://huggingface.co/settings/tokens

# Verify authentication
huggingface-cli whoami
```

Step 2: Upload Dataset

```
# Create example dataset structure
mkdir -p dataset/processed/bailab_demo_data
echo '{"text": "Sample research data", "label": 1}' > dataset/processed/bailab_demo_data/train.jsonl
echo '{"text": "Test research data", "label": 0}' > dataset/processed/bailab_demo_data/test.jsonl

# Upload to HF Hub
from huggingface_hub import upload_folder

upload_folder(
    folder_path="dataset/processed/bailab_demo_data",
    repo_id="bailab/demo-research-dataset",
    repo_type="dataset",
    commit_message="Initial upload of demo research dataset"
)
```

Step 3: Download and Use Dataset

```
# In your training script
from datasets import load_dataset
# Load dataset from HF Hub
dataset = load_dataset("bailab/demo-research-dataset")
# Access train/test splits
train data = dataset['train']
test_data = dataset['test']
print(f"Training samples: {len(train_data)}")
print(f"Test samples: {len(test data)}")
```

Step 4: Upload Trained Model

```
# After training, upload model artifacts
from transformers import AutoTokenizer, AutoModel
# Save model locally first
model.save_pretrained("./trained_model")
tokenizer.save_pretrained("./trained_model")
# Upload to HF Hub
upload_folder(
    folder_path="./trained_model",
    repo_id="bailab/demo-trained-model",
    repo_type="model"
```

Important Considerations

File Size and Storage Limits

GitHub vs HF Hub Guidelines

- Never commit large files to GitHub (>25MB)
- Use HF Hub for datasets (>25MB)
- Use HF Hub for model checkpoints (always large)
- Keep code in GitHub HF Hub for data/models only

Storage Quotas and Limits

```
# Check your storage usage
from huggingface_hub import get_repo_discussions, repo_info

info = repo_info("bailab/your-dataset-repo", repo_type="dataset")
print(f"Repository size: {info.cardData.size if info.cardData else 'Unknown'}")
```

Private vs Public Repositories

- Public repositories Open science, community access
- Private repositories Sensitive data, work-in-progress
- Organization repositories Team collaboration

Token Management

```
# Create tokens with minimal required permissions
# Read: Download models/datasets
# Write: Upload new versions
# Delete: Remove repositories (use carefully)

# Store tokens securely
export HUGGINGFACE_TOKEN="your_token_here"
```

Data Management Best Practices

Version Control Strategy

- **Semantic versioning** for datasets (v1.0, v1.1, etc.)
- **Descriptive commit messages** for changes
- Model cards and dataset cards for documentation
- Tag releases for paper submissions

Repository Organization

Summary: Hugging Face Hub

What We Covered

- ✓ HF Hub fundamentals ML-optimized storage platform
- Advantages over traditional storage Version control + ML features
- **Essential operations** Upload, download, and integration
- Lab-specific workflow Dataset and model management
- Best practices Security, versioning, and organization

Key Takeaways

- 1. **HF Hub is purpose-built for ML research** Not just file storage
- 2. Seamless version control for large datasets and models
- 3. Direct library integration Works with transformers, datasets
- 4. Proper separation of concerns Code in GitHub, data in HF Hub
- 5. **Professional data management** enables reproducible research

Impact on Research Workflow

- Eliminates data storage headaches No more email attachments
- Enables true reproducibility Fixed dataset/model versions
- Facilitates collaboration Team access to shared resources
- Supports open science Easy sharing with research community

Next Steps

- Module 5: Weights & Biases Tracking Experiment management
- Create your first dataset repository on HF Hub
- Upload a model checkpoint from current project