JAX Serving with vLLM & SGLang

Quick Reference

This cheat sheet provides a quick reference for the functions and utilities used to convert a JAX model and serve it with the popular open-source servers vLLM and SGLang.

End-to-End Workflow

The process involves preparing the JAX model weights and then loading them into a serving engine.

Step	Action	Description
Load	Load model weights into JAX/Flax.	Use the load_safetensors utility to read . safetensors files from a directory into a nested dictionary.
Transpose	Transpose weights for specific layers.	Crucial Step: Before flattening, transpose weights for layers like Dense/Linear to match the PyTorch format.
Convert	Flatten the weight dictionary.	Use the flatten_weight_dict utility to convert the (now transposed) nested JAX weights into a flat key-value structure.
Save	Save the flattened weights.	Use safetensors.flax.save_file to write the flattened dictionary to a single model.safetensors file.
Serve	Load the model into a server.	Point either vLLM or SGLang to the directory containing the new model.safetensors file to start the inference engine.

Key Functions & Code Examples

1. Loading Safetensors into JAX/Flax

This function aggregates all .safetensors files in a directory into a single JAX/Flax compatible weight dictionary.

```
from pathlib import Path
from safetensors import safe_open

def load_safetensors(path_to_model_weights):
    weights = {}
    safetensors_files =

Path(path_to_model_weights).glob('*.safetensors')
    for file in safetensors_files:
        with safe_open(file, framework="flax") as f:
        for key in f.keys():
            weights[key] = f.get_tensor(key)
    return weights
```

2. Handling Weight Transposition (Important!)

Before flattening, certain JAX layer weights **must be transposed** to match the dimensional format expected by PyTorch-based servers. Failure to do so will result in errors or incorrect model outputs.

- Dense/Linear Layers: The kernel weights must be transposed. JAX typically uses (input_features, output_features), while PyTorch expects (output_features, input_features).
- Attention Projections (e.g., q_proj, k_proj, v_proj, o_proj): These are linear layers, and their kernel weights must also be transposed.
- **Embedding Layers:** These weights are generally compatible and do **not** need to be transposed.

You must implement logic to iterate through your loaded <code>jax_weights</code> dictionary and apply .T to the appropriate tensors before flattening. The exact implementation will depend on your model's specific structure.

3. Flattening & Saving Weights for Serving

This function converts the nested (and now correctly transposed) JAX weight dictionary into a flat format and saves it.

```
Python
from safetensors.flax import save_file
# Utility to flatten the nested dictionary
def flatten_weight_dict(torch_params, prefix=""):
    flat_params = {}
    for key, value in torch_params.items():
        new_key = f"{prefix}{key}" if prefix else key
        if isinstance(value, dict):
            flat_params.update(flatten_weight_dict(value, new_key)
+ "."))
        else:
            flat_params[new_key] = value
    return flat_params
# --- Usage ---
# jax_weights = load_safetensors(...)
# # ... (Apply transpositions to jax_weights here) ...
# servable_weights = flatten_weight_dict(jax_weights)
# save_file(servable_weights, 'path/to/model.safetensors')
```

4. Serving with vLLM

Initialize the LLM class pointing to your model directory and use .generate().

```
from vllm import LLM, SamplingParams

# Point to the directory with the converted model.safetensors
llm = LLM(model="/path/to/your/model", load_format="safetensors",
dtype="half")
```

```
# Define prompts and sampling parameters
prompts = ["The capital of France is"]
sampling_params = SamplingParams(temperature=0.8, top_p=0.95)
# Generate text
outputs = llm.generate(prompts, sampling_params)
```

5. Serving with SGLang

Initialize the sgl.Engine and use its .generate() method. Note: Requires CUDA 12.4.

```
import sglang as sgl

# Point to the directory with the converted model.safetensors
llm = sgl.Engine(model_path="/path/to/your/model")

# Define prompts and sampling parameters
prompts = ["The capital of France is"]
sampling_params = {"temperature": 0.8, "top_p": 0.95}

# Generate text
outputs = llm.generate(prompts, sampling_params)
```