# JAX & Flax NNX Debugging Quick Reference

### Core Challenge: The JIT Impact

- JAX relies on Compiles Python functions (tracing phase) into optimized code (execution phase).
- **Problem:** Standard Python print(), pdb, breakpoint() only see abstract *tracers* during tracing, not runtime *values* inside JIT-compiled code.
- Goal: Use JAX-specific tools or temporarily disable JIT to inspect runtime behavior.

# Runtime Inspection (Inside JIT / Transformed Functions)

```
jax.debug.print("msg: {var}", var=value, ordered=True)
```

- **Purpose:** The JAX equivalent of print() for runtime values.
- **How:** Embeds print into the compiled graph.
- Notes: Use ordered=True for sequential output. Shows concrete values during execution.

#### jax.debug.breakpoint()

- Purpose: The JAX equivalent of pdb.set\_trace() / breakpoint() for interactive debugging inside JIT.
- **How:** Pauses runtime execution, provides a (jaxdb) prompt.
- Commands: p <var> (print variable), c (continue), q (quit). Limited command set vs pdb.
- Notes: Can be made conditional using jax.lax.cond(). Standard Python breakpoint() can inspect *tracers* during tracing.

### Enabling Standard Python Tools (Eager Execution)

```
jax.disable jit()
```

- **Purpose:** Temporarily forces eager execution (like PyTorch/NumPy), making standard Python tools work.
- How:
  - Context Manager: with jax.disable jit(): ... (Recommended)

- Global: jax.config.update("jax disable jit", True)
- Env Var: JAX DISABLE JIT=1
- Effect: Standard print(), pdb.set\_trace(), breakpoint(), IDE debuggers work directly on runtime values.
- Drawback: Significant performance loss. Use temporarily for debugging.
- Limits: May not fully step into jax.vmap / jax.scan internals.

### **Automatic Error Detection**

#### jax\_debug\_nans Flag

- Purpose: Automatically find the source operation causing NaNs inside JIT.
- **How:** jax.config.update("jax\_debug\_nans", True) **or** JAX\_DEBUG\_NANS=1.
- **Effect:** If NaN detected, JAX re-runs eagerly to pinpoint the error source.
- Limitations: Slows execution; doesn't work with jax.shard\_map; may flag intentional NaNs.

### Flax NNX Specific Tools

#### nnx.display(object)

- **Purpose:** Visualize the structure and state of NNX objects (Modules, Optimizers, State).
- How: Provides a hierarchical view, uses treescope for interactive display in notebooks/Colab if available.
- **Use:** Verify model architecture, check parameters/state initialization, inspect optimizer state. Analogous to print (pytorch model).

```
Module.sow(variable type, name, value, ...)
```

- **Purpose:** Capture intermediate values (e.g., activations) during nnx.Module.\_\_call\_\_ without altering function signatures.
- How: Stores value as an attribute (name) on the module instance. Access via module.<name>.value.
- **Notes:** Default appends to a tuple if called multiple times with the same name. Use variable type (e.g., nnx.Intermediate) for organization.

# Chex for Robustness & Debugging

**Purpose:** Library for reliable JAX code (assertions, testing).

Static Assertions: Check shapes, dtypes, ranks, structure (properties known at trace time).

- Examples: chex.assert\_shape, chex.assert\_type, chex.assert\_rank, chex.assert\_trees\_all\_equal\_shapes.
- Usage: Place directly inside JIT-compiled functions.

Value Assertions: Check properties based on runtime values (NaN/Inf, closeness).

- Examples: chex.assert tree all finite, chex.assert trees all close.
- Usage (in JIT): Requires @chex.chexify() decorator.

#### @chex.chexify

- Purpose: Decorator to enable Chex value assertions inside JIT.
- How: Apply outside @jax.jit. Function returns (error, result). Check error.throw()
  or error.get().
- Notes: Performance overhead; debugging tool; Doesn't work in Colab currently. Usually need chex.block until chexify assertions complete() after call.

```
@chex.assert max traces(n=...)
```

- Purpose: Debug performance by detecting unexpected JIT re-compilations.
- How: Apply inside @jax.jit. Raises AssertionError if function is traced > n times.
- **Use:** Pinpoint causes of slow-downs due to unstable input shapes/types. Use chex.clear\_trace\_counter() if needed.

### Monitoring & Visualization

#### **TensorBoard**

- Purpose: Visualize training metrics, images, profiling data.
- **Setup:** Create a Summary Writer (e.g., via tensorflow, torch.utils.tensorboard, or tensorboardx). Launch tensorboard --logdir <log dir>.

- Logging: Use writer.add\_scalar, writer.add\_image, etc. Remember .item() to convert JAX scalar arrays to Python scalars before logging.
- **Profiling:** JAX profiler (jax.profiler) can output data viewable in TensorBoard.

### Recommended Workflow Summary

- 1. Static Checks: nnx.display, Chex static assertions (assert shape, etc.).
- 2. Runtime JIT Issues: Chex value assertions + @chex.chexify (for NaNs, etc.), jax debug nans, jax.debug.print, jax.debug.breakpoint.
- 3. **Complex Issues:** Temporarily use with <code>jax.disable\_jit()</code>: to enable standard pdb/print.
- 4. **Performance:** chex.assert max traces, jax.profiler + TensorBoard.
- 5. **Monitor:** Use TensorBoard throughout.

# **Profiling**

Profiling is also essential for understanding and improving your code. Xprof is a great tool for profiling JAX and Flax NNX, and is compatible with TensorBoard. <u>An excellent tutorial is available which includes profiling</u>, so developers are encouraged to review that tutorial.

### More Information

- JAX AI Stack https://jaxstack.ai
- Chex https://chex.readthedocs.io
- JAX <a href="https://jax.dev">https://jax.dev</a>
- Flax https://flax.readthedocs.io