```
In [4]: # !pip install sbi
        import torch
        import numpy as np
        from scipy.signal import welch
        from sbi import utils as sbi utils
        from sbi.inference import SNPE
        import matplotlib.pyplot as plt
        import pandas as pd
        import seaborn as sns
        from tensorflow.keras.models import load_model
In [5]: # Load Trained Generator
        generator = load_model("saved_models/qpo_cgan_phy_generator.keras")
        # --- Hyperparameters ---
        latent dim = 100
        num simulations = 5000
        num_test_samples = 100
        # Prior: fc \in [0.01, 1.0], amp \in [0.1, 1.0]
        prior = sbi utils.BoxUniform(low=torch.tensor(
            [0.01, 0.1]), high=torch.tensor([1.0, 1.0]))
          GAN-based simulator for SBI
        def simulator(theta):
            fc, amp = theta.numpy()
            z = torch.randn((1, latent_dim))
            label = torch.tensor([[fc, amp, 1.0]], dtype=torch.float32) # force
            generated = generator([z, label], training=False).numpy().squeeze()
            f, Pxx = welch(generated, fs=1, nperseg=256)
            return torch.tensor(Pxx, dtype=torch.float32)
        # --- Generate Simulations
        print("Generating synthetic QPOs for SBI...")
        thetas = prior.sample((num_simulations,))
        xs = torch.stack([simulator(theta) for theta in thetas])
        # --- Train SBI Posterior
        inference = SNPE(prior)
        density_estimator = inference.append_simulations(thetas, xs).train()
        posterior = inference.build_posterior(density_estimator)
        # Save Trained Posterior
        torch.save(posterior, "saved_models/trained_sbi_posterior.pt")
        print(" SBI Posterior trained and saved.")
       Generating synthetic QPOs for SBI...
        Neural network successfully converged after 52 epochs. SBI Posterior trai
       ned and saved.
In [ ]: import numpy as np
        import torch
        from scipy.signal import welch
```

import matplotlib.pyplot as plt

```
import seaborn as sns
def run_grid_sbi_evaluation(generator, posterior, latent_dim=100,
                            fc_vals=np.linspace(0.01, 0.5, 8),
                            amp vals=np.linspace(0.1, 1.0, 8),
                            num samples=500):
    .....
    Run full-loop inference on a grid of fc/amp values and collect mean e
    Returns 2D arrays of errors.
    fc_error_map = np.zeros((len(fc_vals), len(amp_vals)))
    amp_error_map = np.zeros((len(fc_vals), len(amp_vals)))
    for i, fc in enumerate(fc_vals):
        for j, amp in enumerate(amp_vals):
            # Step 1: Generate GAN sample
            label = torch.tensor([[fc, amp, 1.0]], dtype=torch.float32)
            z = torch.randn((1, latent_dim))
            signal = generator([z, label], training=False).numpy().squeez
            # Step 2: Compute PSD
            f, Pxx = welch(signal, fs=1, nperseg=256)
            x_obs = torch.tensor(Pxx, dtype=torch.float32)
            # Step 3: Run SBI inference
            samples = posterior.sample(
                (num_samples,), x=x_obs, show_progress_bars=False)
            fc pred = samples[:, 0].mean().item()
            amp_pred = samples[:, 1].mean().item()
            # Step 4: Record absolute error
            fc_error_map[i, j] = abs(fc_pred - fc)
            amp_error_map[i, j] = abs(amp_pred - amp)
    return fc_vals, amp_vals, fc_error_map, amp_error_map
def plot_error_heatmaps(fc_vals, amp_vals, fc_errs, amp_errs):
    fig, axs = plt.subplots(1, 2, figsize=(14, 6))
    sns.heatmap(fc_errs, xticklabels=np.round(amp_vals, 2), yticklabels=n
                ax=axs[0], cmap="Blues", annot=True)
    axs[0].set_title("FC Error Heatmap")
    axs[0].set_xlabel("Amplitude")
    axs[0].set_ylabel("FC (Hz)")
    sns.heatmap(amp_errs, xticklabels=np.round(amp_vals, 2), yticklabels=
                ax=axs[1], cmap="Oranges", annot=True)
    axs[1].set_title("Amplitude Error Heatmap")
    axs[1].set_xlabel("Amplitude")
    axs[1].set_ylabel("FC (Hz)")
    plt.suptitle("SBI Inference Error Across FC-Amplitude Grid")
    plt.tight_layout()
    plt.show()
```

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In []: # Load posterior (already trained and saved)
posterior = torch.load("saved_models/trained_sbi_posterior.pt")
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generator = load_model("saved_models/qpo_cgan_phy_generator.keras")

fc_vals, amp_vals, fc_errs, amp_errs = run_grid_sbi_evaluation(
    generator, posterior)
plot_error_heatmaps(fc_vals, amp_vals, fc_errs, amp_errs)
```

