Technology C Relaxation

July 24, 2021

1 RRAM Relaxation Data Notebook

This notebook contains the analysis on empirical RRAM relaxation data across three technologies (A, B, C). It loads and processes the measurements taken for each technology.

[1]:
```
# Imports
import json
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import scipy.signal
import scipy.stats
# from matplotlib.offsetbox import AnchoredText

%config InlineBackend.figure_format = 'svg'
```

1.1 Load the technology and its settings

Below, choose which technology to load data and settings for:

[2]:
```
# Choose technology here
TECH = 'C'

# Load settings for technology
with open(f"data/tech{TECH}/settings.json") as sfile:
    settings = json.load(sfile)
```

1.2 Time series analysis

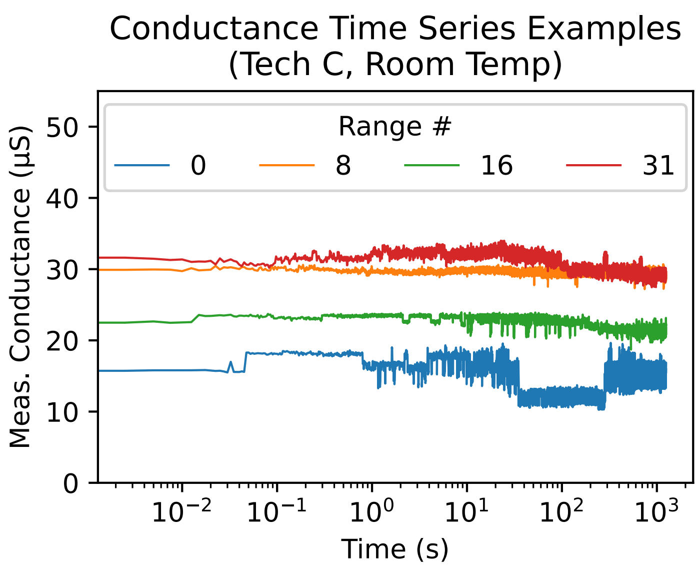
In this section, we will look at example time series data on a log scale and also examine the power spectral density (PSD). First, let us load the time series data:

[3]:
```
# Load data for technology
data = np.loadtxt(f"data/tech{TECH}/tsdata.min.tsv.gz", delimiter='\t',␣
↪usecols=range(settings["ts_npts"]))
```

1.2.1 Example Time Series Data

Below, we can look at the time series data for the ranges chosen above:

[4]:
```
# Plot example time series data
fig = plt.figure(figsize=(4,2.7))
ax = fig.add_subplot(111)
ax.set_title(f"Conductance Time Series Examples\n(Tech {TECH}, Room Temp)")
for i, r in enumerate([0, 8, 16, 31]):
    # Select data
    d = data[i,2:]

    # Plot time series data
    plt.plot(np.arange(0, len(d)/settings["fs"], 1/settings["fs"]), d*1e6,␣
↪label=r, linewidth=0.8)

# Format and display
ax.legend(title="Range #", ncol=4, loc=9)
ax.set_ylim(*settings["ts_ylim"])
ax.set_xlabel("Time (s)")
ax.set_ylabel("Meas. Conductance (µS)")
ax.set_xscale("log")
plt.savefig(f"figs/tech{TECH}/time-series.pdf", bbox_inches="tight")
plt.show()
```



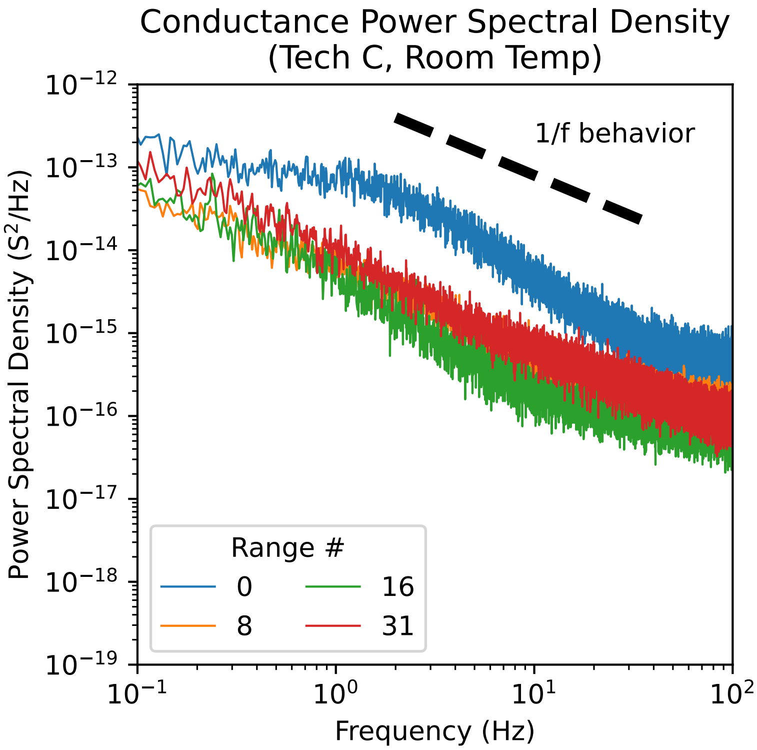
1.2.2 Power Spectral Density (PSD)

In this section, we will look at the PSDs to understand the relaxation behavior better:

[5]:
```
# Plot power spectral density (PSD)
fig = plt.figure(figsize=(4,4))
ax = fig.add_subplot(111)
```

```
ax.set_title(f"Conductance Power Spectral Density\n(Tech {TECH}, Room Temp)")
slopes = []
for i, r in enumerate([0, 8, 16, 31]):
    # Select data
    d = data[i,2:]

    # # Lomb-Scargle PSD
    # f = np.logspace(np.log10(1/600), np.log10(2), 500)
    # f = np.logspace(np.log10(2), np.log10(400/2), 1000)
    # p = scipy.signal.lombscargle(d["time"], d["g"], f)

    # Welch PSD
    f, p = scipy.signal.welch(d, fs=settings["fs"],␣
↪nperseg=settings["psd_nperseg"])
    plt.plot(f, p, label=r, linewidth=0.8)

    # # Power law fit
    # a, b = np.polyfit(np.log(f[[30,-30]]), np.log(p[[30,-30]]), 1)
    # print(f"Range {r} slope: {a}")
    # slopes.append(a)
    # plt.plot(f[1:], np.exp(a*np.log(f[1:]) + b), 'k--', zorder=10,␣
↪linewidth=2)

    # # 1/f fit
    # fitfn = lambda logf, A: A - logf
    # params, _ = scipy.optimize.curve_fit(fitfn, np.log(f), np.log(p))
    # A = params[0]
    # print(A)
    # plt.plot(f, np.exp(fitfn(np.log(f), A)), 'k--', zorder=10, linewidth=2)

# Format and display
# ax.add_artist(AnchoredText("Slopes: [%.2f, %.2f]" % (max(slopes),␣
↪min(slopes)), loc=1, frameon=False))
plt.plot(*settings["psd_fpts"], 'k--', zorder=10, linewidth=4)
if "psd_f2pts" in settings:
    plt.plot(*settings["psd_f2pts"], 'k--', zorder=10, linewidth=4)
ax.legend(title="Range #", ncol=2, loc=3)
ax.set_xlabel("Frequency (Hz)")
ax.set_ylabel("Power Spectral Density (S$^2$/Hz)")
ax.set_xscale("log")
ax.set_xlim(*settings["psd_xlim"])
ax.set_ylim(*settings["psd_ylim"])
ax.set_yscale("log")
plt.text(*settings["psd_ftextloc"], "1/f behavior")
if "psd_f2textloc" in settings:
    plt.text(*settings["psd_f2textloc"], "1/f$^2$")
plt.savefig(f"figs/tech{TECH}/psd.pdf", bbox_inches="tight")
```

```
plt.show()
```



1.3 Relaxation data analysis

Here, we will be analyzing the large dataset relaxation behavior. We will examine: (1) examples of conductance distribution broadening behavior over time, (2) scatterplot of conductance deviation vs. conductance, (3) standard deviation vs. time for different starting conductance values. First let us load the data:

[6]:
```
# Load data for technology
colnames = ["addr", "time", "r", "g", "gi", "range", "timept"]
data = pd.read_csv(f"data/tech{TECH}/relaxdata.min.tsv.gz", names=colnames,
 ↪sep='\t')
data.head()
```

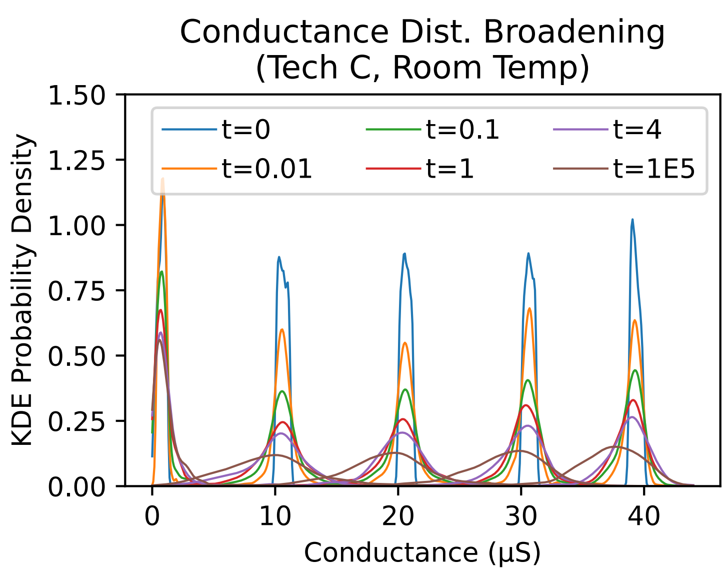
[6]:

	addr	time	r	g	gi	range	timept
0	10354	0.0	2.082326e+06	4.802322e-07	4.802322e-07	0	0.0
1	10191	0.0	2.375569e+06	4.209518e-07	4.209518e-07	0	0.0
2	13027	0.0	1.586060e+06	6.304932e-07	6.304932e-07	0	0.0
3	7397	0.0	2.592918e+06	3.856659e-07	3.856659e-07	0	0.0
4	2216	0.0	2.542619e+06	3.932953e-07	3.932953e-07	0	0.0

1.3.1 Conductance distribution broadening behavior

Below are examples of conductance broadening behavior over time:

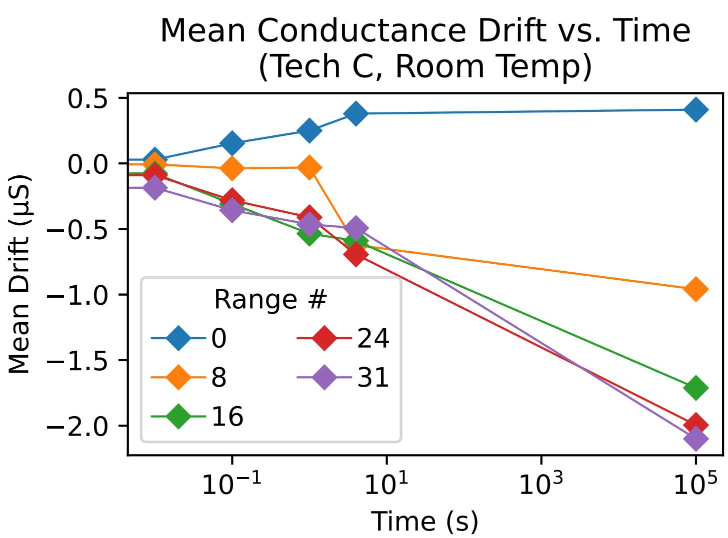
[7]:
```
# Select ranges to study
ranges = [0, 8, 16, 24, 31]

# Conductance broadening behavior
fig = plt.figure(figsize=(4, 2.7))
ax = fig.add_subplot(111)
ax.set_title(f"Conductance Dist. Broadening\n(Tech {TECH}, Room Temp)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for time, color in zip(settings["times"], colors):
    for r in ranges:
        gx = np.linspace(0, settings["gmax"]*1.1e6, 500)
        gvals = data[(data["range"] == r) & (data["timept"] == time)]["g"]
        pdf = scipy.stats.gaussian_kde(gvals*1e6).pdf(gx)
        label = (f"t={time}" if time < 100 else f"t=1E{int(np.log10(time))}")␣
↪if r == 0 else None
        plt.plot(gx, pdf, color=color, label=label, linewidth=0.8)
ax.legend(ncol=3, handletextpad=0.2)
ax.set_ylim(*settings["gbroad_ylim"])
ax.set_xlabel("Conductance (µS)")
ax.set_ylabel("KDE Probability Density")
plt.savefig(f"figs/tech{TECH}/broadening-time.pdf", bbox_inches="tight")
plt.show()
```



1.3.2 Mean Drift and Variance Growth (Time Dependence)

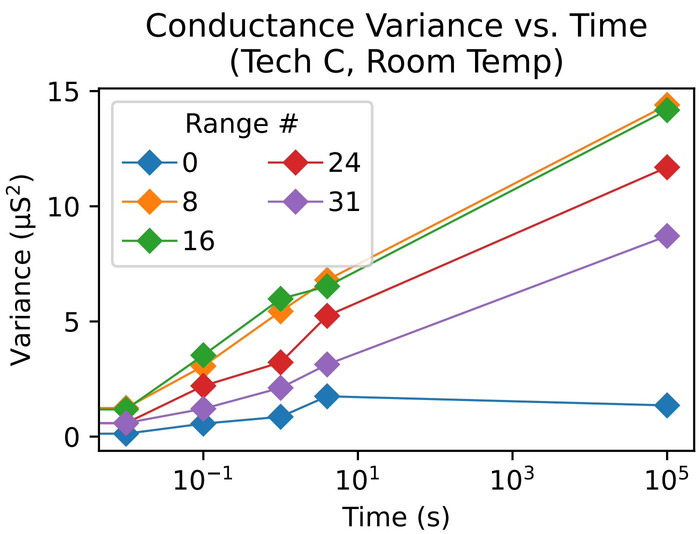
Here, we examine the drift of the distribution mean and variance growth as a function of time.

[8]:
```
# Mean drift behavior (time dependence)
fig = plt.figure(figsize=(4, 2.5))
ax = fig.add_subplot(111)
ax.set_title(f"Mean Conductance Drift vs. Time\n(Tech {TECH}, Room Temp)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for r, color in zip(ranges, colors):
    d = data[(data["range"] == r) & (data["gi"] <= settings["gmax"])]
    gi = d.groupby("timept").mean()["gi"]*1e6
    gf = d.groupby("timept").mean()["g"]*1e6
    deltag = gf - gi
    plt.plot(deltag.index, deltag, '-D', color=color, label=r, linewidth=0.8)
ax.legend(title="Range #", ncol=2 if TECH == 'C' else 3, handletextpad=0.2)
if "gmeandrift_t_ylim" in settings:
    ax.set_ylim(*settings["gmeandrift_t_ylim"])
ax.set_xscale("log")
ax.set_xlabel("Time (s)")
ax.set_ylabel("Mean Drift (µS)")
plt.savefig(f"figs/tech{TECH}/mean-drift-vs-time.pdf", bbox_inches="tight")
plt.show()
```



[9]:
```
# Variance behavior (time dependence)
fig = plt.figure(figsize=(4, 2.5))
ax = fig.add_subplot(111)
ax.set_title(f"Conductance Variance vs. Time\n(Tech {TECH}, Room Temp)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
```

```python
for r, color in zip(ranges, colors):
    d = data[(data["range"] == r) & (data["gi"] <= settings["gmax"])]
    gi = d.groupby("timept").mean()["gi"]*1e6
    gfvar = d.groupby("timept").var()["g"]*(1e6**2)
    plt.plot(gfvar.index, gfvar, '-D', color=color, label=r, linewidth=0.8)
ax.legend(title="Range #", ncol=3 if TECH == 'A' else 2, handletextpad=0.2)
if "gvar_t_ylim" in settings:
    ax.set_ylim(*settings["gvar_t_ylim"])
ax.set_xscale("log")
ax.set_xlabel("Time (s)")
ax.set_ylabel("Variance (µS$^2$)")
plt.savefig(f"figs/tech{TECH}/var-vs-time.pdf", bbox_inches="tight")
plt.show()
```

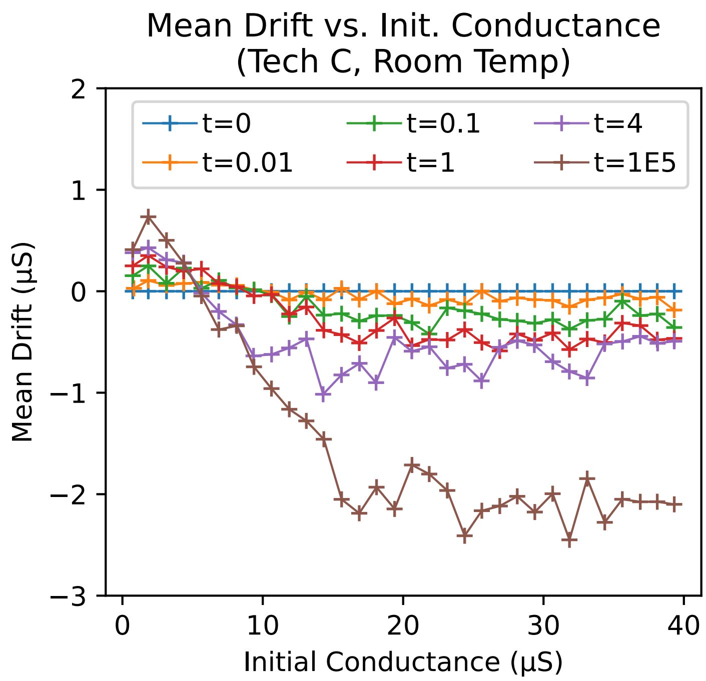


1.3.3 Mean Drift and Variance Growth (Conductance Dependence)

Here, we examine the drift of the distribution mean and variance growth as a function of conductance.

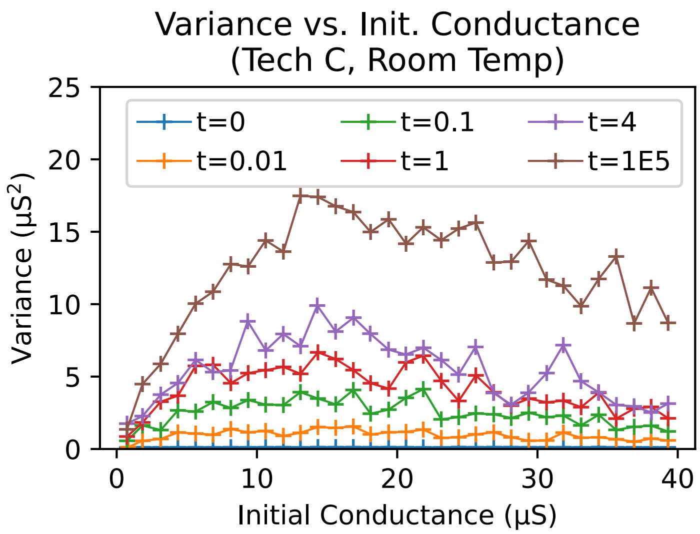
[10]:
```python
# Mean drift behavior (conductance dependence)
fig = plt.figure(figsize=(4, 3.5))
ax = fig.add_subplot(111)
ax.set_title(f"Mean Drift vs. Init. Conductance\n(Tech {TECH}, Room Temp)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for time, color in zip(settings["times"], colors):
    d = data[(data["timept"] == time) & (data["gi"] <= settings["gmax"])]
    gi = d.groupby("range").mean()["gi"]*1e6
    gf = d.groupby("range").mean()["g"]*1e6
    deltag = gf - gi
```

```
    label = (f"t={time}" if time < 100 else f"t=1E{int(np.log10(time))}")
    plt.plot(gi, deltag, '-+', color=color, label=label, linewidth=0.8)
ax.legend(ncol=3, handletextpad=0.2)
ax.set_ylim(*settings["gmeandrift_gi_ylim"])
ax.set_xlabel("Initial Conductance (µS)")
ax.set_ylabel("Mean Drift (µS)")
plt.savefig(f"figs/tech{TECH}/mean-drift-vs-g.pdf", bbox_inches="tight")
plt.show()
```



[11]:
```
# Variance behavior (conductance dependence)
fig = plt.figure(figsize=(4, 2.5))
ax = fig.add_subplot(111)
ax.set_title(f"Variance vs. Init. Conductance\n(Tech {TECH}, Room Temp)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for time, color in zip(settings["times"], colors):
    d = data[(data["timept"] == time) & (data["gi"] <= settings["gmax"])]
    gi = d.groupby("range").mean()["gi"]*1e6
    gfvar = (d.groupby("range")).var()["g"]*(1e6**2)
    label = (f"t={time}" if time < 100 else f"t=1E{int(np.log10(time))}")
    plt.plot(gi, gfvar, '-+', color=color, label=label, linewidth=0.8)
ax.legend(ncol=3, handletextpad=0.2)
ax.set_ylim(*settings["gvar_gi_ylim"])
ax.set_xlabel("Initial Conductance (µS)")
ax.set_ylabel("Variance (µS$^2$)")
```

```
plt.savefig(f"figs/tech{TECH}/var-vs-g.pdf", bbox_inches="tight")
plt.show()
```



1.4 Temperature dependence analysis via 1hr bake

Here, we analyze the effect of baking on the distribution broadening. In particular, we will examine examples of the conductance distributions broadening for different temperatures and then analyze the temperature-dependent mean drift and variance. We can first load the data and preprocess it a little bit:

[12]:
```
# Load data for technology
colnames = ["addr", "time", "r", "g", "temp"]
data = pd.read_csv(f"data/tech{TECH}/bake.tsv.gz", names=colnames, sep='\t')

# Get conductance range
data["gi"] = data.groupby("addr")["g"].transform("first")
data["range"] = np.int32(data["gi"] / settings["gmax"] * 32)

# Show data
data.head()
```

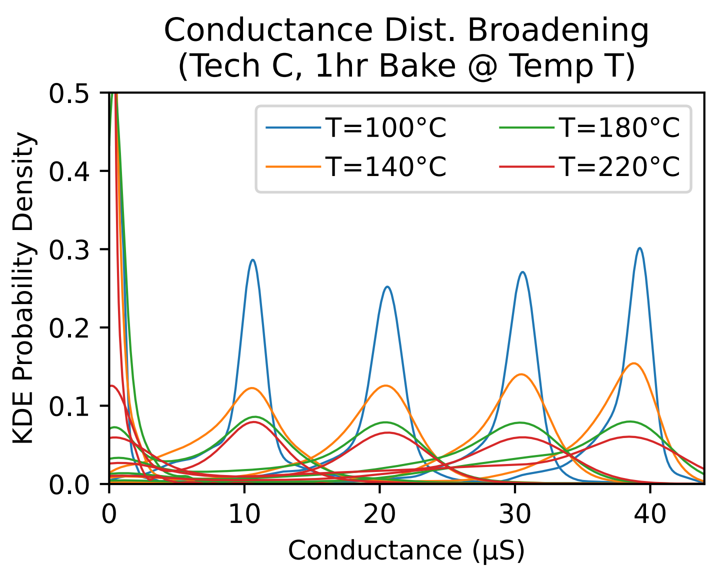
[12]:

	addr	time	r	g	temp	gi	range
0	0	0.129653	25947.041683	0.000039	100	0.000039	30
1	1	0.102764	69769.620577	0.000014	100	0.000014	11
2	2	0.115700	33780.050307	0.000030	100	0.000030	23
3	3	0.105721	28367.615486	0.000035	100	0.000035	28
4	4	0.098823	30310.430310	0.000033	100	0.000033	26

1.4.1 Conductance distribution broadening behavior

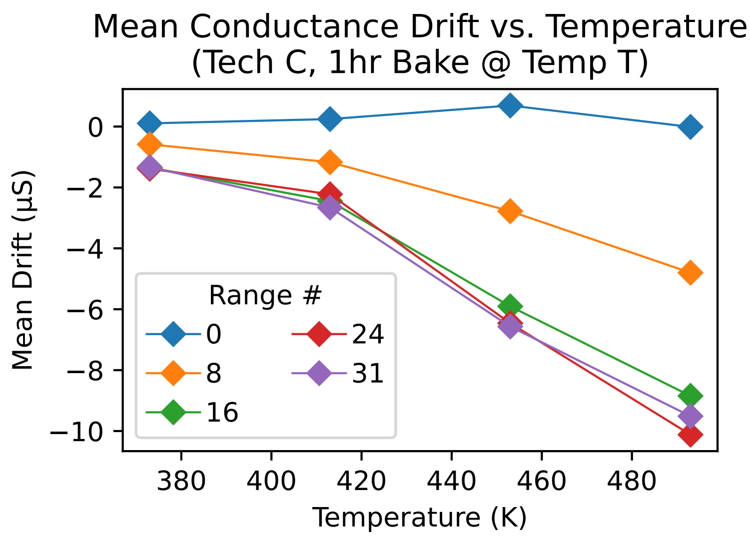
Below are examples of conductance broadening behavior at different temperatures:

[13]:
```
# Select ranges to study
ranges = [0, 8, 16, 24, 31]

# Conductance broadening behavior
fig = plt.figure(figsize=(4, 2.7))
ax = fig.add_subplot(111)
ax.set_title(f"Conductance Dist. Broadening\n(Tech {TECH}, 1hr Bake @ Temp T)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for temp, color in zip(settings["temps"], colors):
    for r in ranges:
        gx = np.linspace(0, settings["gmax"]*1.1e6, 500)
        gvals = data[(data["range"] == r) & (data["temp"] == temp) &␣
↪(data["time"] != 0)]["g"]
        pdf = scipy.stats.gaussian_kde(gvals*1e6).pdf(gx)
        plt.plot(gx, pdf, color=color, label=f"T={temp}°C" if r==0 else None,␣
↪linewidth=0.8)
ax.legend(ncol=2, handletextpad=0.2)
ax.set_xlim(0,settings["gmax"]*1.1e6)
ax.set_ylim(*settings["gbroad_temp_ylim"])
ax.set_xlabel("Conductance (µS)")
ax.set_ylabel("KDE Probability Density")
plt.savefig(f"figs/tech{TECH}/broadening-temp.pdf", bbox_inches="tight")
plt.show()
```



[14]:

```
# Mean drift behavior (temperature dependence)
fig = plt.figure(figsize=(4, 2.5))
ax = fig.add_subplot(111)
ax.set_title(f"Mean Conductance Drift vs. Temperature\n(Tech {TECH}, 1hr Bake @␣
↪Temp T)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for r, color in zip(ranges, colors):
    d = data[(data["range"] == r) & (data["gi"] <= settings["gmax"])]
    gi = d.groupby("temp").mean()["gi"]*1e6
    gf = d.groupby("temp").mean()["g"]*1e6
    deltag = gf - gi
    plt.plot(deltag.index + 273, deltag, '-D', color=color, label=r,␣
↪linewidth=0.8)
ax.legend(title="Range #", ncol=2 if TECH == 'C' else 3, handletextpad=0.2)
if "gmeandrift_temp_ylim" in settings:
    ax.set_ylim(*settings["gmeandrift_temp_ylim"])
ax.set_xlabel("Temperature (K)")
ax.set_ylabel("Mean Drift (µS)")
plt.savefig(f"figs/tech{TECH}/mean-drift-vs-temp.pdf", bbox_inches="tight")
plt.show()
```



[15]:

```
# Variance behavior (temperature dependence)
fig = plt.figure(figsize=(4, 3.5))
ax = fig.add_subplot(111)
ax.set_title(f"Conductance Variance vs. Temperature\n(Tech {TECH}, 1hr Bake @␣
↪Temp T)")
colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
for r, color in zip(ranges, colors):
```

```
    d = data[(data["range"] == r) & (data["gi"] <= settings["gmax"])]
    gfvar = d.groupby("temp").var()["g"]*(1e6**2)
    plt.plot(gfvar.index + 273, gfvar, '-D', color=color, label=r, linewidth=0.
↪8)
ax.legend(title="Range #", ncol=2, handletextpad=0.2)
if "gvar_temp_ylim" in settings:
    ax.set_ylim(*settings["gvar_temp_ylim"])
ax.set_xlabel("Temperature (K)")
ax.set_ylabel("Variance (µS$^2$)")
plt.savefig(f"figs/tech{TECH}/var-vs-temp.pdf", bbox_inches="tight")
plt.show()
```