Chapter 11 INFRARED STIMULATED LUMINESCENCE SIGNALS: DATA ANALYSIS

Code 11.1: CW-IRSL data fitted with KP-CW equation

```
# J1000 CW-IRSL data fitted with KP-CW equation
from scipy import optimize
from sympy import *
import numpy as np
import matplotlib.pyplot as plt
from prettytable import PrettyTable
import warnings
warnings.filterwarnings("ignore")
data = np.loadtxt('J1000cwirsl.txt')
x_data,y_data=data[:,0][1:800], data[:,1][1:800]
y_data=y_data/max(y_data)
plt.subplot(2,1, 1);
def test_func(x, imax_fit,rho_fit, A_fit,bgd_fit):
    return imax_fit*np.exp (-rho_fit*(np.log(1 + A_fit*x))\
** 3.0)*(np.log(1+A_fit*x)**2.0)/(1+x*A_fit)+bgd_fit
params, cov = optimize.curve_fit(test_func,\)
x_data, y_data)
drho= round(np.sqrt(cov[1][1]),5)
dA = round(np.sqrt(cov[2][2]), 2)
dimax = round(np.sqrt(cov[0][2]),2)
plt.scatter(x_data, y_data, label='J1000 feldspar');
plt.plot(x_data, test_func(x_data, *params[0:4]),
label='KP-CW equation');
leg = plt.legend()
leg.get_frame().set_linewidth(0.0)
plt.ylabel('CW-IRSL signal [a.u.]');
```

```
plt.xlabel('Time [s]');
plt.subplot(2,1, 2);
plt.plot(x_data,test_func(x_data, *params[0:4])-\
y_data, "o", label='Residuals');
leg = plt.legend();
leg.get_frame().set_linewidth(0.0)
plt.ylabel('Residuals');
plt.xlabel('Time [s]');
plt.ylim(-.1,.1);
plt.tight_layout();
imax,rho, A, bgd=int(params[0]),round(params[1],5),\
round(params[2],2),round(params[-1],3)
res=test_func(x_data, *params)-y_data
FOM=round(100*np.sum(abs(res))/np.sum(y_data),2)
myTable=PrettyTable(["A",'dA', "rho", "d(rho)",\
"s'(s^-1)","ds'",'bgd']);
myTable.add_row([imax,dimax,rho,drho, A, dA,bgd]);
print('FOM=',FOM,' %')
 FOM= 3.64 %
print(myTable)
  +---+----+
  | A | dA | rho | d(rho) | s'(s^-1) | ds' | bgd |
  +--+---+
  | 2 | 0.08 | 0.01157 | 0.0003 | 12.07 | 0.33 | 0.033 |
  +---+----+
plt.show();
```

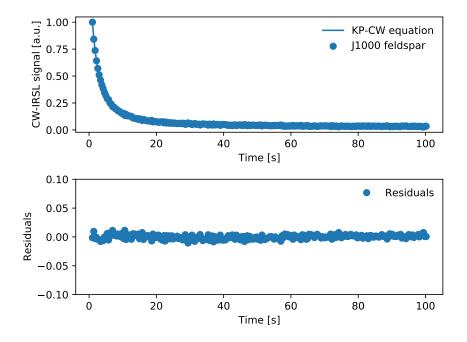


Fig. 11.1: Experimental CW-IRSL curve from a freshly irradiated J1000 feldspar sample, fitted using the KP-CW analytical Eq.(??). For more details and examples from several other types of feldspars, see Pagonis et al. [1].

Code 11.2: LM-IRSL data fitted with KP-LM equation

```
# Deconvolution with the KP-LM equation, fixed bgd
from scipy import optimize
import numpy as np
import matplotlib.pyplot as plt
from prettytable import PrettyTable
import warnings
warnings.filterwarnings("ignore")
data = np.loadtxt('J1000 50degC LMIRSL.txt')
x_data,y_data=data[:,0], data[:,1]
y_data=y_data/max(y_data)
def KVLMOSL(x, A,sprime,rho):
    F=np.log(1+abs(sprime)*x**2/(2*P))
```

```
LMOSL= abs(A)*np.exp (-abs(rho)*\
F** 3.0)*(F**2.0)/(1+abs(sprime)*x**2/(2*P))
   return LMOSL
def total_LMOSL(x, *inis):
   u=np.array([0 for i in range(len(x_data))])
   As, sprimes=inis[0:nPks], inis[nPks:2*nPks]
    rho=inis[-1]
   for i in range(nPks):
       u=u+KVLMOSL(x,As[i],sprimes[i],rho)
    u=u+bgd*x/P
    return u
nPks=2
P=int(max(x_data))
bgd=y_data[-1]
inis=[1,1,3,100,.01]
params, cov = optimize.curve_fit(total_LMOSL,\
x_data, y_data,p0=inis)
plt.scatter(x_data, y_data, label='J1000 feldspar');
plt.plot(x_data, total_LMOSL(x_data, *params),'r-',
        label='KP-LM equation',linewidth=3);
for i in range(0,nPks):
   FOKLMi=KVLMOSL(x_data, params[i], params[i+nPks],
   params[-1]);
    plt.plot(x_data,FOKLMi);
plt.plot(x_data,x_data*bgd/P);
leg = plt.legend()
leg.get_frame().set_linewidth(0.0)
plt.ylabel('LM-IRSL signal [a.u.]');
plt.xlabel('Time [s]');
leg = plt.legend()
leg.get_frame().set_linewidth(0.0)
A1,sprime1,A2,sprime2=[round(params[x],2) for x in range(4)]
rho, drho=f'{params[-1]:.4f}',f'{np.sqrt(cov[4][4]):.4f}'
dA1,dsprime1,dA2,dsprime2=[round(np.sqrt(cov[x][x]),2)\
for x in range(4)]
res=total_LMOSL(x_data, *params)-y_data
FOM=round(100*np.sum(abs(res))/np.sum(y_data),2)
myTable=PrettyTable(["A",'dA', "rho", "d(rho)",\
"s'(s^-1)","ds","FOM"]);
myTable.add_row([A1,dA1,rho,drho, sprime1, dsprime1,FOM]);
myTable.add_row([A2,dA2,' ',' ', sprime2, dsprime2,' ']);
print(myTable)
  +----+
| A | dA | rho | d(rho) | s'(s^-1) | ds | FOM |
```

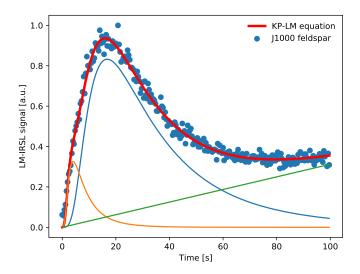
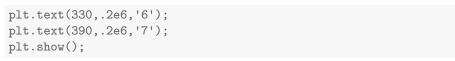


Fig. 11.2: Deconvolution of LM-IRSL signal for sample J1000 based on the KP-LM equation, with two components plus a linearly increasing background. For more details see Pagonis et al. [1]

Code 11.3: J1000 deconvolution with optimal number of peaks N=7

```
# J1000 deconvolution with optimal number of peaks N=7
from scipy import optimize
import numpy as np
import matplotlib.pyplot as plt
from prettytable import PrettyTable
import warnings
warnings.filterwarnings("ignore")
data = np.loadtxt('J1000prompt.txt')
x_data,y_data = data[:, 0], data[:, 1]
z, kB =1.8, 8.617E-5
```

```
def TL(T, B,En ,rho, s):
    return abs(B)* np.exp(-rho*( (np.log(1+z*s*kB*(((T+\
    273)**2.0)/np.abs(En))*np.exp(-En/(kB*(T+273)))*
    (1-2*kB*(T+273)/En)))**3.0))*(En**2.0-6*(kB**2.0)*
     ((T+273)**2.0))*((np.log(1+z*s*kB*(((T+273)**2.0)/
    abs(En))*np.exp(-En/(kB*(T+273)))*(1-2*kB*(T+273)/
    En)))**2.0)/(En*kB*s*((T+273)**2)*z-2*(kB**2.0)*
    s*z*((T+273)**3.0)+np.exp(En/(kB*(T+273)))*En)
def total_TL(T, *inis):
    u=np.array([0 for i in range(len(x_data))])
    Bs, Ens,rho,s= inis[0:nPks], inis[nPks:2*nPks],inis[-2]\
    ,inis[-1]
    for i in range(nPks):
        u=u+TL(T,Bs[i],Ens[i],rho,s)
    return u
nPks=7
B=[2e17]*7
lowB, highB=[0.01*x \text{ for } x \text{ in B}], [50*x \text{ for } x \text{ in B}]
lowrho, highrho, rho= [0.003,.015,.008]
lows, highs, s =[1e11, 1e14, 1e12]
En=[.8, .9,1.06,1.19,1.4,1.6,1.7]
lowEn, highEn = [0.9*x \text{ for } x \text{ in En}], [1.1*x \text{ for } x \text{ in En}]
inis=B+En+[rho]+[s]
lowbnds=lowB+lowEn+[lowrho]+[lows]
highbnds=highB+highEn+[highrho]+[highs]
params, params_covariance = optimize.curve_fit(total_TL,\)
x_data,y_data,p0=inis,bounds=(lowbnds,highbnds),maxfev=10000)
plt.scatter(x_data, y_data,c='r',label='Sample J1000');
plt.plot(x_data, total_TL(x_data,
 *params),c='black',label='KP-TL N='+str(nPks),linewidth=1);
for i in range(0,nPks):
    plt.plot(x_data, TL(x_data, params[i],params[nPks+i],\
    params[-2],params[-1]));
leg = plt.legend();
leg.get_frame().set_linewidth(0.0);
plt.ylabel('TL [a.u.]');
plt.xlabel(r'Temperature T [$^{o}$C]');
res=total_TL(x_data, *params)-y_data
FOM=100*np.sum(abs(res))/np.sum(y_data)
plt.text(90,.2e6,'1');
plt.text(110,1e6,'2');
plt.text(150,1e6,'3');
plt.text(210,.4e6,'4');
plt.text(280,.3e6,'5');
```



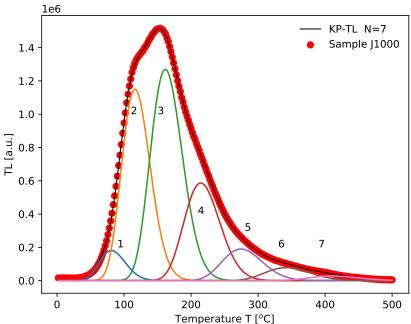
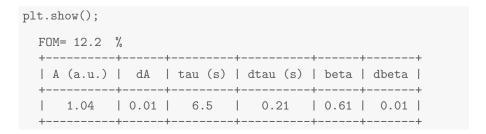


Fig. 11.3: Deconvolution of prompt TL signal from sample J1000 using the original KP-TL equation, with the optimal number of peaks N=7. For more details see Pagonis et al. [1] and Chaps.4-5 of this book.

Code 11.4: Fitting CW-IRSL signal with stretched exponential

```
#Fitting CW-IRSL signal with stretched exponential
# deconvolution of feldspar CW-IRSL with stretched exponential
from scipy import optimize
import numpy as np
import matplotlib.pyplot as plt
from prettytable import PrettyTable
```

```
import warnings
data = np.loadtxt('KST4ph300IR.txt')
x_data,y_data=data[:,0][0:100], data[:,1][0:100]
y_data=y_data/max(y_data)
def STRETCH(t, A,tau,beta):
    CW=A*np.exp(-(t/tau)**beta)
    return CW
def total_CW(t, *inis):
    u=np.array([0 for i in range(len(x_data))])
    As, taus=
                 inis[0:nPks], inis[nPks:2*nPks]
    beta=inis[-1]
    for i in range(nPks):
        u=u+STRETCH(t,As[i],taus[i],beta)
    return u
P=int(max(x_data))
t = np.linspace(0, P, P)
nPks=1
inis=[1,5,.1]
params, cov = optimize.curve_fit(total_CW,\)
x_data,y_data,p0=inis,maxfev=10000)
plt.scatter(x_data, y_data,c='r',label='KST4 feldsparCW-IRSL ');
plt.plot(x_data, total_CW(x_data,
 *params),c='black',label='Stretched Exponential equation');
for i in range(0,nPks):
    CWi=STRETCH(t, params[i],params[nPks+i],params[-1]);
    plt.plot(t,CWi);
leg = plt.legend();
leg.get_frame().set_linewidth(0.0);
plt.ylabel('CW-IRSL [a.u.]');
plt.xlabel(r'Stimulation time [s]');
res=total_CW(x_data, *params)-y_data
FOM=100*np.sum(abs(res))/np.sum(y_data)
As=[round(x,2) for x in params[0:nPks]]
taus=[round(x,1) for x in params[nPks:2*nPks]]
dAs=[round(np.sqrt(cov[x][x]),2) for x in range(0,nPks)]
dtaus=[round(np.sqrt(cov[x][x]),2) for x in\
range(nPks,2*nPks)]
dbeta=round(np.sqrt(cov[2][2]),2)
beta=round(params[-1],2)
myTable = PrettyTable([ "A (a.u.)", "dA", \
'tau (s)','dtau (s)','beta','dbeta'])
myTable.add_row([As[0],dAs[0],taus[0],dtaus[0],beta,dbeta]);
print('FOM=',round(FOM,1),' %')
print(myTable)
```



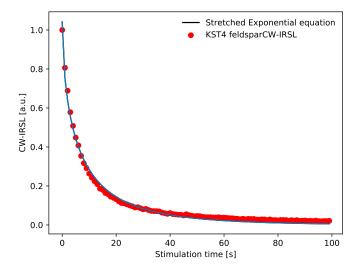


Fig. 11.4: Fitting CW-IRSL signal with the stretched exponential Eq.(??). The experimental data is from feldspar sample KST4 (Polymeris et al. [2])

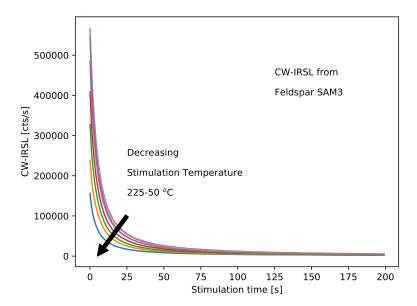


Fig. 11.5: The experimental CW-IRSL curves from feldspar sample SAM3, measured at elevated stimulation temperatures in the range 50-225°C. For experimental details see Şahiner et al. [3]

Code 11.5: Analysis of SAM3 CW-IRSL signals using stretched exponentials

```
# Analysis of CW-IRSL measured at high stimulation temeprature
from scipy import optimize
import numpy as np
import matplotlib.pyplot as plt
[betas, dbetas, As,dAs,taus,dtaus,maxys,bgds,dbgds]= [[] \
for _ in range(9)]
fils=['irrthenst50.txt','irrthenst75.txt','irrthenst100.txt',
'irrthenst125.txt','irrthenst150.txt','irrthenst175.txt',
'irrthenst200.txt','irrthenst225.txt']
def total_CW(t, *inis):
    [As, taus, betas,bgd]= inis
    CW=As*np.exp(-(t/abs(taus))**np.abs(betas))+np.abs(bgd)
    return CW
```

```
def findbeta(fil):
    data = np.loadtxt(fil)
    y_data=data
    x_data=np.arange(0,len(y_data))
    maxys.append(max(y_data))
    y_data=y_data/max(y_data)
    inis=[1,5,.1,min(y_data)]
    params, cov = optimize.curve_fit(total_CW,\
    x_data,y_data,p0=inis,maxfev=10000)
    res=total_CW(x_data, *params)-y_data
    FOM=100*np.sum(abs(res))/np.sum(y_data)
    As.append(params[0]);
    taus.append(params[1])
    betas.append(params[2])
    bgds.append(params[3])
    dAs.append(np.sqrt(cov[0][0]))
    dtaus.append(np.sqrt(cov[1][1]))
    dbetas.append(np.sqrt(cov[2][2]))
    dbgds.append(np.sqrt(cov[3][3]))
    return
for j in range(len(fils)):
    findbeta(fils[j])
temps=[50,75,100,125,150,175,200,225]
plt.subplot(2,2,1);
plt.errorbar(temps, maxys, marker='s',yerr=np.sqrt(maxys),\
label=r'Amplitude A$');
plt.ylim(0,6e5);
plt.ylabel(r'Amplitude A$');
plt.xlabel(r'Stimulation T [$^o$C]');
plt.subplot(2,2,2);
plt.errorbar(temps, betas, marker='s',yerr=dbetas,\
label=r'Coefficient $\beta$');
plt.ylim(0.6,.7);
plt.ylabel(r'Coefficient $\beta$');
plt.xlabel(r'Stimulation T [$^o$C]');
leg = plt.legend();
leg.get_frame().set_linewidth(0.0);
plt.subplot(2,2,3);
plt.errorbar(temps, np.abs(taus), marker='s',yerr=dtaus);
plt.ylim(0,8);
plt.ylabel(r'$\tau$ [s]');
plt.xlabel(r'Stimulation T [$^o$C]');
plt.subplot(2,2,4);
plt.errorbar(temps,np.abs(bgds),marker='o',yerr=dbgds);
```

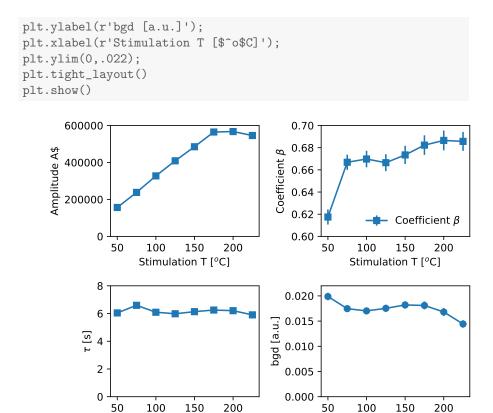


Fig. 11.6: The experimental CW-IRSL curves from Fig.11.5 for feldspar sample SAM3 are analyzed using the stretched exponential function. The data is measured at elevated stimulation temperatures in the range 50-225°C, and is plotted as a function of the stimulation temperature. (a) Amplitude A; (b) coefficient β ; (c) decay time τ ; (d) constant background component. For experimental details see Sahiner et al. [3].

200

100

150

Stimulation T [°C]

200

50

100

150

Stimulation T [°C]

References

- V. Pagonis, N. D. Brown, J. Peng, G. Kitis, G. S. Polymeris, On the deconvolution of promptly measured luminescence signals in feldspars, Journal of Luminescence 239 (2021) 118334. doi:https://doi.org/10.1016/j.jlumin.2021.118334. URL https://www.sciencedirect.com/science/article/pii/S0022231321004506
- G. Polymeris, E. Theodosoglou, G. Kitis, N. Tsirliganis, A. Koroneos, K. Paraskevopoulos, Preliminary results on structural state characterization of K-feldspars by using thermoluminescence, Mediterranean Archaeology and Archaeometry 13 (3) (2013) 155–161.
- 3. E. Şahiner, G. Kitis, V. Pagonis, N. Meriç, G. S. Polymeris, Tunnelling recombination in conventional, post-infrared and post-infrared multi-elevated temperature IRSL signals in microcline K-feldspar, Journal of Luminescence 188 (2017) 514–523.