Analysis of microhardness data in Tl_xln_{1-x}Se

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Microhardness measurements have been performed on TI_xIn_{1-x} Se semiconductors ($x = 0.0, 0.2, 0.3 \dots 1.0$). The microhardness, H, as a function of x shows a maximum at x ca. 0.5. A statistical evaluation of the obtained results leads to a log-normal distribution of the microhardness rather than, as expected, a polynomial one.

1. Introduction

TlInSe₂ and some related compounds are ternary semiconductors with very unusual electrical properties. In some recent publications [1–4] their S-type current–voltage characteristic and the chaotic voltage oscillations appearing in the corresponding NDRregion are examined. In another report the influence of the stoichiometry is examined [5], i.e. how changes of x in the system $Tl_xIn_{1-x}Se$ affect the non-linear electrical properties. However, the mechanical properties of the $Tl_xIn_{1-x}Se$ compounds had not been examined in a systematic way. The aim of the present work is to report some data concerning the dependence of microhardness on x.

2. Experimental procedure

The samples used for the microhardness measurements were $Tl_xIn_{1-x}Se$ (with $x = 0, 0.2, 0.3 \dots 1.0$). These compositions form the InSe–TlSe quasi-binary cross-section of the concentration triangle of the ternary system In–Tl–Se (Fig. 1). The crystals were prepared by crystallization from stoichiometric melts in evacuated ampoules.

Being too small (a few cubic millimetres) and of irregular shape, the samples were encapsulated in dental resin (Kallocryl CP-rz), enabling easy handling. After hardening, which took 24 h, the capsules were mechanically polished with 5 mm alumina powder until flat surfaces of the samples were obtained. Microscopic observation revealed that these surfaces consisted of lapped and cleaved areas. Indentations were made on both areas.

Microhardness measurements were carried out at room temperature with PMT-3 microhardness tester [6]. The Vickers diamond indentor (a square-based pyramid) was maintained in contact with the sample's surface for 60 s. This was an unusually long time, but preliminary measurements showed that shorter contact times gave unstable indentations. The indentor was loaded with 10 g. The indentation marks produced were of irregular contours; cracks could be seen around the indentations due to material brittleness.

The microhardness of the $\text{Tl}_x \text{In}_{1-x}$ Se samples as a function of x is presented in Fig. 2. In order to reduce experimental uncertainty, at least five, and typically 10, indentations on a given sample were produced. It can be seen that the dependence was characterized by a maximum at x = 0.3-0.5. TISe is harder than InSe, a fact consistent with its broader gap. The relatively large scattering of experimental data is presumably due to: (1) the samples not being monocrystalline; and (2) the indentations having irregular contours. It can be seen that the scattering of the data is smaller in the case of the binary compounds.

According to the Kurnakov rules for the variation of hardness with composition in metal solid solutions [7]:

1. The formation of solid solutions is accompanied by an increase in hardness.

2. The variation of hardness over a continuous compositional range of solid solutions yields a continuous curve possessing a maximum.

Later, Goryunova et al. [6] showed that these rules were also valid for semiconductor solid solutions. Hence, our results represent another verification of Kurnakov rules. Goryunova et al. assumed that their behaviour might be due to "healing of defects" which took place in certain alloy compositions. But the following consideration may also be taken into account: production of an indentation is accompanied by material transfer to the adjacent parts of the crystal. This transfer is facilitated by the presence of defects, e.g. dislocations – the bigger the dislocation density, the softer the material. On the other hand, it is known that the dislocation density in some materials can be reduced by doping, as in the case of GaAs and InP doped with Zn, S, Te, Al or N [8,9]. Several explanations have been proposed to aid understanding this experimental fact. One of them (the so-called "pinning effect") is based on the assumption that dislocations



Figure 1 The In-Tl-Se concentration triangle with the InSe-TlSe quasibinary cross-section.



Figure 2 Microhardness, H, of the $Tl_xIn_{1-x}Se$ samples versus x. The best fit using Equation 1 is shown by the curve in the middle and the 95% confidence limits are shown as the upper and lower curves, respectively.

are "stopped" by the impurity atoms, due to the stronger bonds the latter are forming. We suppose that the formation of mixed crystals may be accompanied by such a decrease in dislocation density.

3. Analysis and discussion

One of the main reasons for the large statistical deviations in microhardness measurements on semiconductor compounds and their solid solutions is related to the lack of precise measuring methods due to their high brittleness in comparison with metals and alloys.

The study of a great number of solid solutions [6, 7] has shown that the microhardness dependence on composition in quasi-binary systems $A_xB_{1-x}C$ can be expressed by the following empirical equation, which holds also for many other physical quantities (like the bandgap, lattice parameter, etc.)

$$H(x) = H_{\rm A}x + H_{\rm B}(1-x) + Kx(1-x)$$
(1)

where H_A and H_B are the values of the microhardness of the pure constituents of the solid solution, x, 1 - x are their corresponding concentrations in mole fractions and K a constant. To explain such variations of the properties of a mixed crystal, one has to recall the fact that the mole fraction, x, is a random variable bound to the limits x = 0 and x = 1 [10]. Composition fluctuations in a mixed crystal can be described by the deviation Δx of the fractional concentration of the A-type atoms which has an average value of x. If we choose a volume R^3 with linear size R, the average number of A-type atoms is $\overline{N}_A = N_A R^3$, and the total number of lattice sites is $\overline{N}N = NR^3$. The local deviation Δx in the concentration of the A-atoms (the fluctuation in composition) is given by

$$\Delta x = \Delta N_{\rm A} / NR^3 = (x NR^3)^{1/2} / NR^3$$
 (2)

The most simple assumption is that A- and B-type atoms in the solid solution are distributed in a completely random way. Then the random function $\Delta x(r)$ $= [\bar{N}_A(r) - N_A]/\bar{N}$ which measures the local deviation in the concentration of A-type atoms in a small, but macroscopic, volume around point r will be a Gaussian random function of mean average equal to zero and with a pair correlation function

$$\langle \Delta x(r)\Delta x(r')\rangle = N^{-1}x(1-x)\delta(r-r')$$
(3)

where brackets denote averaging over the volume of the system, N is the concentration of all A- and B-type atoms and $\delta(r)$ is Dirac's delta function [10]. Now, for small fluctuations, one can expand the dependence

$$H[x(r)] = H(x) + [dH/dx]_x[x(r) - x]$$
(4)

The second term has a Gaussian distribution with a mean average value equal to zero and a root-mean square fluctuation equal to

$$\sigma = |dH/dx| [x(1-x)/\bar{N}]^{1/2}$$
(5)

Already in this simple model one can describe semiempirically the bandgap bowing, proportional to x(1-x) (see, for example, [11]). Thus, we have tried to explain the variation of our experimental data as a function of x assuming that the microhardness and its standard deviation are given by Equations 1 and 5, respectively.

The fitting of our data by means of Equation 1 is shown in Fig. 2 (middle curve – the upper and lower curves being the 95% confidence limits). The maximum in the dependence is at x = 0.53, consistent with the assumed x(1 - x) variation. The parameters of the curve $H(x) = a + bx + cx^2$ are listed in Table I, together with their standard error, *t*-value and 95% confidential limits, as well as other parameters evaluating statistical reliability of the model curve.

In contrast, the data on the standard deviation as a function of x shown in Fig. 3 cannot be explained assuming a dependence of the type of Equation 5. In fact, the experimental data show a maximum, whereas the prediction by means of Equation 5 has a minimum at compositions 0.4-0.5. We have found that using the log-normal distribution a much better fit can be obtained for both the microhardness and its standard deviation

$$f(x) = a + b \exp\{-0.5 \left[\ln(x/c)/d\right]^2\}$$
(6)

Rank 125 Eqn 1	$003 \ y = a + bx + cx^2$				
<i>r</i> ² Coef det 0.3631692648	DF adj r ² 0.34069288	59	Fit std err 0.6322438023	F-value 24.521866681	
Parm a b c	Value 0.608008952 5.614307952 - 5.31725739	Std error 0.210676751 0.840007103 0.759529095	t-value 2.885980291 6.683643424 – 7.0007290	95% confidence 1 0.189216144 3.944503710 3 — 6.82708379	imits 1.026801759 7.284112194 - 3.80743098
Area $X_{min}-X_{max}$ 1.6427437989 Function min 0.6080089525 1st Deriv min - 5.020206820 2nd Deriv min - 10.63451483 Soln vector Direct r^2 Coef det 0.3631692648 Source Regr Error	area precision 1.979984e-19 X-value 1.66924e-10 X-value 1.000000000 X-value 0.2619504842 Covar matrix LUDecomp DF adj r^2 0.3406928859 Sum of squares 19.604361 34.376971	Fun 2.08 1st 5.61 2nd - 1 Fit 0.63 DF 2 86	action max 199972912 Deriv max 43079501 Deriv max 10.63451464 std err 122438023 Mean square 9.8021803 0.39973223	X-value 0.5279326864 X-value 1.669243e-10 X-value 0.5631963212 F 24.5219	
Total X variable: X	53.981332	88			
X_{\min} : X_{mean} : $X@Y_{min}$:	0.000000000 0.5820224719 0.0000000000	X_{max} : X_{std} : $X@Y_{max}$:	1.0000000000 0.2933202554 0.5000000000	$ \begin{array}{c} X_{\text{range}}: \\ X_{\text{median}}: \\ X @ Y_{\text{range}}: \end{array} $	1.0000000000 0.600000000 0.5000000000
Y variable: micro Y_{min} : Y_{mean} : $Y@X_{min}$:	ohardness (GPa) 0.2930000000 1.6221011236 0.2930000000	Y_{max} : Y_{std} : $Y@X_{max}$:	3.7200000000 0.7832140373 1.3630000000	$Y_{range}:$ $Y_{median}:$ $Y @ X_{range}:$	3.4270000000 1.519000000 1.070000000



Figure 3 Experimental dependence of the standard deviation of H as a function of composition. The best fit is given by log-normal distribution (Equation 6) with 95% confidence limits shown by the upper and lower curves, respectively.

The corresponding parameters are given in Tables II and III, respectively, and the corresponding curves are shown in Fig. 3 for the standard deviation and in Fig. 4 for H(x), together with their confidence limits. Finally, in Fig. 5 the predicted dependences according to Equations 1 and 6 for the microhardness are compared. The calculated values of four most frequently used statistical criteria obtained from the fit of microhardness data by means of Equations 1 and 6 are given in the first rows of Tables I and II respectively. The comparison shows that the log-normal distribu-

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tion describes the experimental dependence H(x) in a much better way according to all four criteria. In case of the standard deviation, the log-normal distribution is the best physically plausible fitting compared to a variety of about 320 different functional forms which have been tried.

One possible explanation of these results may be related to the fact that the monocrystalline materials of this family possess highly anisotropic mechanical and other properties in the direction of the c-axis and in a plane perpendicular to it. In fact, our specimens are polycrystalline with crystallites randomly oriented to the plane of the indentation. In addition, the environment where successive indentations have been produced is quite different due to different contents of impurities, defects, dislocations, etc. In our case, the random results of the experiment are expressed by means of the area of the indentation which is affected by many different and independent causes. If the effects of them were additive, the asymptotic distribution according to the Central Limit Theorem would be the Gaussian one and equations analogous to Equations 1 and 5 would be valid for the hardness and its standard deviation. However, the assumption that the interaction between different causes is expressed by simple addition is not always plausible. A much more natural assumption is that every cause produces some pulse, the result of which is proportional to the strength of the pulse itself. Let us suppose that we have *n* pulses, R_1, R_2, \ldots, R_n , which are independent

Rank 1 Eqn 800	$5 y = a + b \exp\{-0.5 [\ln(b + b)]\}$	$x/c)/d$] ² } (log-n	ormal)				
<i>r</i> ² Coef det 0.7349451771	DF adj r ² 0.522901313	38	Fit std err 0.152245528	d err F-v 2455281 5.54		alue 156087842	
Parm a b c d	Value 0.055443217 0.644373725 0.376769410 0.594999379	Std error 0.148554950 0.167704662 0.044404413 0.173369504	<i>t</i> -valu 0.373 3:842 8.484 3.431	e 216892 312539 954251 972550	95 -	5% confidence lin - 0.30957134 0.232306382 0.267663270 0.169012945	nits 0.420457776 1.056441068 0.485875550 1.020985813
Area $X_{\min}-X_{\max}$ 0.4237358415 Function min 0.0554432166 1st Deriv min - 1.041522173 2nd Deriv min - 21.80618559	area precision 4.258404e-10 X-value 0.0010178024 X-value 0.5872161156 X-value 0.2644376475	Fu 0.6 1st 3.4 2no 41.	nction max 998169418 Deriv max 834031641 1 Deriv max 737971396	2 0 2 0 2 0 0 2 0 0 0 0	X-value).3767693954 X-value).1696689692 X-value).0889373823		
r ² Coef det 0.7349451771 Source Regr Error Total	DF adj r ² 0.5229013188 Sum of square: 0.38562002 0.1390722 0.52469223	Fit 0.1 5 DF 3 6 9	std err 522455281 7 Mear 0.128 0.023	square 54001 178701		F 5.54561	
X variable: $XX_{min}:X_{mean}:X@Y_{min}:$	0.000000000 0.540000000 0.000000000	$egin{array}{l} X_{ ext{max}}:\ X_{ ext{std}}:\ X@Y_{ ext{max}}: \end{array}$	1.000 0.320 0.300	0000000 4163958 0000000		X_{range} : X_{median} : $X @ Y_{range}$:	1.000000000 0.550000000 0.300000000
Y variable: micr Y_{min} : Y_{mean} : $Y@X_{min}$:	ohardness (GPa) 0.0528000000 0.4262860000 0.0528000000	Y _{max} : Y _{std} : Y@X _{max} :	0.842 0.241 0.216	0600000 4521406 2000000		Y _{range} : Y _{median} : Y@X _{range} :	0.7892600000 0.3643000000 0.1634000000

TABLE III Numeric summary for the best fit of H(x) by means of log-normal distribution, as shown in Figs 4 and 5

Rank 9 Eqn 8005 $v = a + b \exp\{-0.5 \lceil \ln(x/c)/d \rceil^2\}$ (log-normal)

Rank 9 Eqn 800	$y = a + b \exp\{-0.5 [m($	$x/c)/a]^{-} $ (log-noi	mai)			
r ² Coef det 0.4924053889	of det DF adj r ² 1053889 0.468234216		Fit std err 0.5677679028		F-value 27.485488840	
Parm a b c d	Value 0.388310213 1.944326059 0.396279869 0.639357395	Std error 0.211996011 0.231663235 0.022906725 0.086340097	t-value 1.831686412 8.392898673 17.29971737 7.405103985	95% confidence limits - 0.03317448 1.483739527 0.350737342 0.467698367		ts 0.809794906 2.404912590 0.441822395 0.811016423
Area X _{min} -X _{max} 1.5859052865 Function min 0.3883102130 1st Deriv min - 2.691100115 2nd Deriv min - 55.92070746	area precision 1.601936e-09 X-value 0.0006147410 X-value 0.6320473221 X-value 0.2633109703	Fun 2.332 1st I 9.870 2nd 118.	ction max 26362717 Deriv max 51445873 Deriv max 35652470	X-value 0.3962799668 X-value 0.1650914073 X-value 0.0808750802		
r ² Coef det 0.4924053889 Source Regr Error Total	DF adj r ² 0.4682342169 Sum of squares 26.580699 27.400633 53.981332	Fit s 0.56 DF 3 85 88	td err 77679028 Mean square 8.8602329 0.32236039	2	F 27.4855	
X variable: X X_{\min} : X_{mean} : $X@Y_{min}$:	0.000000000 0.5820224719 0.0000000000	$egin{array}{l} X_{\max}:\ X_{ m std}:\ X@Y_{ m max}: \end{array}$	1.000000000 0.293320255 0.500000000) 4)	X_{range} : X_{median} : $X@Y_{range}$:	1.000000000 0.600000000 0.500000000
Y variable: micr Y_{min} : Y_{mean} : $Y@X_{min}$:	ohardness (GPa) 0.2930000000 1.6221011236 0.2930000000	Y_{max} : Y_{std} : $Y@X_{max}$:	3.720000000 0.783214037 1.363000000) 3)	Y _{range} : Y _{median} : Y@X _{range} :	3.4270000000 1.5190000000 1.0700000000



Figure 4 The data from Fig. 2 are fitted by means of log-normal distribution (Equation 6) with 95% confidence limits shown by the lower and upper curves, respectively.



Figure 5 Comparison of the best fits from Figs 2 and 4 using curves of Equations 1 and 6.

random variables and let them affect the random variable X_0 (with an initial value X) so that after the action of $R_1 \ldots R_v$, it successively had the values $X_1 \ldots X_v$, respectively. Furthermore, suppose the increase in X due to the (v + 1)th pulse is proportional to R_{v+1} and to some function $g(X_v)$ of the value of X after the vth pulse

$$X_{\nu+1} = X_{\nu} + R_{\nu+1}g(X_{\nu})$$
 (7)

If every pulse has changed the value of X only slightly, then approximately we would have

$$R_1 + R_2 + \dots + R_n = \sum_{0}^{n-1} \frac{X_{\nu-1} - X_{\nu}}{g(X_{\nu})} \approx \int_{x_0}^x \frac{\mathrm{d}t}{g(t)}$$
(8)

where $X = X_n$ is the final value of the variable. Since the R_j s are independent, their sum is a Gaussian random variable. In the case g(t) = t (the effect of each pulse is proportional to the value of X at moment t) one finds that $\log(X)$ is normally distributed, i.e. X obeys the log-normal distribution (Equation 6). Using the above mechanism, first proposed by the Dutch astronomer Kapteyn, Kolmogorov [12] has shown that the size of the crushed ores obeys a lognormal distribution.

4. Conclusions

In conclusion, it seems unlikely that, as a rule, the dependence of microhardness on composition obeys Equation 1, which is intuitively the most natural empirical law in solid solutions. We have found that in the system $\text{Tl}_x \text{In}_{1-x}\text{Se}$, H(x) is better described by the log-normal distribution, Equation 6, with a maximum at x ca. 0.5. Results obtained on similar systems, e.g. $(\text{TlGaTe}_2)_x$ - $(\text{TlGaS}_2)_{1-x}$ and $(\text{TlGaSe}_2)_x$ - $(\text{TlInS}_2)_{1-x}$, also show a pronounced maximum at x ca. 0.5 [13] in accordance with Kurnakov's rules. Concerning the member with x = 0.5, i.e. TlInSe₂, measurements of the Knoop hardness, H_K (reported in [14]), gave values for H_K in the range 0.7–0.74 GPa, not very different from those reported in the present work.

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