FT-IR spectroscopy used for the analysis of isotopic effects in H_2O , the rotational behaviour of HBr and the water absorption by DMSO, the identification of different polymers and the determination of ethanol contents of alcohol samples

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Abstract

An IR spectrum using an FT-IR spectroscope from an liquid ethanol sample was taken to become more familiar with the measurement basics. The reproductibility of the measurements were approved to be more applicable for liquid than solid samples. The hygroscopic property of DMSO could be pointed out by measuring the increasing O-H absorption band with increasing ambient air exposition time. On the basis of the IR spectra from three unknown polymer samples their identity could be determined as nitrile, polyethylene terephthalate and polystyrene. The IR spectra of ethanol/water solutions, with known concentrations were used to asses the ethanol volume percentage of three alcohol samples. The determination resulted in $16 \pm 1\%$ for white wine, $36 \pm 1\%$ for rum and $40.8 \pm 2.2\%$ for anise. The vibration wavenumber shift caused through isotopic substitution from H_2O to D_2O could be calculated as 0.728. The equilibrium bond length of HBr in the gaseous phase could be determined as 142 pm with the help of the transition wavenumbers within the measured IR spectrum.

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Introduction

Infrared spectroscopy (IR spectroscopy) has great value in the structure determination of molecules. Most molecules absorb infrared radiation and convert it into molecular oscillations. The absorption of the infrared radiation is characteristic for the different binding ratios in a certain molecule. Using a spectroscope the absorption is measured as a function of the wavenumber $[cm^{-1}]^1$. The resulting IR spectrum is specific to each molecule and depends on the masses involved, the bond distances and the bond strengths. Regrading the *Bouger-Lambert-Beer-law* the intensity of the absorbance is proportional to the thickness of the irradiated substance layer d and the concentration of the absorbing material c.

$$A(\tilde{\nu}) = \epsilon(\tilde{\nu})cd \tag{1}$$

 ϵ is the proportionality constant and dependant on the wavenumber $^2.$

These day it is common to use Fourier-transformation-infrared-spectroscopes (FT-IR spectroscopes) for measuring IR absorption spectra. The advantage is that the samples can be radiated permanently with the complete spectrum of light and therefore a lot of time can be saved ². A schematic scheme of a FT-IR spectroscope can be seen in Fig.1 and the functionality is further explained in the description of the figure. The radiation of the poly chromatic light source reaches the detector after it passes through the interferometer. The detected signals are transferred to the computer where the Fourier transformation is carried out to obtain the IR absorbance spectrum ³.

The FT-IR spectrometers can be equipped with an attenuated total reflection (ATR) module to generate better measurements with aqueous samples 2 . An IR beam with an intensity I_0 is directed onto an optically dense crystal with a high refractive index n_1 at a certain angle. At the interface between the crystal and the sample the light beam gets totally reflected. In the region, where the sample absorbs energy, the evanescent wave will be attenuated. The attenuated beam with intensity I_1 is than directed to the detector, where the attenuated IR beam is recorded as an interferogram 4 .

In one of our experiments the reproductibility of measurements with a FT-IR spectroscope equipped with an ATR module were analyzed to assess the significance of the IR spectra from the other experiments, which were only measured once. Some of the other experiments had the aim to illustrate the hygroscopic property of dimethyl sulfoxide (DMSO), to identify the material of a plastic sample or to determine the ethanol content of an alcohol sample.

Using the model of an harmonic oscillator the stretching vibrations of two atoms, which are chemically bonded, can be approximately described. The behaviour of the two atoms is similar to two masses, which are connected by a spring. The resulting harmonical potential function is a parable and symmetric regrading the stretching and compression of the bond. For the energy of the oscillating atom pair only discrete energy states are allowed, which are given by the vibration quantum numbers ($\nu = 0, 1, 2, 3...$). This leads to the fact that neighboring vibration levels always have the same energy difference. The wavenumber ν_0 for the harmonic oscillation can be described as

$$\nu_0 = \frac{1}{2\pi c} \sqrt{\frac{k}{\mu}},\tag{2}$$

whereas the reduced mass μ

$$\mu = \frac{m_1 \cdot m_2}{m_1 + m_2} \tag{3}$$

can be calculated using the mass of the two atomic nuclei ². One of our experiment involved the verification of the wavenumber shift for the harmonic oscillation upon isotopic substitution.

Molecules can not only vibrate, but also rotate, which is also contributing to the total energy of the molecule. Equal to the changes in the state of vibration the changes in the state of rotation are quantized and therefore linked to discrete energy changes (the rotational level $J=0,\,1,\,2,\,3...$). In an IR spectrum the differences in energy between the vibration and rotation levels can be observed as absorption lines (transition wavenumber). Each of these lines correspond to a transition happening at the wavenumber position of the line. The fundamental band consists of a group of absorption lines with only vibration transitions from the basic vibration state ($\nu=0$) to the first excited vibration state ($\nu=1$). These lines are symmetrical and consist of a R-and a P-branch. The R branch consists of absorption lines with differences in rotational levels of $\Delta J=+1$ and in the P-branch the differences are $\Delta J=-1$. The transition wavenumbers can be calculated as

$$\tilde{\nu} = \tilde{\nu_0} - 2\tilde{\nu_0}x_e + B_e[J'(J'+1) - J''(J''+1)] - \frac{1}{2}\alpha_e[3J'(J'+1) - J''(J''+1)]. \tag{4}$$

J' stands for the rotational level at the basic vibration state and J'' for the rotational level at the excited state. B_e is the rotational constant, α_e a constant for the measure for the effect of vibration excitation on B_e and D_e is the centrifugal distortion constant ². The aim of our last experiment was to determine the equilibrium bond length of an diatomic molecule, which can be calculated with

$$R_e = \sqrt{\frac{\hbar}{8B_e\pi^2c\mu}},\tag{5}$$

where \hbar is the Planck's constant and c the speed of light.

Experimental

Chemicals

For cleaning the measurement instrument and the verification of the reproducibility of the IR measurements ethanol, $M_{\rm C_2H_5OH}=46.07~{\rm g~mol^{-1}}^5$, not pure, provided from the HCI shop was used. For the ethanol/water solutions the ethanol purity was > 99.8% and the water, $M_{\rm H_2O}=18.02~{\rm g~mol^{-1}}^6$, was deionized. Benzoic acid, $M_{\rm C_6H_5COOH}=122.12~{\rm g~mol^{-1}}^7$, > 99% purity was used. The alcohol samples white wine, rum and anise were provided from the assistant. The same applies for the three polymer samples. DMSO, $M_{\rm (CH_3)_2SO}=78.14~{\rm g~mol^{-1}}^8$, < 0.03% H₂O, provided from VWR was used. For isotopic substitution analysis deuertium oxide, $M_{\rm D_2O}=20.03~{\rm g~mol^{-1}}^9$, 99.9% purity, provided from CIL was used.

Sample preparation

Before the measurement benzoic acid was mortared using mortar and pestle. For the ethanol determination 13 solutions containing ethanol and distilled water were prepared. Each solution held a different ethanol concentration. Solutions with ethanol volume percentages of 0%, 9.1%, 16.7%, 18.6%, 37.5%, 44.4%, 47.4%, 50%, 60%, 66.7%, 75%, 80% and 100% were made.

Measurements

All measurements are made with BRUKER Alpha FT-IR spectrometer using the ATR module unless otherwise specified. A simplified version of a Michelson interferometer, which is part of a FT-IR spectroscope, is shown in Fig.1. The sketch for the ATR module is shown in Fig.2. The data for the gaseous HBr sample was measured using the transmission module and provided by the assistant. For all measurements except for the gaseous HBr sample the following settings were used.

type	setting
resolution	$2~\mathrm{cm}^{-1}$
sample scan time	16 scans (measurement time > 35 s)
save data from	$4000 \text{ to } 400 \text{ cm}^{-1}$
result spectrum	absorbance
spectra extended ATR correction	standard

The spectroscope was connected with a laptop, which calculated the interferogram into an absorption spectrum. The used software was OPUS (version 7.5). Before the measurements a background sample was taken, which was deducted from all subsequent measurements. To get in touch with the measurement instrument an IR spectrum of ethanol was taken. For the verification of the reproductibility of IR measurements IR spectra of liquid ethanol and solid benzoic acid were measured each three consecutive times, replacing the samples after each measurement. From three different polymer samples an IR spectrum was taken for the identification of each sample. DMSO was placed at the measurement instrument and every three minutes an IR spectrum was taken, starting with time zero and ending at time 27 min. For the calibration function for the ethanol content determination from the 13 ethanol/water solutions, which were further described in the sample preparation, IR spectra were measured. In addition from three alcohols samples white wine, rum and anise an IR spectrum was taken. Last, an IR spectrum of $\rm H_2O$ and $\rm D_2O$ was measured.

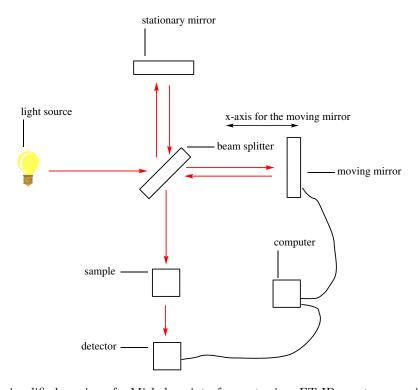


Figure 1: A simplified version of a Michelson interferometer in a FT-IR spectroscope is illustrated. The polychromatic light of the light source is splitted at the beam splitter and reflected at the stationary and the moving mirror. The polychromatic light radiates through the sample and the not absorbed light will be recorded by the detector. For the interferogram, the detector signal is recorded as a function of the position of the moving mirror. For the calculation of the IR spectrum the computers uses some fourier transformation. All our experimental measurements were taken by a FT-IR spectroscope.

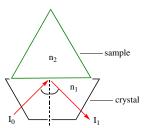


Figure 2: A schematic scheme of the ATR module. The entering light beam I_0 gets totally reflected at the interface of the crystal and the sample on top of it. In the region, where the sample absorbs energy, the evanescent wave will be attenuated. The attenuated beam with intensity I_1 is than directed to the detector, where the attenuated IR beam is recorded as an interferogram.

All measurements were plotted and the results calculated with Python (Version 3.8) as can be seen in the appendix.

Results and Discussion

The measured IR absorption spectrum for ethanol as liquid with its prominent bands is shown in Fig.3. The peaks for the characteristic properties of ethanol from the measured IR spectrum compared with a reference IR spectrum are listed below.

$\begin{array}{c} \text{measured} \\ \text{wavenumber } [\text{cm}^{-1}] \end{array}$	reference wavenumber [cm ⁻¹]	assignment 10
3340	3358 11	O-H stretch (hydroxyl group)
2974	$2974^{\ 11}$	C-H stretch
1049	1050^{-11}	C-O stretch (primary alcohol)

The measured IR spectrum shows the main significant peaks for the main functional group, the O-H and the C-O stretch of the primary alcohol. The absorption bands match those of the reference spectrum. Therefore a certain measuring accuracy of the measurement instrument can be assumed.

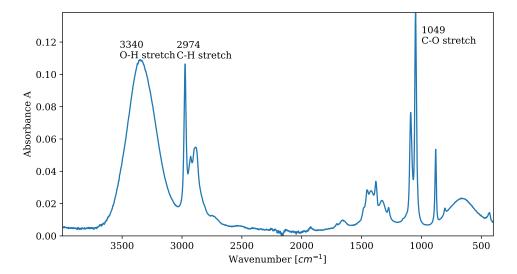


Figure 3: The IR spectrum of ethanol as liquid was measured using the ATR module. The characteristic bands for ethanol are labeled in black.

For the analysis of the reproducibility of FT-IR measurements from three samples of liquid ethanol and from three samples of solid benzoic acid an IR spectrum was taken. As can be seen in Fig.4 the reproductibility of the measurements of the liquid ethanol is better ensured than for the solid benzoic acid. The positions of the individual peaks remain constant for the liquid and the solid samples, but only for the liquid sample the intensity of the absorbance remain nearly consistent as well. The slight differences in the absorbance intensity of the ethanol samples can be neglected, because the cause is probably the fast evaporating of ethanol. The reproductibilty for the measurements with liquid samples can be ensured, but with fast volatile substances, care must be taken that the substances are not exposed to air for too long before the measurement. Before the measurement the solid benzoic acid was mortared leading to different particle sizes, which

make the sample less homogeneous. In addition the layer thickness of the solid sample on the measurement instrument differentiated for each of the three measurements. So in conclusion the layer thickness and the concentration of the solid sample was different for the three measurements, which led with inclusion of the Bouger-Lambert-Beer-law (Eq.1) to different absorbance intensities. In order to ensure the reproductibilty of IR spectra with solid samples, the sample preparation and the sample thickness on the measuring plate has to be strictly equal for each measurement.

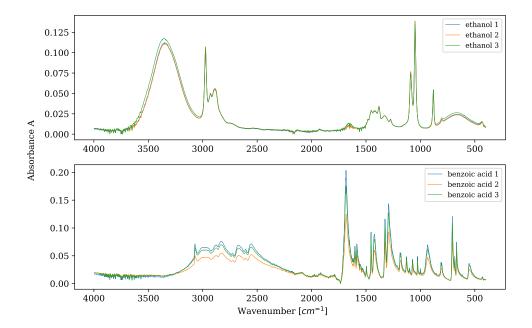


Figure 4: The IR spectrum of ethanol as liquid was measured using the ATR module for three times. The sample was replaced after each measurement. The same applies to the solid benzoic acid. For the liquid sample the measurement reproductibility is better ensured than for the solid.

The hydroxyl group has a characteristic peak between 3670 and 2500 cm⁻¹ ¹⁰. For pure DMSO there is no band visible at this range, because the molecule has no hydroxyl group. As can be seen in Fig.5 the DMSO contained from the beginning some water as the first measurement at time zero shows a peak at the hydroxyl group region. The longer DMSO was exposed to ambient air, the more water it absorbed and the stronger the hydroxyl group peak is visible. This confirms the fact that DMSO is a highly hygroscopic substance ¹². At the wavenumber 3442 cm⁻¹ the absorbance and the exposition time of all ten measurement were plotted against each other, which is seen in Fig.6. The relation between absorption and time of ambient air exposition was fitted using a fourth degree polynomial function with a confident interval of 0.0003, which confirms the accuracy of the fitted model.

$$A(t) = 12(5) \cdot 10^{-8}t^4 - 8(3) \cdot 10^{-6}t^3 + 12(5) \cdot 10^{-5}t^2 + 22(3) \cdot 10^{-4}t + 297(5) \cdot 10^{-4}$$
 (6)

The fitted function can be seen as an orange graph in Fig.6. After 18 minutes of exposition time the function flattens and the ratio is not linear as in the beginning. DMSO forms hydrogen

bonds with water. Over the exposition time all DMSO molecules are saturated with water and no additional hydrogen bonds can be formed. The water absorption of DMSO has a capacity limit, which can be seen with the flattening of the function graph.

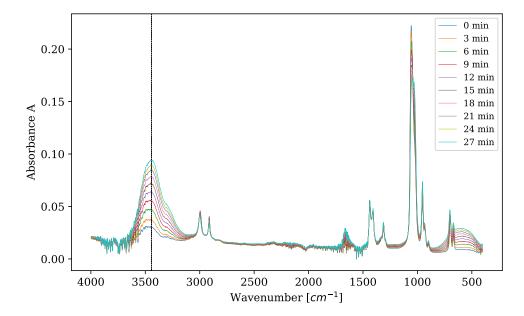


Figure 5: For 27 minutes after every three minutes an IR spectrum of DMSO, which was not replaced between the measurements, was taken. DMSO is highly hygroscopic and tends to absorb water by exposition to ambient air, which can be seen with the increasing intensity of the peak for the O-H stretch at wavenumber $3442~{\rm cm}^{-1}$ marked by the dashed line.

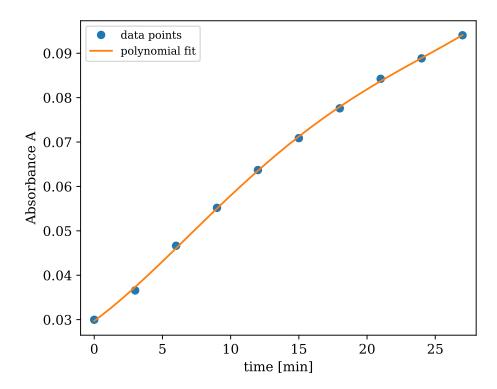


Figure 6: For all ten measurements of DMSO at different times the intensity of the absorbance at wavenumber $3442~{\rm cm^{-1}}$ was plotted against the time since DMSO was exposed to ambient air. The longer DMSO was exposed to ambient air, the more water was absorbed, but the relation is not linear. The beginning has a linear ratio, but towards longer exposition to ambient air, the function flattens.

The three IR spectra from the polymer samples were compared with spectra from the literature and thereby their identity was determined. The results can be seen in Tab.1. Since all samples could be assigned, IR spectroscopy is a suitable method for the identification of poylmer samples.

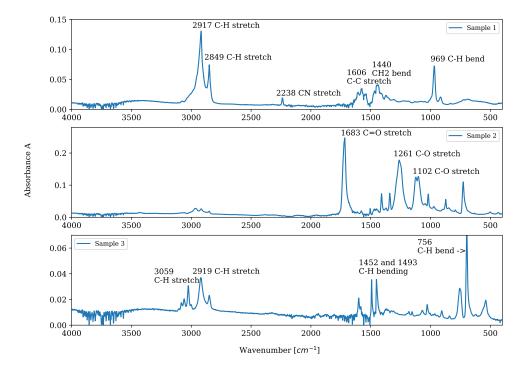


Figure 7: From three different polymer sample an IR spectrum was taken using ATR module. The characteristic bands for each sample are labeled in black and further explained in Tab.1.

sample	identification	$ \begin{array}{c c} \text{measured} & \text{reference} \\ \text{wavenumber } [\text{cm}^{-1}] & \text{wavenumber } [\text{cm}^{-1}] \\ \end{array} $		assignment ¹⁰	
1	nitrile	2917 and 2849	$2917 \text{ and } 2849^{13}$	C-H stretch	
1	nitrile	2238	2237^{-13}	${ m CN~stretch} \ { m (nitrile)}$	
1	nitrile	1606	1605^{13}	C-C stretch	
1	nitrile	1440	1440^{13}	CH_2 bend	
1	nitrile	969	967^{13}	C-H bend	
2	polyethylene terephthalate	1683	1730^{-14}	C=O stretch	
2	polyethylene terephthalate	1261	1285^{-14}	C-O stretch (carboxyl group)	
2	polyethylene terephthalate	1102	1096^{-14}	C-O stretch (ester group)	
3	polystyrene	3059	$3025\ ^{15}$	aromatic C-H stretch	
3	polystyrene	2919	2921^{-15}	C-H stretch	
3	polystyrene	1452 and 1493	1451 and $1493\ ^{15}$	aromatic C-H stretch vibration	
3	polystyrene	756	$749^{\ 15}$	aromatic C-H deformation vibration	

Table 1: The spectral bands from the three polymer samples assigned with their types of oscillation.

In Fig.8 different absorbance spectra from different ethanol/water solutions are presented. It is shown that the absorbance intensity of the O-H peak decreases with increasing ethanol content. The same pattern applies for the fingerprint region between 800 and 600 cm⁻¹. The intensity of the O-H peak decreases with increasing ethanol content, because ethanol has only one O-H bond and therefore less hydrogen bonds and O-H bonds can cause oscillation in the region between 3550 and 2500 cm⁻¹ ¹⁰. Unlike water, ethanol has C-H bonds in its molecule structure, which causes increasing peak intensities at 2975 and 2890 with rising ethanol content. The wavenumbers 3384, 2975 and 692 cm⁻¹ were selected to generate a function to calculate the ethanol content dependant on the absorbance intensity at these specific wavenumbers. Using the calibration points with the known ethanol contents a linear model was fitted using linear regression, which can be seen in Fig.9. The results of the ethanol content determination from three alcohol samples are shown in Tab.2.

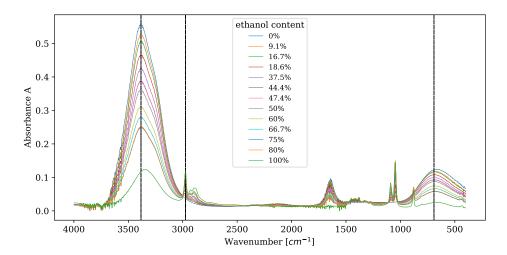


Figure 8: From known solutions with different ethanol and water contents IR spectra were measured. The higher the ethanol content of the solution the less the intensity of the O-H peak at wavenumber 3384 cm⁻¹ is visible. The same pattern applies in the fingerprint region at wavenumber 692 cm⁻¹. In return the C-H peak at wavenumber 2975 cm⁻¹ increases with higher ethanol content. The changes in the band intensities at these three wavenumbers, which are marked with dashed lines, were used to determine the ethanol content of three different alcohol samples.

linear model	alcohol sample	calculated ethanol content [vol%]	reference ethanol content [vol%]	
at O-H peak region	white wine	16.2 ± 1.8	13.5	
at O-H peak region	anise	39.3 ± 1.8	35	
at O-H peak region	rum	36.4 ± 1.8	40	
at C-H peak region	white wine	18.1 ± 2.4	13.5	
at C-H peak region	anise	46.0 ± 2.4	35	
at C-H peak region	rum	36.9 ± 2.4	40	
at fingerprint region	white wine	14 ± 3	13.5	
at fingerprint region	anise	37 ± 3	35	
at fingerprint region	rum	34 ± 3	40	

Table 2: In this table the results from the ethanol content determination of three alcohol samples is shown. The used linear model can be seen in Fig.9.The reference ethanol contents were provided from the assistent Luis Fábregas Ibáñez.

The calculated means and standard deviations for the three alcohol samples are for the white wine $16 \pm 1\%$, for the rum $36 \pm 1\%$ and for the anise $40.8 \pm 2.2\%$. The calculations at the fingerprint region are the most truthful, because except for the rum the other two results fall within the confidence interval. It seems like the samples anise and rum interchanged, because rum should have the higher ethanol content than anise. Perhaps before the anise measurement the ethanol for cleaning the measurement instrument was not completely evaporated and therefore increased the ethanol content of the anise sample. The low percentage of the rum sample could

be explained as the measurement was not taken immediately after the sample was placed on the measurement instrument and some ethanol evaporated before the measurement. For the inaccurate results from all three samples has to be considered that the alcohol samples not only contain water and ethanol. Therefore the other ingredients can falsify the calculations, because the calibration function is based on water/ethanol solutions. In addition certain inaccuracies could be caused through the sample preparation of the water/ethanol samples and the measuring of the IR spectra from these samples. Ethanol is volatile and therefore by the sample preparation due to not fast enough closing of the sample containers some ethanol could have evaporated. After every ethanol/water sample measurement the measuring instrument was cleaned with ethanol. If not all of the ethanol evaporated it could have mixed up with the ethanol/water samples and increased their ethanol content.

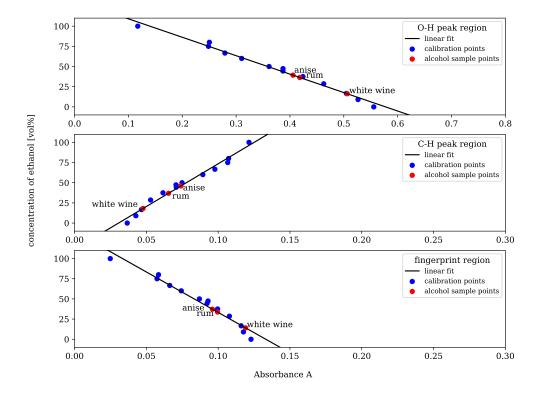


Figure 9: From the IR spectra from the ethanol and water solutions, which can be seen in Fig.8 at three specific wavenumbers, namely 3384, 29475 and $692~{\rm cm}^{-1}$, the absorbance was plotted against the ethanol content. The ratios at all three wavenumbers were linear and therefore using linear regression a calibration function was determined. Using the calibration function the ethanol content of three samples was calculated.

The absorption spectra of H_2O and D_2O are shown in Fig.10. The absorption spectra show the same absorption pattern, but their peaks positions are shifted. The shifts are caused upon isotopic substitution, since deuterium oxide has the higher reduced mass than water. The O-D peak is shifted to the right compared to the O-H peak. The O-H peak is at wavenumber 3389 cm⁻¹ and the O-D peak at wavenumber 2496 cm⁻¹. Therefore the O-H peak is shifted by a value of 0.737. By transforming Eq.3 under the assumption that the spring constant for the O-H and O-D bond is equal the shift of the vibration wavenumber can be calculated as

$$\frac{\tilde{\nu}_{O-D}}{\tilde{\nu}_{O-H}} = \sqrt{\frac{16+2}{16\cdot 2} \cdot \frac{16\cdot 1}{16+1}} = 0.728. \tag{7}$$

Comparing the calculated shift number of 0.728 with the experimentally determined shift number of 0.737 it can be assumed that the measured absorption peaks of the O-H and the O-D band are accurate.

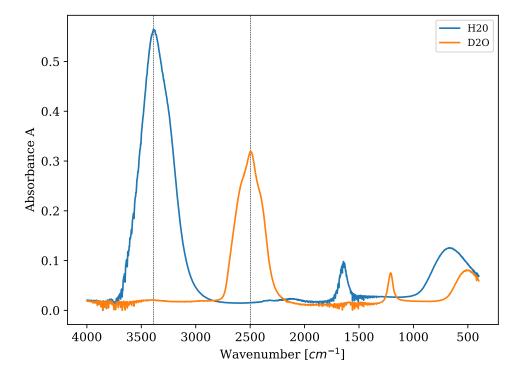


Figure 10: From distilled water and deuterium oxide IR spectra using ATR module were taken. The peak for the O-H stretch is at $3389~\rm cm^{-1}$ and the peak for the O-D stretch at $2496~\rm cm^{-1}$, which is marked by the dashed lines. The peak for the O-D stretch is at a lower wavenumber, because deuterium oxide has the bigger reduced mass than distilled water.

An excerpt of the gaseous absorption spectrum of HBr, which can be seen in Fig.11, shows the fundamental band with its transition wavenumbers of the P- and R-branches. The transition wavenumbers of the P- and R-branches of the fundamental band are linear dependant on the coefficients q1, q2 and q3.

$$\tilde{\nu} = B_e q 1 - \alpha q 2 - D_e q 3 + \tilde{\nu}_0 \tag{8}$$

The coefficients q1, q2 and q3 can be calculated from the vibration and rotation levels of the transition wavenumbers as

$$q1 = J'(J'+1) - J''(J''+1)$$

$$q2 = (1(\nu=1) + \frac{1}{2})J'(J'+1) - \frac{1}{2}J''(J''+1)$$

$$q3 = (J'(J'+1))^2 - (J''(J''+1))^2,$$
(9)

which comes from the Eq.4 and the results can be seen in Tab.3. Since only the fundamental band is analyzed the first two arguments in Eq.4 results in $\tilde{\nu}_0$, which is the intercept of Eq.8. A multiple linear regression was used to determine the rotational constant B_e , which resulted in 8.463 ± 0.016 cm⁻¹. Using Eq.5 the bond length of HBr could be determined as 142 pm. In comparison with the value in literature of 141 pm ¹⁶ the calculations were accurate.

transition state	wavenumber $[cm^{-1}]$	q1	q2	q3
R(0)	2575	2	3	4
R(1)	2591	4	8	32
R(2)	2606	6	15	108
P(1)	2542	-2	-1	-4
P(2)	2525	-4	0	-32
P(3)	2507	-6	3	-108

Table 3: The transition wavenumber of three R- and P-branches and the coefficients calculated from the vibration and rotation quantum numbers of the involved states are shown. The coefficients and the transition wavenumbers were used for the determination of the the rotation constant B_e of the molecule HBr.

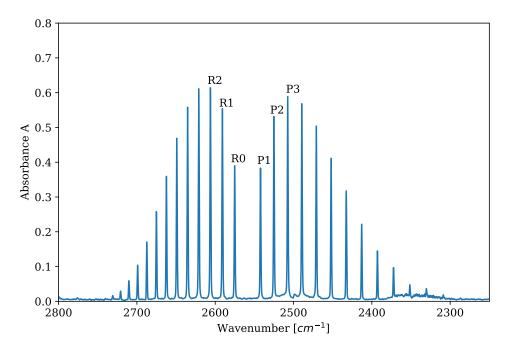


Figure 11: The for the calculation of the bond length relevant cutout of the gaseous HBr absorption spectrum, which was measured with the transmission method, is shown. The spectrum was provided from the assistant Luis Fábregas Ibáñez. The for the calculation used R- and P-branch transition states are labeled in black.

References

- [1] BRUKER, Grundlagen der FT-IR-Spektroskopie, https://www.bruker.com/content/bruker/int/de/products-and-solutions/infrared-and-raman/ft-ir-routine-spectrometer/what-is-ft-ir-spectroscopy.html (visited on 04/19/2021).
- [2] E. Meister, *Grundpraktikum Physikalische Chemie: Theorie und Experimente*, prov. 3.Aufl.,vdf Hochschulverlag AG an der ETH Zuerich, **Version 22. Januar 2021**.
- [3] A. K. Shukla, S. Iravani, *Green synthesis*, characterization and applications of nanoparticles, Elsevier, **2018**, Chapter 12, pp. 303–319.
- [4] T. F. S. US, FTIR Sample Techniques: Attenuated Total Reflection (ATR), https://www.thermofisher.com/ch/en/home/industrial/spectroscopy-elemental-isotope-analysis/spectroscopy-elemental-isotope-analysis-learning-center/molecular-spectroscopy-information/ftir-information/ftir-sample-handling-techniques/ftir-sample-handling-techniques-attenuated-total-reflection-atr.html (visited on 04/19/2021).
- [5] Ethanol 100983, https://www.sigmaaldrich.com/catalog/product/mm/100983?lang=en (visited on 04/13/2021).
- [6] Di water 848333, https://www.sigmaaldrich.com/catalog/product/mm/848333?lang=en®ion=GB (visited on 04/13/2021).
- [7] Benzoic acid, https://www.sigmaaldrich.com/catalog/substance/benzoicacid122126585011? lang=en®ion=GB&gclid=Cj0KCQjw1PSDBhDbARIsAPeTqrdwRK5KWt69ckxLdhg 2H2Tx7BlLJO36JvCCGX0tJXnqLvI58f2PFysaAop1EALw_wcB (visited on 04/19/2021).
- [8] Dimethyl sulphoxide, dehydrated, https://sg.vwr.com/store/product/2381554/dimethyl-sulphoxide-dehydrated-max-0-03-h2o-99-5-analar-normapur-analytical-reagent (visited on 04/19/2021).
- [9] https://shop.isotope.com/supplyimages/MSDS025/DEUTERIUM_OXIDE__D__99.
 9 DLM 4 GHS V.8.2.pdf (visited on 04/19/2021).
- [10] E. Pretsch, P. Buehlmann, M. Badertscher, Structure Determination of Organic Compounds: Table of Spectral Data, Springer, 2009.
- [11] SDBS, SDBS No.: 1300, https://sdbs.db.aist.go.jp/sdbs/cgi-bin/landingpage?sdbsno=1300 (visited on 04/16/2021).
- [12] R. Ellson, R. Stearns, M. Mutz, C. Brown, B. Browning, D. Harris, S. Qureshi, J. Shieh, D. Wold, *Comb. Chem. High Throughput Screening* **2005**, *8*, 489 –498.
- [13] M. Jung, D. Horgen, S. Orski, V. C., K. Beers, G. Balazs, T. Jones, T. Work, K. Brignac, S.-J. Royer, D. Hyrenbach, B. Jensen, J. Lynch, Marine Pollution Bulletin 2017, 127, 704–716.
- [14] A. P. d. S. Pereira, M. H. P. d. Silva, É. P. Lima Júnior, A. d. S. Paula, F. J. Tommasini, Materials Research 2017, 20, 411–420.
- [15] A. Ashraf, FTIR UV-Vis analysis of Polymer(Polystyrene, LDPE) samples, 2014.
- [16] Properties of Ionic Diatomic Molecules, http://hydrogen.physik.uni-wuppertal.de/hyperphysics/hyperphysics/hbase/tables/diatomic.html (visited on 04/19/2021).

Appendix

- A1 Python Scripts
- A2 Lab journal, dated March 30 2021.
- A3 Task sheet
- A4 Regression summaries

A1 - Python Scripts

Python Code for part A

Python Code for part B

```
import numpy as np
from scipy import stats as stats
import matplotlib.pyplot as plt
from scipy.optimize import leastsq
from scipy.signal import find_peaks

font = {'family': 'serif',
    'weight': 'normal',
    'size': 11}

plt.rc('font', **font)
Smallsize = 11

plt.rc('font', size=Smallsize)
plt.rc('axes', labelsize=Smallsize)
plt.rc('legend', fontsize=9)

cm = 1/2.54
# controls default text sizes
# fontsize of the x and y labels
```

```
fig , ax = plt.subplots(2,1,figsize=(25*cm, 16*cm),dpi=600)
fig.text(0.5, 0.04, 'Wavenumber [$cm {-1}$]', ha='center')
fig.text(0.04, 0.5, r'Absorbance A', va='center', rotation='vertical')
filenamesEthanol = ['Bl_Ethanol.dpt','B2_Ethanol.dpt','B3_Ethanol.dpt']
filenamesBenz = ['Bl_Benzoesaeure.dpt','B2_Benzoesaeure.dpt','B3_Benzoesaeure.dpt']
                     dataframesEthanol = []
                     labelse = ['ethanol 1', 'ethanol 2', 'ethanol 3']
labelsb = ['benzoic acid 1', 'benzoic acid 2', 'benzoic acid 3']
                     dataframesBenz= []
                   \begin{array}{l} \text{dataframesBenz} = [] \\ \text{n} = [0\,,1\,,2] \\ \text{for } i,f,k,l,m \text{ in } zip(n,\text{filenamesEthanol},\text{filenamesBenz},\text{labelsb},\text{labelse}): \\ \text{dataframesEthanol}.append(np.loadtxt(f, delimiter=',', skiprows=1, dtype=float)) \\ \text{dataframesBenz}.append(np.loadtxt(k, delimiter=',', skiprows=1, dtype=float)) \\ \text{y} = \text{dataframesEthanol}[i][:,1] \\ \text{x} = \text{dataframesEthanol}[i][:,0] \\ \text{ax}[0].plot(x,y,\text{label} = m,\text{linewidth} = 0.5) \\ \text{ax}[0].legend() \\ \text{y} = \text{dataframesPeng}[i][:,1] \\ \text{y} = \text
 26
 30
 31
 32
33
                                                    ax[0].legend()
y = dataframesBenz[i][:,1]
x = dataframesBenz[i][:,0]
ax[1].plot(x,y,label = l,linewidth = 0.5)
ax[1].legend()
34
 36
 37
 38
                    ax[0].invert_xaxis()
ax[1].invert_xaxis()
39
 40
                     plt.show
                      plt.savefig('TeilB.pdf')
```

Python Code for part C

```
import numpy as np
   import statsmodels.formula.api as smf
   import pandas as pd
from scipy import stats as stats
import matplotlib.pyplot as plt
from scipy.optimize import leastsq
from scipy.signal import find_peaks
   plt.rc(', size', : 1}
plt.rc(', font', **font)
   Smallsize = 11
   plt.rc('font', size=Smallsize)
plt.rc('axes', labelsize=Small
                                                    # controls default text sizes
            'axes', labelsize=Smallsize) # controls default text sizes # fontsize of the x and y labels |
  plt.rc('lege
cm = 1/2.54
  23
       y = dataframes[i][:,1]
x = dataframes[i][:,0]
27
       28
29
30
31
        ax.legend()
        ax.set_ylabel(r'Absorbance A')
ax.set_xlabel('Wavenumber [$cm^{-1}$]')
33
34
  ax.invert_xaxis()
plt.savefig('TeilC1.pdf')
```

```
#Unterschied = 10 verschiedene Absorbanzen bei WellenlÄdnge 3441.65
39
    Unterschied = np.zeros((10,2))
40
    for i in range(0,10):
Unterschied[i] = dataframes[i][545]
fig, ax = plt.subplots(figsize=(15*cm, 12*cm), dpi=600)
fity = Unterschied[:,1]
41
42
43
    \begin{array}{l} {\rm fitx} = {\rm np.array} \, (\, [\, 0\,, 3\,, 6\,, 9\,, 12\,, 15\,, 18\,, 21\,, 24\,, 27\,]) \\ \# \ {\rm Koeffizienten} \ \ {\rm berechnen} \end{array}
46
    # Moerifierten befechnen
coeffs,cov = np.polyfit(fitx, fity, 4, cov = True)
# macht Polynom fÄijr Plot
47
    poly = np.poly1d(coeffs)
    new_x = np. linspace(fitx[0], fitx[-1])
    # gemessenen X-Werte in Polynomfunktion eingesetzt
y_val = poly(fitx)
   y_val = poly(fitx)
plt.xlabel(r'time [min]')
plt.ylabel(r'Absorbance A')
plt.plot(fitx, fity, 'o', label = 'data points')
plt.plot(new_x, new_y, label = 'polynomial fit')
plt.xlim([fitx[0]-1, fitx[-1] + 1 ])
plt.rc('font', size=Smallsize)  # control
plt.rc('axes', labelsize=Smallsize)  # fontsiz
plt.rc('legend', fontsize=9)
plt.legend()
                                                                        # controls default text sizes
# fontsize of the x and y labels
60
61
62
    plt.savefig('TeilC2.pdf')
    plt.show
65
66
    Vertrauensintervall Berechnung
67
68
    # Standardfehler Mittelwert
69
70
    n = np.std(fity-y_val)/np.sqrt(10)
    #t-taktil
    ts = stats.t(df=9).ppf(0.975)
    Verintervall = n*ts
    Standardfehler fÄijr die einzelnen Coeffs
    Standardfehlercoeffs = np.sqrt(np.diag(cov))
                                                                                     teigung:\n', Standardfehlercoeffs)
    print (
```

Python Code part D

```
24 CD = ['D3_CD.dpt']
25 dataframesHandschuh = []
26 dataframesPET = []
               dataframesCD = []
               \begin{array}{l} n = [0] \\ \text{for } i, f, k, m \text{ in } zip(n, Handschuh, PET, CD): \end{array} 
                               dataframesHandschuh.append(np.loadtxt(f, delimiter=',', skiprows=1, dtype=float))
dataframesPET.append(np.loadtxt(k, delimiter=',', skiprows=1, dtype=float))
dataframesCD.append(np.loadtxt(m, delimiter=',', skiprows=1, dtype=float))
y = dataframesHandschuh[i][:,1]
x = dataframesHandschuh[i][:,0]
peaks, properties = find_peaks(y,height=None)
Peakpunkte = dataframesHandschuh[i][peaks]
ax[0].axis([400, 4000, 0, 0.15])
ax[0].plot(x,y,label = 'Sample 1')
ax[0].text(2990,0.13600, '2917 C-H stretch')
ax[0].text(2890,0.08285, '2849 C-H stretch')
ax[0].text(2290,0.02300, '2238 CN stretch')
ax[0].text(1700,0.04280, '1606 \nC-C stretch')
ax[0].text(1490,0.05600, '1440 \nCH2 bend')
ax[0].text(1000,0.08000, '969 C-H bend')
ax[0].legend()
                                   dataframesHandschuh.append(np.loadtxt(f, delimiter=',', skiprows=1, dtype=float))
32
33
34
 35
 36
 37
 38
39
 40
 41
 42
 43
                               ax [0]. text (1490, 0.05600, "1440 \nCH2 bend")
ax [0]. text (1000, 0.08000, "969 C-H bend")
ax [0]. legend ()
y = dataframesPET [i][:,1]
x = dataframesPET [i][:,0]
#peaks, properties = find_peaks(y, height=0.02)
#Peakpunkte = dataframesPET [i][peaks]
ax [1]. axis ([400, 4000, 0, 0.28])
ax [1]. plot (x,y, label = 'Sample 2')
ax [1]. text (1750, 0.255, "1683 C=O stretch")
ax [1]. text (1150, 0.135, "1102 C-O stretch")
ax [1]. text (1150, 0.135, "1102 C-O stretch")
ax [1]. legend ()
y = dataframesCD [i][:,1]
x = dataframesCD [i][:,0]
#peaks, properties = find_peaks(y, height=0.02)
#Peakpunkte = dataframesPET [i][peaks]
ax [2]. axis ([400, 4000, 0, 0.07])
ax [2]. plot (x,y, label = 'Sample 3')
ax [2]. text (3310, 0.033, "3059 \nC-H stretch")
ax [2]. text (2990, 0.04, "2919 C-H stretch")
ax [2]. text (1110, 0.056, "756 \nC-H bending")
ax [2]. legend ()

Ol invert xaxis()
 44
 45
 46
 47
 49
 50
51
 55
 56
58
59
60
61
62
63
65
66
               ax[0].invert_xaxis()
              ax[1].invert_xaxis()
ax[2].invert_xaxis()
plt.savefig('TeilD.pdf')
 69
              plt.show
```

Python Code part E

```
import numpy as np
from scipy import stats as stats
import matplotlib.pyplot as plt
from scipy.optimize import leastsq
from scipy.signal import find_peaks

"""

GrÃússeneinstellungen / Schrifteinstellungen

"""

font = {'family': 'serif',
    'weight': 'normal',
    'size': 11}

plt.rc('font', **font)
Smallsize = 11
cm = 1/2.54
"""
```

```
Daten fÄijr die Kalibration mit den bekannten EthanolgehÄdltern
      {\rm fig} \ , \ {\rm ax} \ = \ {\rm plt.subplots} \, (\, {\rm figsize} \, = \, (25 \, * {\rm cm} \, , \ 12 \, * {\rm cm}) \, , {\rm dpi} \, = \, 600)
    20
23
24
              y = dataframes[i][:,1]

x = dataframes[i][:,0]
              plt.rc('font', size=Smallsize)  # controls default text sizes
plt.rc('axes', labelsize=Smallsize)  # fontsize of the x and y labels
plt.rc('legend', fontsize=9)
ax.plot(x,y, label = 1, linewidth = 0.5)
x1 = dataframes[i][602,0]
x2 = dataframes[i][1003,0]
x3 = dataframes[i][3241,0]
x3 = dataframes[i][3241,0]
x4 = xyline(x1 = color=16; linestyle=16; linewidth = 0.4)
28
20
30
31
32
33
34
              x3 = dataframes [1][3241,0]

ax.axvline(x1, color='k', linestyle='—

ax.axvline(x2, color='k', linestyle='—

ax.axvline(x3, color='k', linestyle='—

plt.legend(title="ethanol content")

ax.set_ylabel(r'Absorbance A')

ax.set_xlabel('Wavenumber [$cm^{-1}$]')
                                                                                                              , linewidth = 0.4)
35
                                                                                                               , linewidth = 0.4)
36
37
                                                                                                               .linewidth = 0.4)
38
40
     ax.invert_xaxis()
plt.savefig('TeilE1.pdf')
41
42
43
44
     Daten der zu bestimmenden Substanzen
     \begin{array}{ll} fig\;,\;\;ax=plt.subplots(\,figsize\,{=}\,(12,\;4)\,)\\ filenames2=\left[\;'E\_Rum.\,dpt'\;,\,'E\_Weisswein.\,dpt\;'\;,\,'E\_Anise.\,dpt\;'\;\right]\\ dataframes2=\left[\;\right] \end{array}
47
48
49
     m \, = \, \left[\, 0 \,\, , 1 \,\, , 2 \,\right]
50
     for i, f in zip (m, filenames 2):
               dataframes2.append(np.loadtxt(f, delimiter=',', skiprows=1, dtype=float))
              y = dataframes2[i][:,1]

x = dataframes2[i][:,0]
              ax.plot(x,y)
ax.set_ylabel(r'Absorbance')
ax.set_xlabel('Wavenumber [$cm^{-1}$]')
56
57
      ax.invert_xaxis()
      plt.show
      plt.figure()
60
61
      Modelfunktionberechnung:
62
            -bei WellenlÄdnge 3383.52, bei Index 602 fÄijr OH stretch

-bei WellenlÄdnge 2974.6, bei Index 1003 fÄijr CH stretch

-bei WellenlÄdnge 692.409, bei Index 3241 in Fingerprint Region
63
64
66
      Unterschied = np.zeros((13,2))
67
     Unterschied = np.zeros((13,2))
Unterschied1 = np.zeros((13,2))
Unterschied2 = np.zeros((13,2))
for i in range(0,13):
        Unterschied[i] = dataframes[i][602]
        Unterschied1[i] = dataframes[i][1003]
        Unterschied2[i] = dataframes[i][3241]
Samples = np.zeros((3,2))
Samples1 = np.zeros((3,2))
Samples2 = np.zeros((3,2))
for i in range(0,3):
68
69
75
      for i in range (0,3):
               Samples[i] = dataframes2[i][602]
Samples1[i] = dataframes2[i][1003]
Samples2[i] = dataframes2[i][3241]
80
     fig , ax = plt.subplots(3,1,figsize=(26*cm, 20*cm),dpi=600)
plt.rc('font', size=Smallsize)  # controls default text sizes
plt.rc('axes', labelsize=Smallsize)  # fontsize of the x and y labels
plt.rc('legend', fontsize=9)
82
86 fig.text(0.5, 0.04, r'Absorbance A', ha='center')
```

```
87 | fig.text(0.04, 0.5, 'concentration of ethanol [vol%]', va='center', rotation='vertical'
              # X Achse die verschiedenen Absorbanzen, Y Achse gibt den Ethanolgehalt
              fitx = Unterschied[:,1]
fitx1 = Unterschied1[:,1]
               fitx2 = Unterschied2 [:,1]
              fitx2 = Onterschied2[.,1]
fitxSample = Samples[:,1]
fitxSample1 = Samples1[:,1]
fitxSample2 = Samples2[:,1]
fity = np.array([0,100/110*10,100/120*20,100/140*40,100/160*60,100/180*80,100/190*
90,50,100/250*150,100/300*200,100/400*300,100/500*400,100])
              # Line fitting
               model, cov = np. polyfit (fitx, fity, 1, cov = True)
            model1, covv = np. polyfit (fitx1, fity,1,cov = True)
model2, covvv = np. polyfit (fitx1, fity,1, cov = True)
predict = np. poly1d (model)
predict1 = np. poly1d (model1)
predict2 = np. poly1d (model2)
 100
              x_{lin}reg = np.linspace(0, 1,1000)
103 x_lin_reg = np.linspace(0, 1,100)
104 y_lin_reg = predict(x_lin_reg)
105 y_lin_reg1 = predict1(x_lin_reg)
106 y_lin_reg2 = predict2(x_lin_reg)
107 y_val = predict(fitx)
108 y_val1 = predict1(fitx1)
109 y_val2 = predict2(fitx2)
              y_Samples = predict(fitxSample)
            y_Samples1 = predict(fitxSample1)
y_Samples1 = predict2(fitxSample1)
y_Samples2 = predict2(fitxSample2)
ax[0]. axis([0, 0.8, -10, 110])
ax[0]. scatter(fitx, fity, c = 'b', label = 'calibration points')
ax[0]. scatter(fitxSample, y_Samples, c = 'r', label = 'alcohol sa
ax[0]. plot(x_lin_reg, y_lin_reg, c = 'k', label = 'linear fit')
ax[0]. apportant(*white wines* (0.51, 17))
 113
                                                                                                                                                                                                                                                                                                           sample points')
              ax [0]. annotate ("white wine", (0.51, 17))
ax [0]. annotate ("rum", (0.43,36))
ax [0]. annotate ("anise", (0.408, 42.3))
ax [0]. legend (title="O-H peak region")
 118
 119
            ax [0]. almotate (almise ', (0.408, 42.5))
ax [1]. axis([0, 0.3, -10, 110])
ax [1]. scatter(fitx1, fity, c = 'b', label = 'calibration points')
ax [1]. scatter(fitxSample1, y_Samples1, c = 'r', label = 'alcohol s
ax [1]. plot(x_lin_reg, y_lin_reg1, c = 'k', label = 'linear fit')
ax [1]. legend(title='C-H peak region')
ax [1]. annotate("white wine", (0.012, 20.1))
ax [1]. annotate("rum", (0.068, 30.2))
ax [1]. annotate("anise", (0.0755, 40.5))
ax [2]. axis([0, 0.3, -10, 110])
ax [2]. scatter(fitx2, fity, c = 'b', label = 'calibration points')
ax [2]. scatter(fitx2, fity, c = 'b', label = 'linear fit')
ax [2]. scatter(fitx2, fity, c = 'b', label = 'linear fit')
ax [2]. annotate("white wine", (0.12, 15))
ax [2]. annotate("white wine", (0.12, 15))
ax [2]. annotate("anise", (0.075, 35.7))
ax [2]. legend(title="fingerprint region")
plt. savefig('TeilE2.pdf')
 120
                                                                                                                                                                                                                                                                                                                   sample points')
 124
 126
 130
                                                                                                                                                                                                                                                                                                                  sample points')
 136
               plt.savefig('TeilE2.pdf')
              plt.show
 138
 139
 140
               Vertrauensintervall Berechnung
 141
 142
             # Standardfehler Mittelwert
 143
# January | # Janu
               #t-taktil
                 ts = stats.t(df=12).ppf(0.975)
 149
               Verintervall = n*ts
Verintervall1 = n1*ts
 150
               Verintervall2 = n2*ts
               Standardfehler des Koeffiezienten und der Steigung
               error1 = np.sqrt(np.diag(cov))
 157 error2 = np.sqrt(np.diag(covv))
```

```
error3 = np.sqrt(np.diag(covvv))
                print('Stdfehler fAjjr Funktion bei O-H peak:\nKoeffizient:',error1[0],'\n','Steigung:',
 160
                                   error1 [1])
                                                                                  er fÄijr Funktion bei C-H peak:\nKoeffizient:',error2[0],'\n','Steigung:',
                print ('Std
 161
                                 error2[1])
                print('Stdf
                                                                                               fAijr Funktion bei Fingerprint:\nKoeffizient:',error3[0],'\n','Steigung
 162
                                           , error3 [1])
 164
                Alkoholgehalt der Samples bestimmen
 165
                Referenzrum = np.array([40,40,40])
 167
                Referenzweisswein = np. array ([13.5,13.5,13.5])
            Referenzweisswein = np. array ([13.5,13.5,13.5])
Referenzanise = np. array ([35,35,35])
Rum = np. array ([y_Samples [0],y_Samples1 [0],y_Samples2 [0]])
mRum = np. mean(Rum)
Weisswein = np. array ([y_Samples [1],y_Samples1 [1],y_Samples2 [1]])
mWeisswein = np. mean (Weisswein)
Anise = np. array ([y_Samples [2],y_Samples1 [2],y_Samples2 [2]])
mAnise = np. mean (Anise)
#Stead and following the following the second content of the second con
#Standardfehler Mittelwert
stdrum = np.std (Rum-Referenzrum)/np.sqrt (3)
stdweisswein = np.std (Weisswein-Referenzweisswein)/np.sqrt (3)
stdanise = np.std (Anise-Referenzanise)/np.sqrt (3)
```

Python Code for part F

```
import numpy as np
    from scipy import stats as stats
    import matplotlib.pyplot as plt
    from scipy.optimize import leastsq
    from scipy.signal import find_peaks
    import data
    H2O = 'F\_H20.dpt
    D2O = F_D2O \cdot dpt
    dataH = np.loadtxt(H2O, delimiter=',', skiprows=1, dtype=float)
dataD = np.loadtxt(D2O, delimiter=',', skiprows=1, dtype=float)
    Schrift und GrÄűsseneinstellungen fÄijr Plots
    font = {'family' : 'serif';
    'weight' : 'normal';
    'size' : 11}
1.8
    plt.rc('font',
19
    Pit.rc('font', **font)
Smallsize = 11
plt.rc('f
20
21
    plt.rc('font , size=Smallsize)
plt.rc('axes', labelsize=Smallsize)
plt.rc('legend', fontsize=9)
                                                                     # controls default text sizes
                                                               # fontsize of the x and y labels
   plt.rc('lege
cm = 1/2.54
24
25
26
    Plot H2O und D2O Spektrum
27
28
    \label{eq:fig_size} \mbox{fig , ax = plt.subplots(figsize=(18*cm, 13*cm), dpi=600)}
   x = dataH[:,0]
y = dataH[:,1]
x1 = dataD[:,0]
y1 = dataD[:,1]
30
31
32
    x2 = dataH[597,0]
    x3 = dataH[1472, 0]
   peaks, properties = find_peaks(y1, height=0.3)
#peaks, properties = find_peaks(y, height=0.5)
PeakpunkteD = dataD[peaks]
#PeakpunkteH = dataH[peaks]
39
#PeakpunkteH = dataH[peaks]

plt.ylabel(r'Absorbance A')

plt.ylabel('Wavenumber [$cm^{-1}$]')

2 ax.axvline(x2, color='k', linestyle='-', linewidth = 0.4)

3 ax.axvline(x3, color='k', linestyle='-', linewidth = 0.4)

4 plt.plot(x,y, label = 'H20')

plt.plot(x1,y1, label = 'D2O')

ax.invert_xaxis()
    plt.legend()
    plt.savefig ('TeilF.pdf')
49
    plt.show()
50
51
   OH Peak bei H2O bei Wellenzahl 3389, Index 597
OH Peak bei D2O bei Wellenzahl 2496, Index 1472
    Berechnungen fÄijr die reduzierte Masse
57
    # Massen H, O und D in u
58
    u = 2.65686*10**-23 \# in gramm
59
   mH = 1*u
   mO = 16*u
   mD = 2*11
62
   c = 299792458*10**3 \# Lichtgeschwindigkeit cm/s
63
   muH = mH*mO/(mH+mO) \# in Einheit u
65
   muD = mD*mO/(mD+mO) # in Einheit u
66
    Berechnung Federkonstante —> Annahme sind gleich
69
70
_{71} kH = ((3389*2*c)**2)*muH
```

Python Code for part G

```
import numpy as np
     from sklearn import linear_model
     import statsmodels.api as sm
    from scipy import stats as stats
import matplotlib.pyplot as plt
from scipy.optimize import leastsq
from scipy.signal import find_peaks
10
     Schrift und GrÃűsseneinstellungen fÃijr Plots
     14
                                       : 11}
15
     Smallsize = 11
plt.rc/'
    plt.rc('font', size=Smallsize)
plt.rc('axes', labelsize=Smallsize)
plt.rc('legend', fontsize=9)
cm = 1/2.54
                                                                                # controls default text sizes
# fontsize of the x and y labels
20
    cm = 1/2.54
21
     import data and plot for spectrum
24
25
26
     \label{eq:fig_size} \texttt{fig} \;,\;\; \texttt{ax} \;=\; \texttt{plt.subplots} \left(\; \texttt{figsize} = (20 * \texttt{cm} \,,\;\; 13 * \texttt{cm} \,) \;, \texttt{dpi} = 600 \right)
27
     data = np.loadtxt(HBr, delimiter=',', skiprows=1, dtype=float)
    x = data[:,0]y = data[:,1]
30
     {\tt peaks}\;,\;\;{\tt properties}\;=\;{\tt find\_peaks}\,({\tt y}\,,{\tt height}\,{=}\,0.3)
    peaks, properties = find_peaks(y, heigheats)
Peakpunkte = data [peaks]
plt.axis([2250, 2800, 0, 0.8])
plt.annotate('R0', (2580, 0.4))
plt.annotate('R1', (2595, 0.56))
plt.annotate('R2', (2610, 0.625))
plt.annotate('P1', (2546.5, 0.395))
plt.annotate('P1', (2530, 0.54))
plt.annotate('P2', (2530, 0.54))
plt.annotate('P3', (2509.7, 0.6))
plt.ylabel('Absorbance A')
plt.xlabel('Wavenumber [$cm^{-1}$]')
plt.plot(x,y)
33
36
37
38
39
40
     plt.plot(x,y)
     ax.invert_xaxis()
     plt.show
44
    plt.savefig('TeilG.pdf')
45
48
     Berechnung reduzierte Masse
49
    u = 1.66054019999999*10**-24
50
    mH = 1*u
51
    mBr = 80*u
52
    muHBr = 1/((1/mH)+(1/mBr)) \# Einheit gramm
55
56
     bei Wellenl\mbox{\tt Adnge} 2800 bis 2400 / Zuordnung der Wellenl\mbox{\tt Adngen}
```

```
vR2 = 2606.18
 61
     vP1 = 2542.19
 62

    \begin{array}{r}
      \text{vP1} = 2542.13 \\
      \text{vP2} = 2524.99 \\
      \text{vP3} = 2507.40
    \end{array}

 63
     v = np. array ([vR0, vR1, vR2, vP1, vP2, vP3])

q0 = np. array ([0,0,0,0,0,0])

q1 = np. array ([1,1,1,1,1,1])

q2 = np. array ([2,2,2,2,2,2])

q3 = np. array ([2,4,6,-2,-4,-6])

q4 = np. array ([3,8,15,-1,0,3])

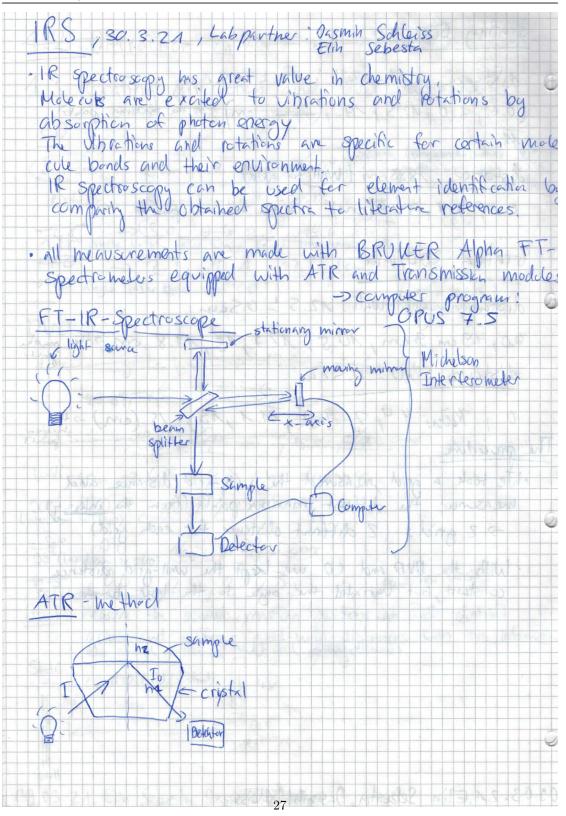
q5 = np. array ([4,32,108,-4,-32,-108])
 73
      multiple Lineare Regression (wurde dann nicht verwendet fÄijr Bericht, da Fehler nicht gut ermittelbar sind)
 74
 75
 76
     X = np.array([q3,q4,q5]).transpose()
     y = v
 79
     \begin{array}{ll} lm = linear\_model.\,LinearRegression\,()\\ lm.\,fit\,(X,\ y)\\ coeffs = lm.\,coef\_ \end{array}
 80
 81
      intercept = lm.intercept_
      y2 = lm.predict(X)
      print(coeffs)
 86
 87
      multiple Linear Regression with OLS
 88
 89
     X = sm.add\_constant(X)
      model = sm.OLS(y, X).fit()
ypred = model.predict(X)
 91
 99
      coef = model.params
 93
      print(model.summary())
 94
 95
 97
      Berechnung Fehler
 98
     Differenz = y-ypred

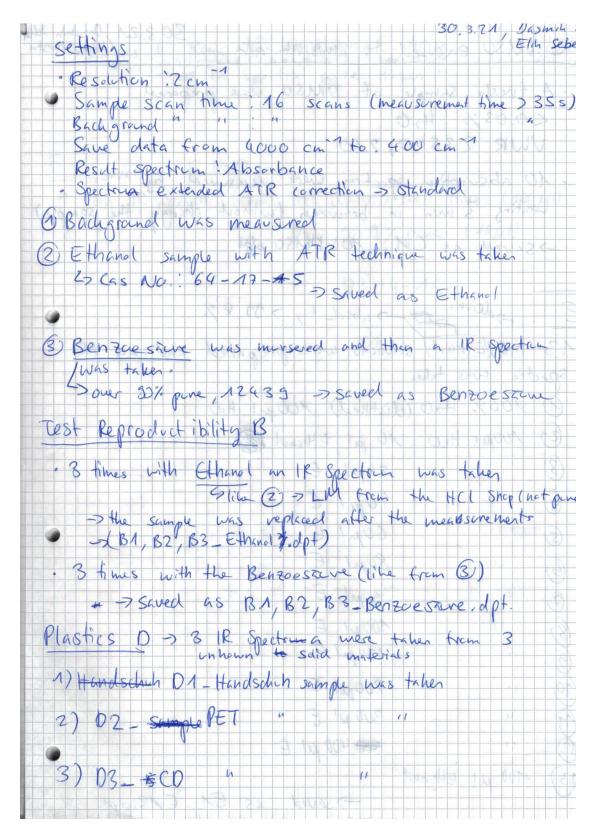
sterr = np.std(Differenz)/np.sqrt(6)

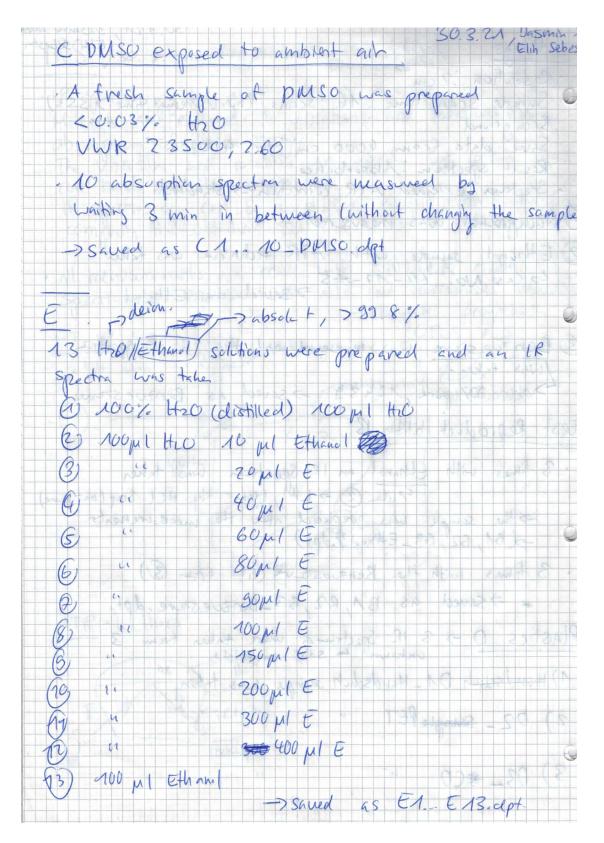
ts = stats.t(df=5).ppf(0.975)

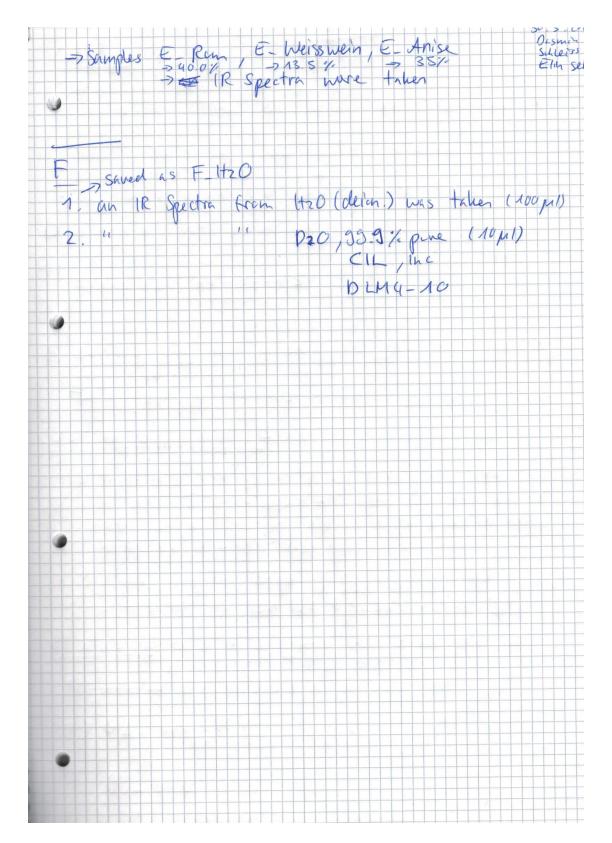
Verintervall = sterr*ts
 99
100
102
104
      Berechnung\ Bindungsl\tilde{A}dnge\ im\ Gleichgewichtszustand
105
106
c = 299792458*10**2\# Lichtgeschwindigkeit cm/s h = 6.62607015*10**-27 # Plankkonstante in cm^2*g/s^3 B = coef[1] # Einheit cm^-1
110 R = \text{np.sqrt}(h/(8*B*np.pi**2*c*muHBr)) #Einheit cm
```

A2 - Lab journal









A3 - Task sheet

IRS Infrared Spectroscopy Experiments

Location: HCI H290 Group: 25

How to prepare to the introduction

Read the document [1] as a good preparation to the IR experiment. Additional recommended references are listed at the end of this task list.

Experiments

Note: Spectrometer settings and sample details have to be recorded clearly. Each team works strictly independently to collect its own data (preparing solutions, measuring spectra) unless otherwise specified. Measurements are made with Bruker Alpha FT-IR spectrometers, equipped with ATR and Transmission modules, respectively.

A — Measurement basics

With the ATR module measure the IR absorption spectrum of pure liquid or solid substance. Background and sample interferograms are measured and, with the spectrometer software, processed to obtain the single intensity spectra and finally the transmission and the absorption spectrum of the sample. With this example, explain and illustrate the route from the interferograms to the absorption spectrum. Compare the absorption spectrum with one taken from literature [2]. Assign prominent bands in the spectrum to the vibration of certain functional groups.

B — Verification of measurement reproducibility

Check, with liquid and solid samples, whether the ATR technique gives reproducible spectra (with respect to absorbance values and wavenumber positions of individual peaks). Take 3 measurements of each substance, replacing the sample between the measurements.

C — Absorption spectrum of liquid DMSO exposed to ambient air

DMSO is a highly hygroscopic substance and tends to absorb water by simple exposition to ambient air. IR allows us to monitor this phenomenon. Prepare a sample of fresh DMSO and measure 10 absorption spectra by waiting (e.g. 3 minutes) between measurements (without changing the sample). Plot all spectra, identify the water wavenumbers, and plot the absorbance values A against time t. Find

Plot all spectra, identify the water wavenumbers, and plot the absorbance values A against time t. Find an appropriate model function $A(\phi_t)$ that fits the data.

D — Absorption spectra and identification of plastics

ATR-IR spectroscopy has a lot of industrial applications, one of them being quality control. Databases exist to identify e.g. polymers [3,4].

Measure absorption spectra of selected polymer samples. Draw the spectra in an overview for easy identification or differentiation. Assign prominent bands in the spectra to the vibration of certain functional groups.

E — Absorption spectra of binary solutions of water and ethanol

IR spectra of binary solutions of water (w) and ethanol (e) strongly depend on the composition. Prepare 12 binary solutions (including pure substances) with known volume fractions $\phi_{\rm e} = V_{\rm e}/(V_{\rm e} + V_{\rm w})$ and measure the absorption spectra of these samples with the ATR technique. Plot all spectra (full range and details) and discuss the spectral changes.

At a selected wavenumber determine absorbance values A and plot them against ϕ_e . Find an appropriate model function $A(\phi_e)$ that fits the data. Repeat the procedure with absorbance data taken at two other wavenumbers.

Measure the absorption spectrum of a sample of a liquor (e.g. wine, grappa, whisky, etc. feel free to bring a sample from home) and determine the alcohol content using the above calibration graphs.

F — Vibrational spectra and transition energies of isotopomers

Measure the IR absorption spectrum of a compound (e.g. H_2O or $CHCl_3$) and of its deuterated analogon. Verify (by calculation of the reduced masses μ) the shift of the vibration wavenumber upon isotopic substitution [1].

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G — Rotational-vibrational spectra of diatomic molecules

At high resolution, IR absorption spectra of small molecules may be observed with discrete rotationalvibrational transitions, see [1]. Aim of this experiment is to prepare several gaseous samples, to measure well-resolved spectra in a gas cell and to study and analyze the spectra of diatomic molecules to obtain specific molecular parameters such as bond lengths with high precision.

The discrete vibrational-rotational energy levels of a diatomic molecule can be described by

$$\begin{split} E_{\rm vib,rot} = h\nu_0 \left(v + \tfrac{1}{2}\right) - h\nu_0 x_{\rm e} \left(v + \tfrac{1}{2}\right)^2 + hc \left(B_{\rm e} - \alpha(v + \tfrac{1}{2})\right) J(J+1) \quad , \\ B_{\rm e} = \frac{h}{8\pi^2 \mu c R_{\rm e}^2} \quad , \end{split}$$

v and J: vibrational and rotational quantum number, respectively, ν_0 : frequency of the harmonic vibration, $x_{\rm e}$: anharmonicity constant, $B_{\rm e}$: equilibrium rotational constant, α : rotation-vibration constant, $R_{\rm e}$: equilibrium bond length.

Rotation-vibration absorption transitions $(v=0,J'') \rightarrow (v=1,J')$ within the fundamental vibration transition occur at discrete transition wavenumbers

$$\tilde{\nu} = \frac{E_{v=1,J'} - E_{v=0,J''}}{hc} = \tilde{\nu}_0 - 2\tilde{\nu}_0 x_e + \left(B_e - \frac{3}{2}\alpha\right) J'(J'+1) - \left(B_e - \frac{1}{2}\alpha\right) J''(J''+1) \quad .$$

A simple method uses two differences of out of four transitions, e.g.

transition	ΔJ	change of J	transition	ΔJ	change of J
R(0) transition	+1	$J'' = 0 \to J' = 1$	P(1) transition	-1	$J'' = 1 \rightarrow J' = 0$
R(1) transition	+1	$J''=1 \rightarrow J'=2$	P(1) transition	-1	$J''=2 \rightarrow J'=1$

Explicit expressions of $\tilde{\nu}$ for each of the four transitions are given in [1]. In the experiment you will determine the wavenumbers of these transitions from the absorption spectrum and further calculate the equilibrium bond length $R_{\rm e}$ of the molecule.

Literature .

- [1] E. Meister, Infrarotspektroskopie (provisorische Versuchsanleitung), ETH Z $\tilde{\mathbf{A}}_{4}^{1}$ rich (5.4.2020).
- [2] J. Coates, Interpretation of Infrared Spectra, A Practical Approach, in: R.A. Meyers, Encyclopedia of Analytical Chemistry, John Wiley & Sons, 2006.
- [3] SDBS Spectral Database for Organic Compounds, National Institute of Advanced Industrial Science and Technology (AIST), Japan. sdbs.db.aist.go.jp/sdbs (9.3.2020).
- [4] Polymer Sample Identification Using Quest ATR, Specac Application Note 42 www.specac.com/en/documents/application-notes/polymer-identification-quest-application-note (9.3.2020).

Lab report

Consult the appendices B and C in the praktikum book and the document *Lab Report Correction Checklist* when preparing your report! The report should present quantitative as well as qualitative findings of your experiments, clearly described and with context to theory and literature references. Just consider your report a research paper!

Deadline to hand in the signed report (per email to luis.fabregas@phys.chem.ethz.ch and C.C. to erich.meister@phys.chem.ethz.ch) is April 23 2021.

Please add to your report: - This task list

- copies of your lab journal
- program codes used
- all regression summaries

Please attach a ZIP file with all measured data, R files and report PDF to the email.

A4 - Regression summaries

for part C

The standard errors for the fourth degree polynomial function:

```
In [22]: runfile('C:/Users/sebes/OneDrive/ETH/4.Semester_2020_ETH/PC Praktikum/
2.Versuch_IRS/TeilC/C_DMSO.py', wdir='C:/Users/sebes/OneDrive/ETH/4.Semester_2020_ETH/
PC Praktikum/2.Versuch_IRS/TeilC')
Standardfehler t^4, t^3, t^2, t und Steigung:
[5.33479060e-08 2.90440032e-06 5.08218316e-05 3.10475243e-04
5.36884708e-04]
```

for part E

The standard errors for the three linear regressions:

```
In [23]: runfile('C:/Users/sebes/OneDrive/ETH/4.Semester_2020_ETH/PC Praktikum/2.Versuch_IRS/TeilE/E_Ethanolcontent.py', wdir='C:/Users/sebes/OneDrive/ETH/4.Semester_2020_ETH/PC Praktikum/2.Versuch_IRS/TeilE')
Stdfehler für Funktion bei 0-H peak:
Koeffizient: 7.390012199359168
Steigung: 2.882195820574911
Stdfehler für Funktion bei C-H peak:
Koeffizient: 46.17353509479053
Steigung: 3.682984870367805
Stdfehler für Funktion bei Fingerprint:
Koeffizient: 56.865354683960426
Steigung: 5.132956112610616
```

for part G

The regression summary (OLS function used):

	J (-					
=======	=========	=======			========	=======
	coef	std err	t	P> t	[0.025	0.975]
const	2558.8621	0.019	1.38e+05	0.000	2558.782	2558.942
x1	8.4628	0.007	1131.278	0.000	8.431	8.495
x2	-0.2305	0.003	-70.847	0.000	-0.244	-0.216
x3	0	0.000	0	1.000	-0.002	0.002
=======		======		=======	======	======