KnowItAll Software Training

SmartSpectra Databases and Classification Models

Using SmartSpectra in KnowltAll

Purpose

These exercises demonstrate how to use SmartSpectra in KnowltAll ID Expert and Searchlt.

Objectives

These exercises will teach you:

- ➤ How to use SmartSpectra IR spectra in KnowItAll SearchIt
- ► How to use SmartSpectra Raman spectra in KnowItAll SearchIt
- ➤ How to use SmartSpectra IR spectra in KnowItAll ID Expert
- > How to use SmartSpectra Raman spectra in KnowltAll ID Expert

Training Files Used in This Lesson:

- 4-(Pyridin-3-yl)-2-2,6,2-terpyridine.irf
- 4,13-DIDECYL-1,7,10,16-TETRAOXA-4,13-DIAZACYCLOOCTADECANE.irf
- X-Phos.irf
- 1-acetyl-1234-tetrahydroquinoline.irf
- 17-Hydroxy-17-alpha-pregn-4-en-20-yn-3-one.irf

Note: The training files used are for example purposes only. The user should utilize their own IR spectra when following this training.

KnowItAll Applications Used:

- KnowltAll SearchIt
- KnowItAll ID Expert



Background

Wiley has the largest quantity of experimental IR spectra, which includes the Sadtler IR collection. Even with the continued development in the chemical industry, there is still insufficient progress to cover the newly-discovered chemical space. Sample collection is also challenging, time-consuming and costly. Wiley's IR SmartSpectra Collection is an attempt to increase the catalog coverage of Wiley's IR chemical space. This does not mean that this database will increase the chemical space, but that it will increase the amount of coverage within the bounds of our current libraries' chemical space. With the development of computer modeling technology, the feasibility of using computed IR spectra to compensate for the lack of samples was investigated and found to increase unknown compound classification.

These libraries can be helpful when used as tools for searching unknown compounds, illuminating the composition of an unknown spectrum, or accurately predicting the compound's spectrum for the associated structure and functional groups. Due to the predicted nature of this data, it should be mentioned that the results have a margin of error. Wiley acknowledges that some results may not be entirely correct. It is advised that users employ these libraries as additional tools to help the user classify components that are in the spectrum in order to characterize their unknown spectrum.

KnowItAll IR and Raman Search Algorithms

A background in the algorithms used by KnowltAll will be beneficial. For IR and Raman spectral comparison, KnowltAll uses the following algorithms:

Correlation

This is the default algorithm for searching in KnowltAll and it conforms to the industry standard for correlation algorithms. The Correlation algorithm is similar to the Euclidean Distance algorithm. The difference between the two is in the way the spectra are treated before the comparison. Each spectrum is mean-centered prior to performing the dot product normalization. This approach can improve search results for noisy spectra and spectra that have baseline issues, particularly with a baseline offset that is the result of a negative spike or chemical noise. It is slightly more time-consuming than the Euclidean Distance algorithm. The search speed is slower because each spectrum in the database must be mean-centered and then normalized prior to the comparison. The search results that are obtained with the Correlation algorithm are spectrally similar to the unknown, even if the unknown compound is not in the database. The Correlation algorithm is heavily weighted by peak areas. Broad features are weighted much more strongly than sharp features. The algorithm is most tolerant to peak shifts and non-linearities in relative band intensities.

Correlation (Classic)

The Correlation algorithm that was found in all versions of KnowltAll prior to KnowltAll 2020 is similar to the Euclidean Distance algorithm. However, it did not conform to the industry standard for correlation algorithms. Beginning with KnowltAll 2020, the Correlation algorithm does conform to the industry standard and it is the default algorithm used for searching in KnowltAll. To provide backward compatibility for customers who want to reproduce prior search results, the previous correlation algorithm is now provided as Correlation (Classic).

Euclidean Distance

The Euclidean Distance algorithm measures the point-to-point differences between a pair of spectra. The results that are obtained with the Euclidean Distance algorithm are spectrally similar to the unknown, even if the unknown compound is not in the database. This algorithm, however, can yield degraded search results when the unknown spectrum has a sloping or offset baseline. The Euclidean Distance algorithm is heavily weighted by peak areas. Broad features are weighted much more strongly than sharp features. This algorithm is most tolerant to peak shifts and non-linearities in relative band intensities.



First Derivative Euclidean Distance

Use this algorithm to reduce the effects of baseline slope or offset in the unknown. Although the search speed is slightly slower than with the Euclidean Distance algorithm, the First Derivative Euclidean Distance sometimes gives improved search results, especially when the unknown spectrum is a mixture of two or more compounds. The First Derivative Euclidean Distance algorithm is heavily weighted by changes in slope. Sharp features are weighted much more strongly than broad features. The algorithm is also very sensitive to peak shifts. Small shifts can make the algorithm miss a similar result.

Second Derivative Euclidean Distance

Use the Second Derivative Euclidean Distance algorithm to compare the second derivative of a reference spectrum to that of the query spectrum.

Optimized Corrections: A Breakthrough Technology for Spectral Searching

Spectral searching is one of the most important tools researchers use to classify and/or identify materials, yet it continues to be plagued by errors and imperfections. During a spectral search, a sample spectrum is compared to a database of reference spectra. To ensure that an optimal match is found in the database, spectra can be adjusted to compensate for differences between spectra caused by variability in instruments, accessories, environmental conditions, and other factors.

According to ASTM's guide on spectral searching¹, various algorithms and manual methods exist to adjust spectra to get reasonable match scores when two compared spectra of the same compound differ for various reasons. While these methods may work in selected cases, subtle discrepancies such as a shift of the X-axis are very hard to identify and correct manually. The inflexible mathematical algorithms typically employed do not compensate for these types of errors in spectra that are flawed.

Manual corrections can be made by expert spectroscopists, but those less experienced in spectroscopy are often unaware of how to perform the necessary corrections on their sample spectrum to achieve the best search result. To address this growing concern, Wiley has introduced Optimized Corrections, a breakthrough patented technology that performs a computationally complex set of multiple corrections on query and reference spectra in a search to find the optimal match between the query and each individual reference spectrum. This training guide will demonstrate how the Optimized Corrections technology yields better matches between query and reference spectra than can be attained using rigid search algorithms alone or with manual methods to optimize spectra for searching.

Optimized Corrections consider the full spectrum during a selected range(s) search.

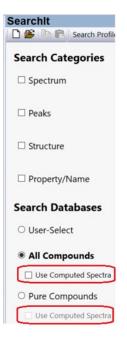
¹ E2310-04 - Standard Guide for Use of Spectral Searching by Curve Matching Algorithms with Data Recorded Using Mid-Infrared Spectroscopy, 2009. ASTM International Web Site. http://www.astm.org/Standards/E2310.htm (accessed March 4, 2015).



SearchIt

Introduction

When a user license contains the subscription to SmartSpectra IR spectra, it can be accessed in the **SearchIt** application using **Search Databases > All Compounds** or **Databases > Pure Compounds** and by checking the box for the **Use Computed Spectra** option:



Example 1

Example File: 4,13-DIDECYL-1,7,10,16-TETRAOXA-4,13-DIAZACYCLOOCTADECANE.irf

	Action	Result
1	In the SearchIt application, click Open Spectrum or Structure icon () located on the Standard Toolbar.	Note: The training files used in this document are for example purposes only. The user should utilize their own IR spectra when following this training.
2	Navigate to find 4,13-DIDECYL-1,7,10,16- TETRAOXA-4,13- DIAZACYCLOOCTADECANE.irf in "C:\Users\Public\Documents\Wiley\ KnowltAll\Samples\Computed Spectra\IR" folder. Click Open. Confirm that the Search Method is set to Correlation algorithm. To change the method, use the dropdown menu to select it as the option. Search Method: Correlation Under the Search Databases section, select All Compounds with the 'Use Computed Spectra' box checked as well. All Compounds Use Computed Spectra	The image on this page is the correct compound structure associated with this spectrum. Upon executing the search, Minelt application displays the query results. Image I
	Click Search.	Sort in Processing Special Analysis



In Minelt, the hit list table displays match results featuring a mix of predicted and empirical matches.

The structure below is the correct match for the query spectrum's structure.

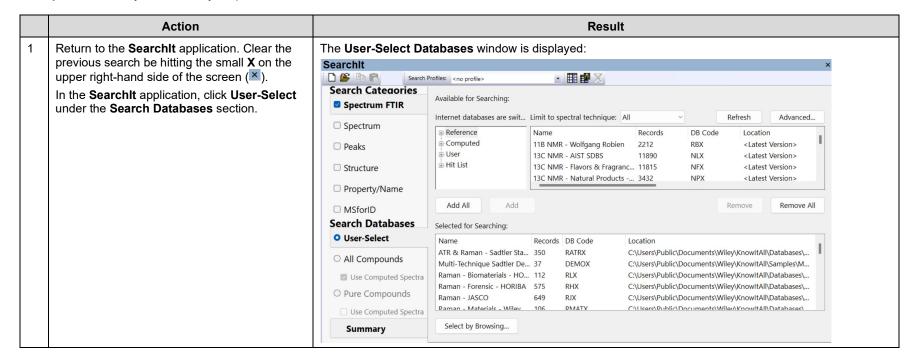
The structure below is an exact match with the predicted data which appears as the top hit in the table.

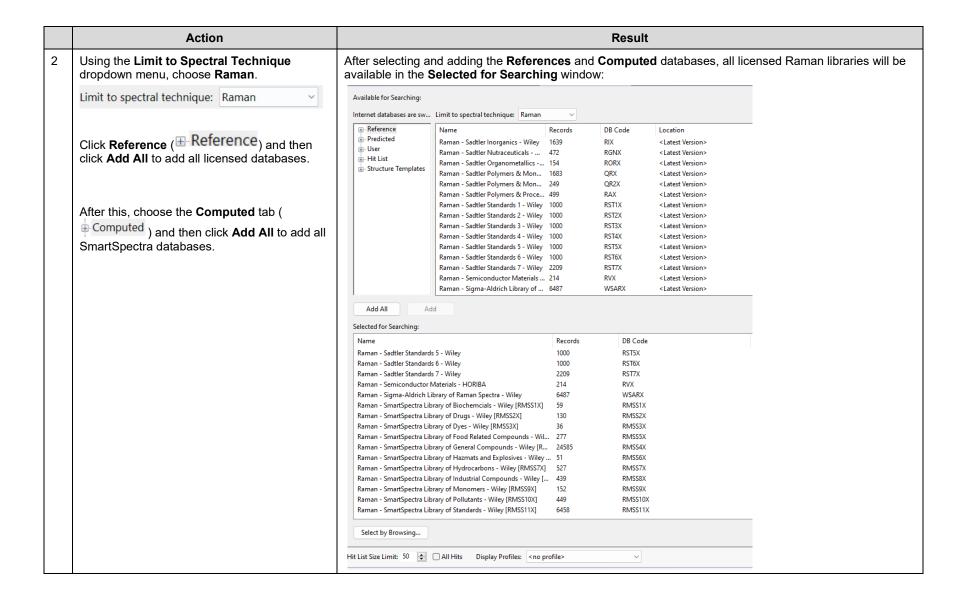
Note: Specific results will depend on the available databases in the user license.

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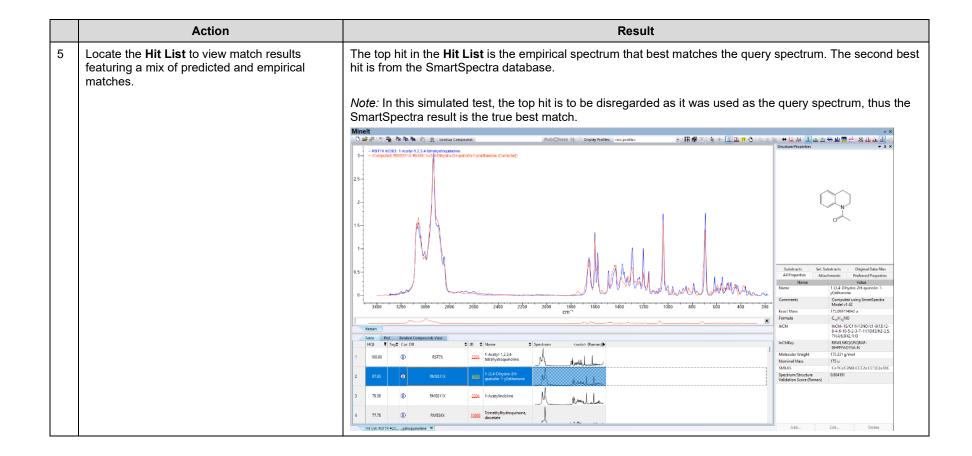
Example 2

Example File: 1-acetyl-1234-tetrahydroguinoline.irf





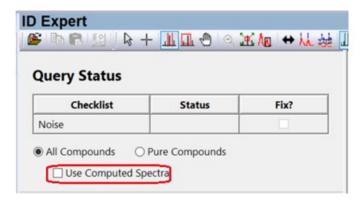
Action Result Click Open Spectrum or Structure icon () The selected file opens in the **Spectrum Search** window: located on the Standard Toolbar. 1 3 + 1 1 1 3 3 3 2 2 1 + 2 1 1 1 1 + 2 1 Navigate to find 1-acetyl-1234tetrahydroguinoline.irf in "C:\Users\Public\Documents\Wiley\ KnowItAII\Samples\ Computed Spectra\Raman". Click Open. *Note:* The image on this page is the correct compound structure associated with this spectrum. In the Spectrum Search window, click The Advanced Settings pop-up window is displayed. Upon clicking OK the window closes. Upon Advanced Settings and deselect Remove executing the search, Minelt loads with the query results. **Duplicates** and **Remove Replicates**. Click Advanced Settings OK. ☐ Remove Duplicates Optimized Corrections Click Search. ✓ Enabled ☐ Remove Replicates ✓ Baseline ✓ Vertical Clipping ✓ Intensity Distortion ✓ Horizontal Offset ✓ Vertical Offset Set As Default ☑ Raman Intensity Distortion Reset To Default



ID Expert

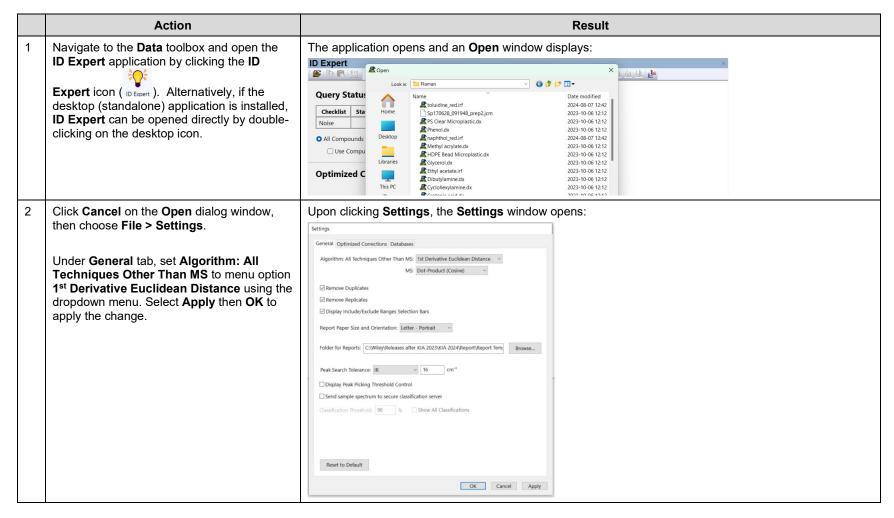
Introduction

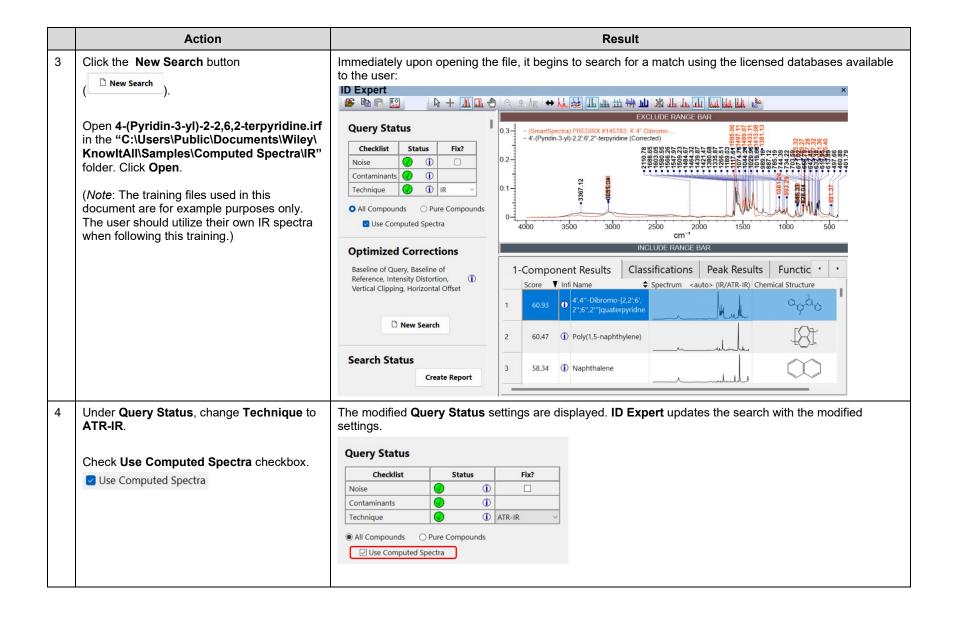
When a user's license contains the subscription for predicted IR spectra, the ID Expert interface Query Status shows the Use Computed Spectra option:

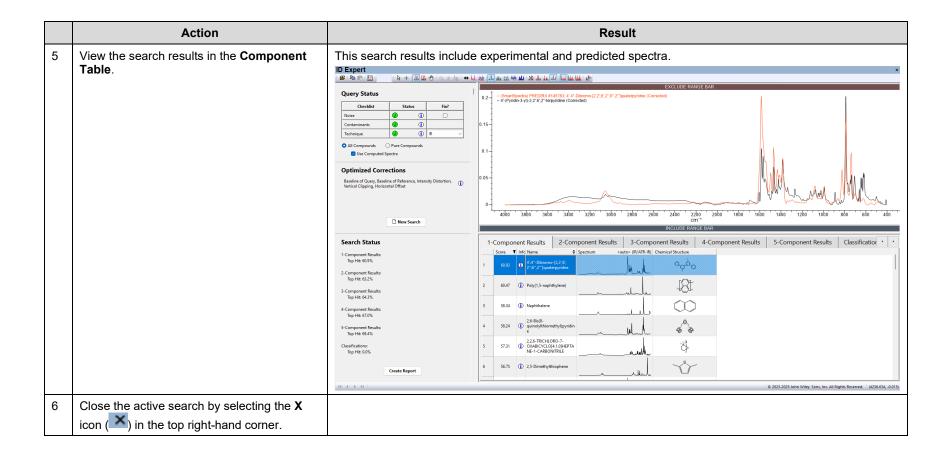


Example 3

Example File: 4-(Pyridin-3-yl)-2-2,6,2-terpyridine.irf

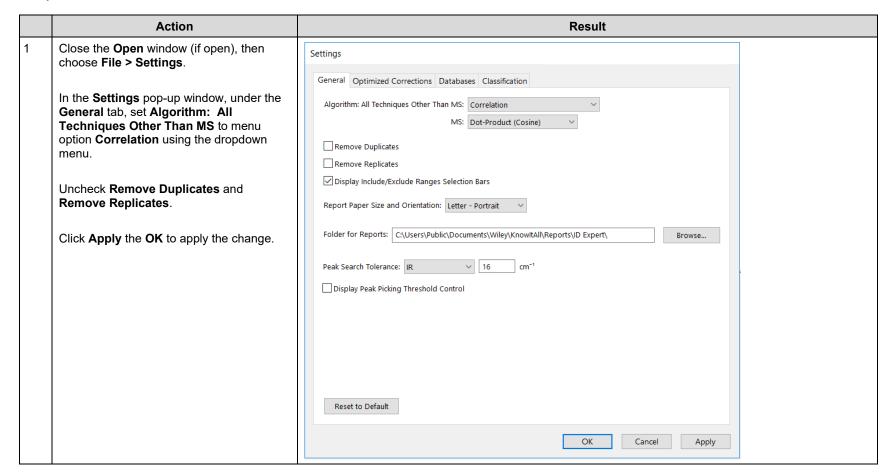


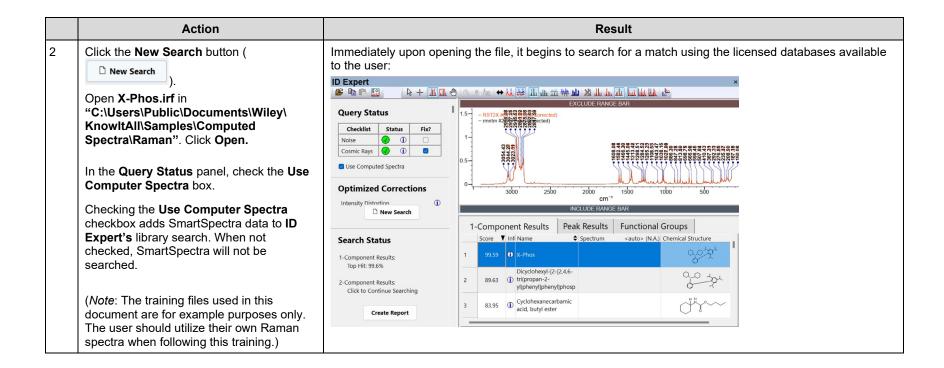


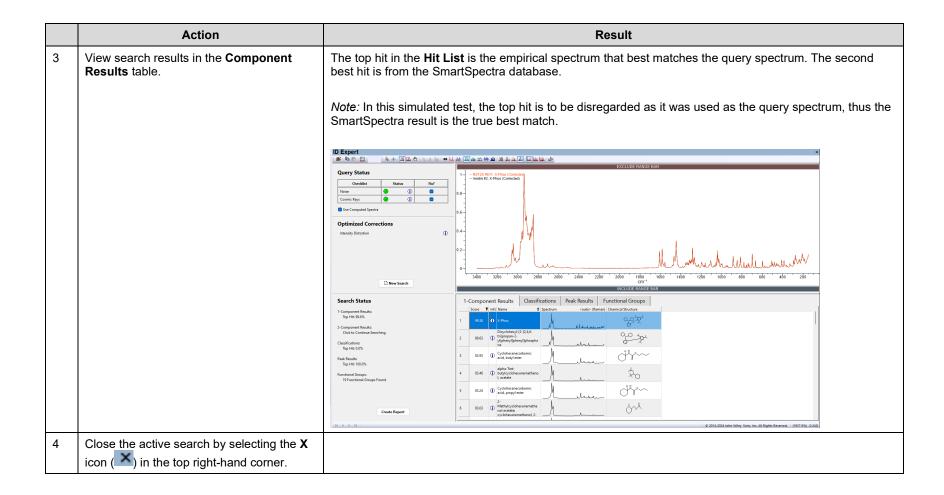


Example 4

Example File: X-Phos.irf







Classification Models

Purpose

These exercises demonstrate how to use classification models in KnowltAll ID Expert and Minelt.

Objectives

These exercises will teach you:

- > How to use the KnowltAll Classification models in ID Expert
- > How to use the KnowltAll Classification models in Minelt

Training Files Used in This Lesson:

- 17-Hydroxy-17-alpha-pregn-4-en-20-yn-3-one.irf
- SmartSpectraFTIROxycodone.irf
- SmartSpectraRamanAndrosteroneacetate.irf

Note: The training files used are for example purposes only. The user should utilize their own IR spectra when following this training.

KnowItAll Applications Used:

- KnowItAll ID Expert
- KnowltAll Minelt

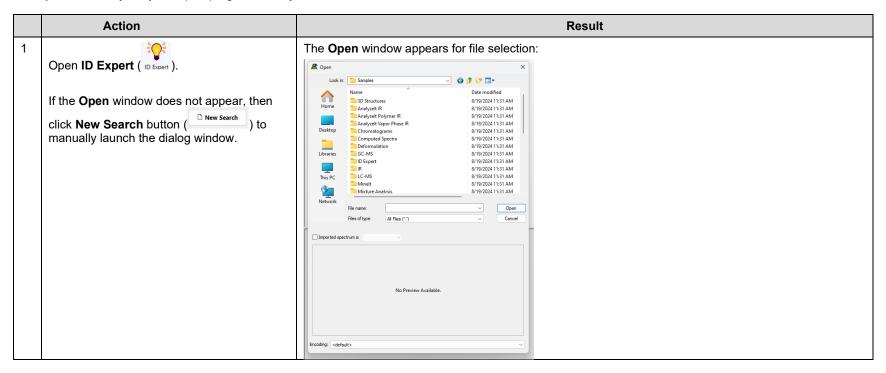


Background

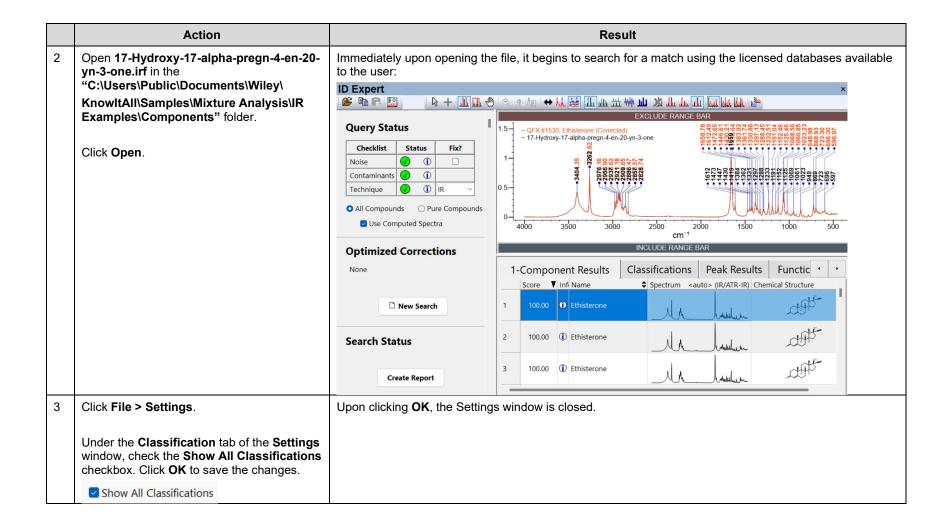
Classification models are a type of supervised learning used to label or categorize data into predefined classes or groups. These models predict the said label based on feature data. The model must be trained on labeled data or have an architecture that leads to the data being trained on to be self labelled by the model. Our models are binary classifiers that involve either a true or false result. The model also uses the neural network algorithm architecture, which has gained much popularity in recent years. The models have been optimized for FT-IR, Raman and GC-MS techniques within KnowltAll's ID Expert and Minelt applications.

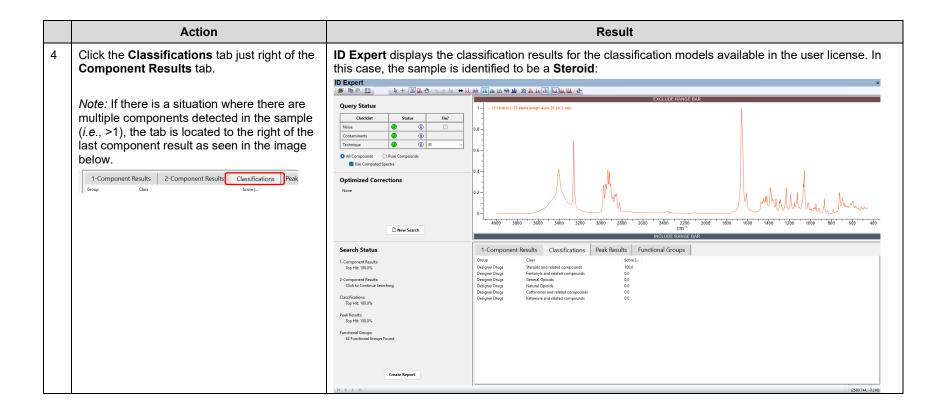
ID Expert

Example File: 17-Hydroxy-17-alpha-pregn-4-en-20-yn-3-one.irf





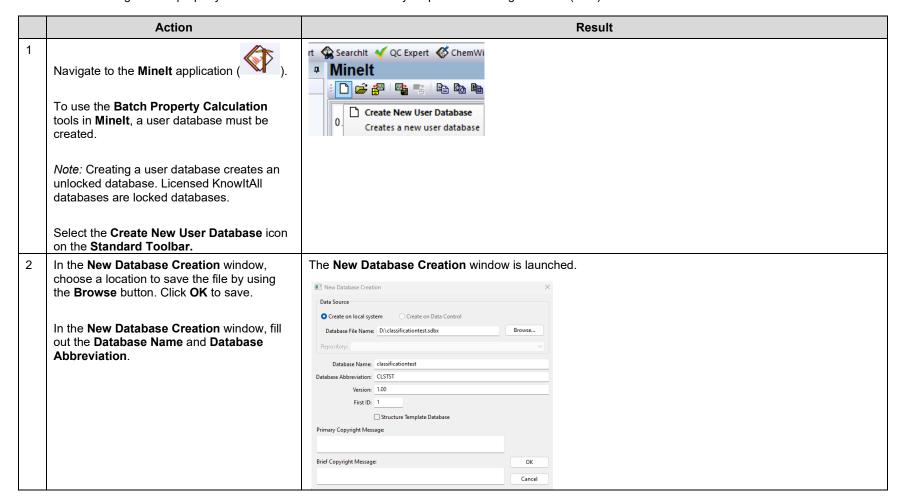




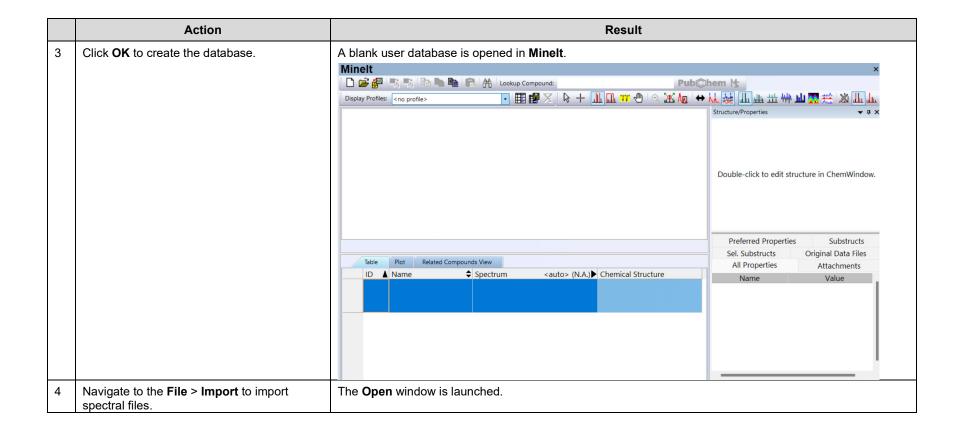
Classification Batch Property Calculation in Minelt

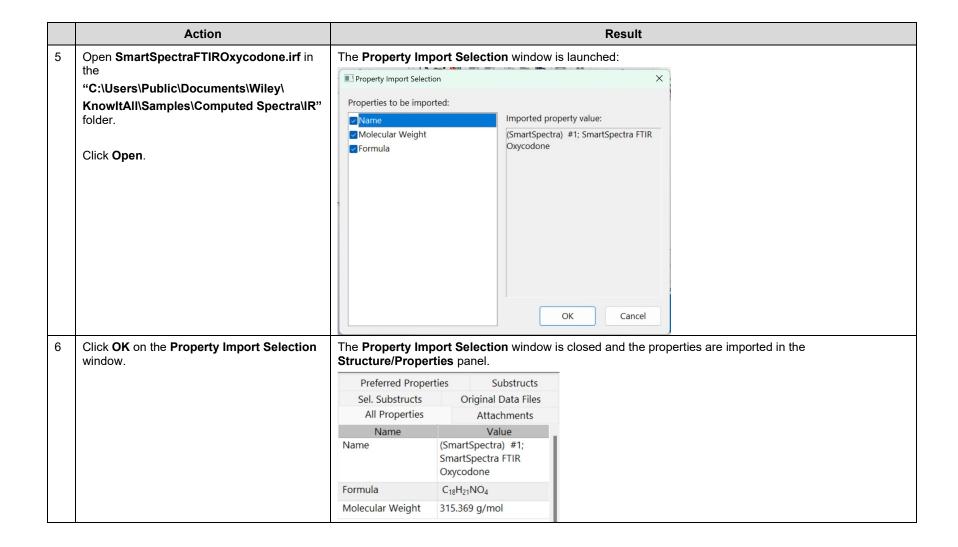
Example Files: SmartSpectraFTIROxycodone.irf, SmartSpectraRamanAndrosteroneacetate.irf

Batch property calculation is a calculation that will be applied to a series of compounds at one time. In this exercise, an unlocked user database will be created that to simulate executing a batch property calculation. This workflow can only be performed using unlocked (user) databases.

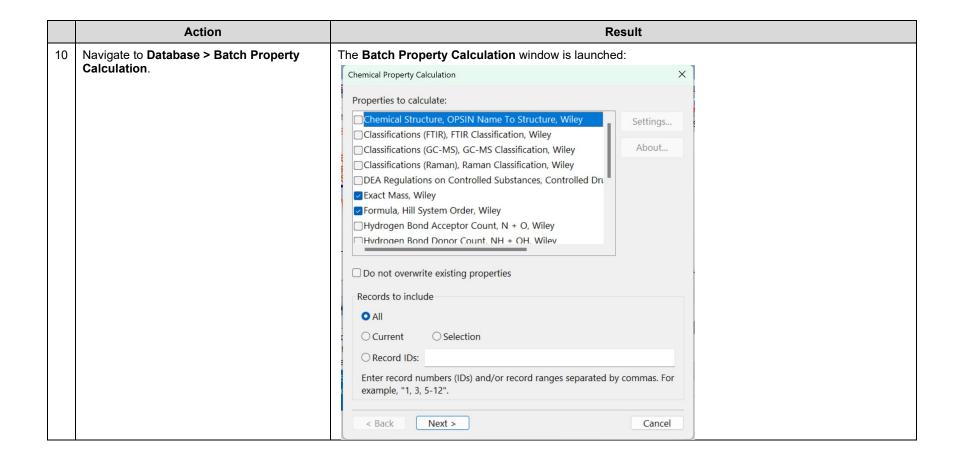


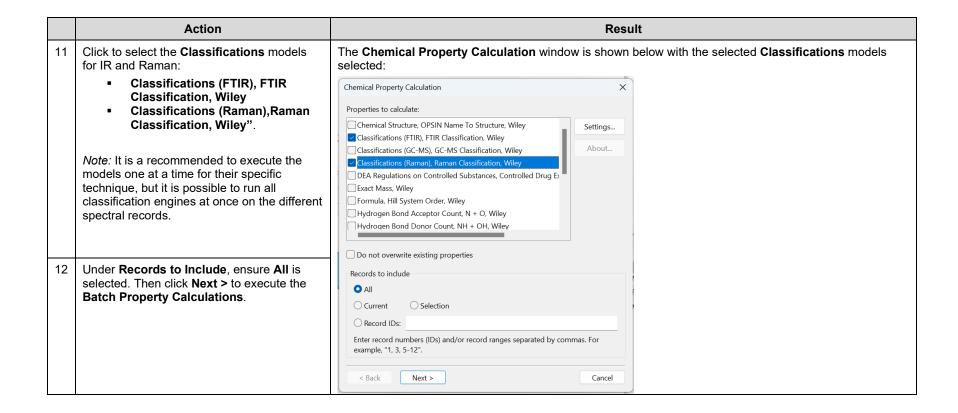


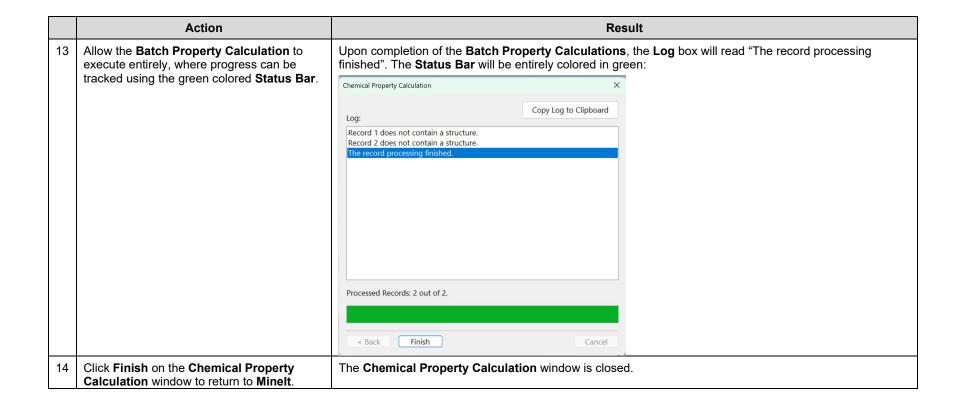




	Action	Result			
7	Click on the row below the first record in the Table .	The second row is shaded in the Table :			
		Table Plot Related Compounds View ID ▲ Name			
8	Navigate to the File tab and select Import to import a second file into the user database.	The Open dialog window is launched.			
9	Repeat steps 4 through 8 using SmartSpectraRamanAndrosteroneacetate .irf in the "C:\Users\Public\Documents\Wiley\Knowl tAll\Samples\ Computed Spectra\Raman" folder.	The Property Import Selection window is launched. The Raman sample file is imported into the user database.			







	Action			Result
15	The resulting classification can be found in the Structure/Properties panel for the selected record.	Record ID 1 I (100.0%)".	has results Class	ifications (FTIR) "General Opioids (100.0%)" and "Natural Opioids
	Scientifica record.	Preferred Prope	erties Substructs	
		Sel. Substructs	Original Data Files	
		All Properties	Attachments	
		Name	Value	
		Name	(SmartSpectra) #1; SmartSpectra FTIR Oxycodone	
		Classifications	General Opioids (100.0%) Natural Opioids (100.0%)	
		Formula	C ₁₈ H ₂₁ NO ₄	
		Molecular Weight	315.369 g/mol	
		Sel. Substructs All Properties Name		
		Classifications	Steroids and related compounds (100.0%)	
		Formula	C ₂₁ H ₃₂ O ₃	
		Molecular Weight	332.484 g/mol	
16	The final compound structures are provided in the Results cell for reference. These are the structural representations of the two spectra used in the batch property calculations added for comparison to show that they are classified correctly.	Record ID 1:	(SmartSpectra)	#1; SmartSpectra FTIR Oxycodone.

Action	Result
	Record ID 2: #2; SmartSpectra Raman Androsterone Acetate.
To view all the different spectral techniques in the database spectrum pane, select the black triangle to the right of ' <auto> (IR)' or</auto>	Spectrum <auto> (IR) <auto> <auto> <auto <aut="" <aut<="" <auto="" th=""></auto></auto></auto></auto>
' <auto> (Raman)' part of the pane.</auto>	Vapor Phase IR Near IR Raman
When the dropdown appears, select the	Raman (SERS) UV-Vis
technique you would like to view.	XRD XRF
	MS (GC) MS (LC)
	MS (IMS) GC LC
	IMS 1 ^t H NMR
	1 ³ C NMR 1 ⁵ B NMR
	1 ⁵ N NMR 1 ⁷ O NMR
	19F NMR 29Si NMR
	³¹ P NMR Other