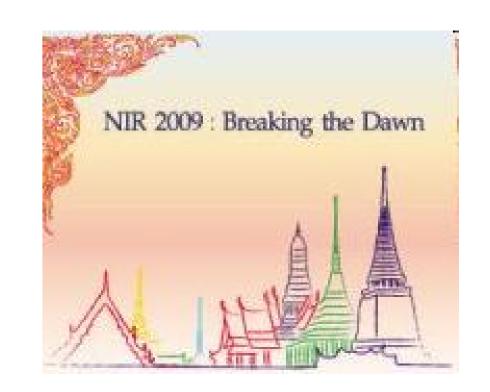


Multi-spectral assessment of ingredients and physical properties of apricot



Ferenc Firtha, Zoltán Gillay, Eszter Vozáry, Tímea Kaszab, Anikó Lambert-Meretei and József Felföldi

Corvinus University of Budapest, Department of Physics and Control, 14-16 Somlói str, Budapest, H-1118, Hungary, ferenc.firtha@uni-corvinus.hu

About us

At the Corvinus University of Budapest (physics.uni-corvinus.hu)

•Faculty of Horticulture •Faculty of Food Science

Department of Physics and Control

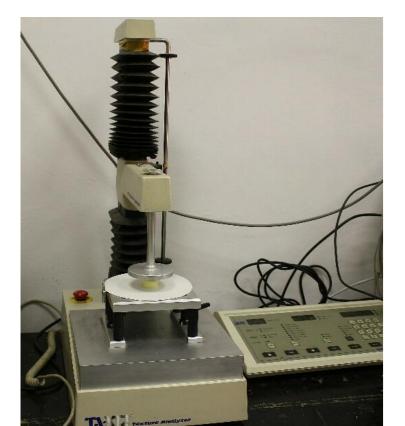
physical properties of food, their raw materials, fruits and vegetables are investigated:

•Rheological by static and dynamic methods

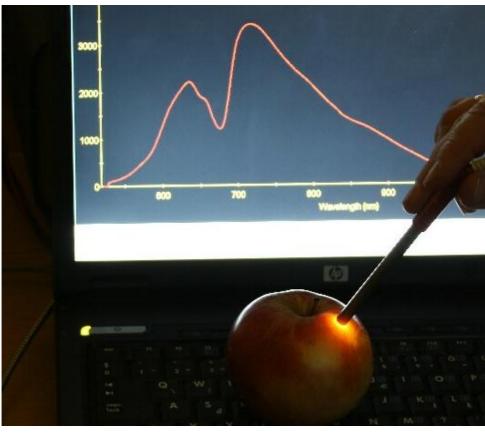
•Dielectric, chemical (e-tongue)

•Optical by image processing, scattering, spectroscopy, multiand hyperspectral imaging methods

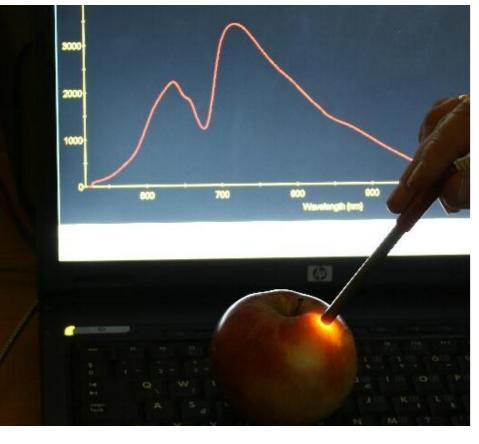
for basic research and industrial quality control, automation purposes.

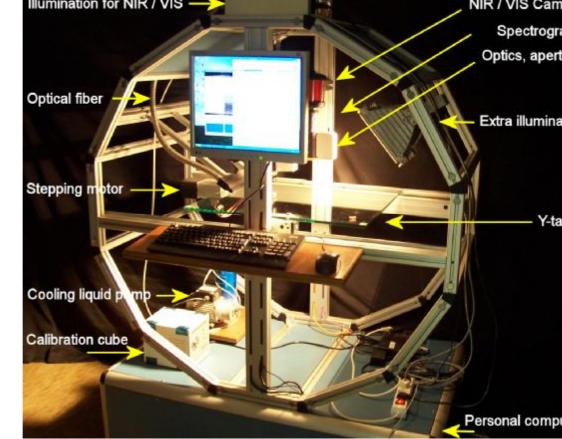


Penetrometer (static method)



Scattering at multiple wavelength





Capital of Hungary at night

Zeutec hyperspectral measurement setup

Introduction

Some quality-related internal parameters of apricot can be predicted by the reflected NIR spectrum. According to the recent publications (Bureau et al, 2009, Camps and Christen, 2009), using 800-2500nm range, the soluble solids content (SSC) and the titratable acidity (TA) can be predicted properly, but other quality traits, like malic and citric acid, individual sugars, ethylene production and firmness were not satisfactory modeled.

The non-destructive, non-contact and fast optical measurement methods, like hyper- or multispectral imaging are more and more demanded for on-line industrial quality control tasks. These methods combine the advantages of spectroscopy and conventional image processing, obtain the spatial distribution of spectral properties on nonhomogeneous surfaces (Figure 1), but have much less spectral information in usually noisy environment.

For testing the feasibility of multispectral industrial application, the internal-, rheological and optical properties of apricot cultivars were investigated. Samples of three cultivars, three ripening state and three further categories by storage time were measured with 20 samples in each group. All the measurements were taken on both blushed and un-blushed side as well (Figure 2).

The optical properties were measured and checked in different spectral ranges with different instrumentation. Mechanical properties of the samples were measured on dynamical way, with two impact methods and an acoustic response system. The chemical properties were measured after all non-destructive methods mentioned above.

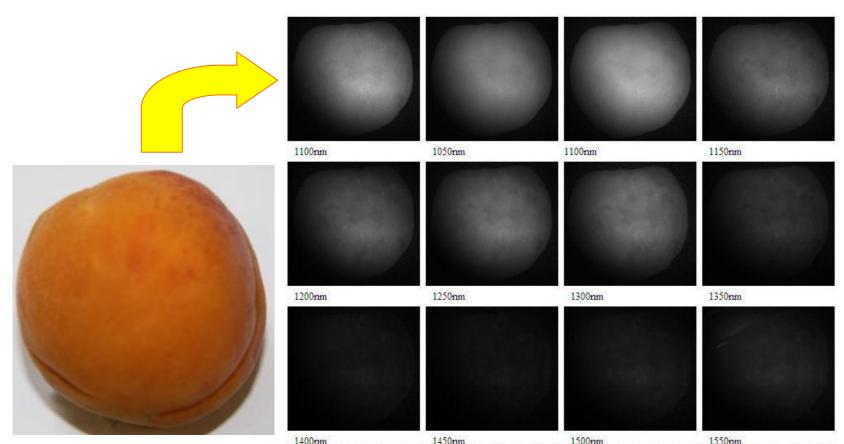




Fig. 1: Multispectral images of defected Bergarouge sample

Fig. 2: Bergeron cultivar in three ripeness state

Materials and methods

In the experiment described below, samples of Bergeron, Bergarouge and Zebra apricot cultivars were tested grouping in three ripeness category (1. immature, 2. ripened for processing, 3. ripened for consumption). The second category was stored for one week (4. category) and two weeks (5. category).

The mass (m) and the three perpendicular diameters (d_1, d_2, d_1) were recorded. The following optical, mechanical and chemical parameters were inspected on both blushed and un-blushed side (Figure 3):

Optical	RGB Imaging System (RGB images using diffuse illumination):
	average value and variance of XYZ colour components of segmented areas
	• Pigment Analyzer (400-1090 nm range, 3.25nm resolution):
	reflected spectra, Normalized Difference Vegetation Index, Normalized Anthocyanin Index
	• ColorLite sph850 spectrophotometer (400-700 nm range, 10nm resolution):
	reflected spectra, CIE Lab, Luv and XYZ coordinates
	• PCM Spectralyzer 10-25 (1000-2500 nm range, 2nm resolution) (Figure 5):
	the average reflectance on a 25mm diameter area
	• NIR Multispectral Imaging system (12 images at 1000-10501550 nm) (Figure 8):
	average value and variance of intensity values on segmented areas
Mechanical	• acoustic resonance method:
	measured: resonance frequency (f, Hz) and width of the resonance peek at -3 dB (bw, Hz)
	calculated: $\mathbf{s_1} = \mathbf{f}^{2*}\mathbf{m}^{2/3}$ and $\mathbf{s_2} = \mathbf{f}^{2*}\mathbf{d_1}^2$ acoustic stiffness coefficients
	• impact method:
	measured: time of deceleration of the impact hammer (dT , ms)
	calculated: $\mathbf{D} = 1/dT^2$ impact stiffness coefficient
	Sinclair Internal Quality tester
	measured: Sinclair firmness coefficient on 1-100 scale (IQ)
Destructive	• pH of the apricot flesh (Vaiseshika pH-conductivity TDS+DO meter and inserted flesh probe)
	• SSC (Brix) of the apricot juice (Atago digital refractometer PAL-1)
	• Sugar-content of groups (fructose, glucose, saccharose, xylose, raffinose) (HPLC)
	• Titratable acid content of groups (titration)





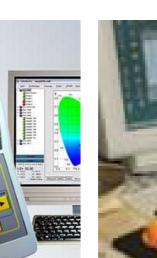








Fig.3: Pigment Analyzer, Color Lite; Acoustic method, Impact method, Sinclair IQ tester; Brix tester, pH-meter

The correlation between NIR spectrophotometer measurement and different quality traits was investigated by Partial Least Squares (PLS) linear regression method. Since randomly selecting two third of whole dataset as calibration subset did not resulted reliable prediction, the first (immature) and third (ripened for consumption) groups were used as calibration set having a wide distribution of internal properties. The optimal number of latent variables was determined on the base of the minimal value of RMSEV and the maximal value of Relative Performance Determinant (RPD). RPD is defined as the standard deviation of the reference values of all samples devided by the standard deviation of the error of the validation set. The overfitting of regression was checked by the B-vector.

The multispectral images were segmented, than the average spectra of investigated area were compared with spectrophotometer measurements. The variance of spectra on investigated area was also checked. The prediction model was also the PLS method calculating the correlation, the RPD, RMSEC and RMSEV values and the B-vector. Multi-linear regression (MLR) method was used to determine the significant wavelengths of given internal properties. All, the statistical algorithm were developed in Mathcad software package (ver. 14.0, MathSoft, USA).

Results

All the measured optical (e.g. Lab, NDVI, NAI) and mechanical (Sinclair-, impact- and acoustic stiffness) parameters were changed monotonically by the ripeness state and by the storage time as well (Table 1). The chemical properties, however, behave irregularly. The pH increased during ripening, but it grown significantly only at the first week of storage. The soluble solids content (Brix) increased during ripening, but has not changed during storage. The change of sugar-content (fructose, glucose, saccharose, xylose, raffinose) and titratable acid (TA) was not commonly monotonous by ripeness state, neither by storage time (Table 2). This paper will focus on the prediction of **pH** and **Brix**.

		Geometric				Stiffness				Reflectance in visible range																					
												un-bl	ushed		ed side		ushed	blushe				un-blu	shed					blushe	d side		
cultivar	ripeness / storage	Mas	s, g	Volum	_	Sinc		Imp		A∞ı	ustic		NI	M		Norma	lized An	thocyani				а		Ŀ	•	L		a'		b	
l	categories	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std	avg	std
	:t (4)	EE 4	40	52,1	4.0	440	2.5	0.440	0.000	20.50000	475643	0.07	0.20	0.00	0.04	0.77	0.45	0.50	0.48	57.0	27	6.7	4.0	44.0	2.7	40.5	40	45.4	E 0	24.2	E 4
<u>!</u>	immature (1)	55,4	4,9	-	4,3	14,0		0,140		2050099 2135610	370329	-0,07	0,20	0,06 -0,02	0,21	0,77	0,15	0,56 0,78	0,18	57,8	2,7	6,7	1,9	41,0	2,7	49,5	4,8 1,7	15,4	5,3	31,3	5,4
ŧ	processing (2) consumption (3)	57,8 65,9	6,6	54,7 61,5	6,6	12,1 9,8		0,133		1572507	294058	-0,10 -0,30	0,18	-0,02	0,20	0,90	0,07	0,78	0,12	51,5 56,3	1,3 2,4	11,5 14,2	1,6	40,3 40,8	2,1	46,5 45,5	3,9	14,0 22,1	2,0	31,9 26,5	2,0 5,6
Bergeron	cors uniption (3)	65,5	0,0	61,5	7,1	3,0	در ا	0,052	0,017	157 2507	254050	-0,30	0,10	-0,24	0,21	0,52	0,00	0,00	0,08	36,3	2,4	14,2	2,0	40,0	2, 1	45,5	3,3	22,1	2,0	20,0	5,0
beige on 1		57.0	00	E4 7	00	424	4.2	0.422	0.017	2425040	370329	0.40	0.10	-0,02	0.20	0.00	0.07	0.70	0.42	EAE	4.2	44 E	4.0	40.2	4.7	40.5	47	44.0	20	24.0	20
<u> </u>	processing (2)	57,8 59,4	6,6	54,7 54.4	6,6	12,1		0,133		2135610 1011637	485148	-0,10	0,18	-	0,20	0,90	0,07	0,78	0,12	51,5 49,3	1,3	11,5 12,5	1,6 3,6	40,3 33,7	1,7 2,8	46,5	1,7 4,9	14,0	2,0	31,9	2,0 5,5
i	1 week (4) 2 weeks (5)	44.2	7,8	37.7	6,9 8,6	10,3	1,2	0,073	0,020	186192	107327	-0,46 -0,47	0,06	-0,48 -0,51	0,07	0,84	0,13	0,61 0,75	0,10	45,8	2,8 4,2	11,8	2,9	33,0	4,0	41,1 36,0	4,0	19,4 19,2	4,1 2,8	23,5 19,6	6,1
<u></u>	2 Weeks (5)	44,2	8,8	31,1	0,0	-		0,035	0,010	100132	107327	-0,47	0,07	-0,51	0,04	0,50	0,05	0,75	0,12	40,0	4,2	11,0	2,5	33,0	4,0	36,0	4,0	15,2	2,0	13,6	0, 1
<u>T</u>	immature (1)	64,7	8,3	60,6	77	14,9	2.5	0,141	0,027	2955759	1117084	-0,11	0,15	-0,08	0,19	0,69	0,23	0,35	0,33	58,1	1,7	7,2	2,7	41,5	2,4	49,7	4,2	14,9	E.B.	31,6	5,5
į		-		_	0.0			-			847019	-		-0,00		_				_		_		_				_	5,6		
į	processing (2)	64,3	9,8	58,9	8,9	11,2		0,110		2085061 956134	315487	-0,29	0,11	-0,21	0,11	0,89	0,11	0,71	0,15	56,0 54,1	2,5	12,1	2,2	40,1 39,5	3,0	43,0	5,3	22,8	4,4	24,1	6,2 4,7
Bergarouge	consumption (3)	62,0	11,7	57,2	11,0	8,2	2,0	0,073	0,016	336134	310467	-	-	-	-	-	-	_	-	34,1	1,8	14,7	2,1	35,5	1,9	41,2	4,0	23,6	3,5	22,5	4,7
oe galouge	propossing (2)	64,3	9,8	58,9	8,9	11.2	2.6	0,110	0,022	2085061	847019	-0,29	0.11	-0,27	0,11	0,89	0,11	0,71	0,15	56,0	2.5	42.4	22	40,1	3,0	43,0	5,3	22,8	4.4	24,1	82
İ	processing (2) 1 week (4)	51,4	5,0	46,2	4,9	,				461506	202082	-0,25	0,08	-0,27	0,11	0,84	0,09	0,60	0,18	47,7	2,5	12,1 10,5	3,4	_	1,8		3,8		3,9	23,1	6,2 4,4
+	(-)	45,4	7,3	39,4		13,5	12,4	0,034	0,012	186482	131904		0,00		0,07			0,60	0,10	45,9	2,0	11,2	2,1	31,6 33,1	2,1	41,6 37,3	3,4	16,9 17,0	2,9	21,9	4,9
<u>i</u>	2 weeks (5)	45,4	7,3	39,4	7,0	-		0,034	0,011	100402	131904	-	-	-	-	-	-	-	_	45,5	2,0	11,2	2,1	33,1	2,1	31,3	3,4	17,0	2,5	21,9	4,5
T	immature (1)	51,3	5.2	46,6	4,9	21,8	2.1	0,228	0,041	4262904	1220702	0,22	0,28	0,21	0,27	0,83	0,10	0,76	0,10	55,5	1,9	15,4	2,9	40,5	2,4	44,4	5,4	23,7	3,4	26,4	7.1
ŧ	processing (2)	56,3	5,0	52,2	4.4	20,3	4.1	0,228	0,030	3339701	1033934	-0,08	0,19	0,10	0,25	0,83	0,09	0,79	0,10	50,6	1,9	13,8	1,8	42,6	2,0	45,1	2,8	17,5	2,8	34,5	4,0
ŧ	consumption (3)	57,1	10,0	51,5	8,4	13,2	4.2	0,123	0,030	1609299	554809	-0,32	0,13	-0,31	0,25	0,91	0,08	0,77	0,11	50,6	1,5	16,1	1,2	39,4	2,3	45,1	2,3	20,4	1,7	31,5	3,2
Zebra	cors uniption (3)	57,1	10,0	51,5	0,4	10,2	7,2	0,123	0,027	1003233	554665	-0,32	0,11	-0,51	0,15	0,51	0,00	0,77	0,14	30,0	1,0	10,1	1,2	30,4	2,0	40,1	2,0	20,4	1,7	31,3	0,2
2000	processing (2)	56,3	5,0	52,2	4.4	20,3	41	0,181	0,030	3339701	1033934	-0,08	0,19	0,10	0,25	0,83	0,09	0,79	0,11	50,6	1,9	13,8	1,6	42,6	2,0	45,1	2,8	17,5	2,8	34,5	4,0
ŧ	1 week (4)	57,6	7,6	50,5	7,3	12,3		-		928187	282781	-0,39	0,13	-0,43	0,25	0,86	0,03	0,75	0,17	47,3	1,0	15,2	1,1	36,3	1,6	42,9	2,2	18,2	1,5	28,8	3,7
ŧ	2 weeks (5)	52,0	6,1	43,7	6,0	12,0	3,1	0,057	0,013	521339	168552	-0,49	0,00	-0,43	0,07	0,88	0,07	0,74	0,17	45,6	1,0	18,2	1,0	36,7	1,0	41,3	2,0	19,9	1,4	28,5	3,3
	Z WOEKS (U)	32,0	U, I	40,7	0,0	-		0,007	0,011	32 1333	100302	70,40	0,02	-0,51		0,00	0,07	C	1.4		1,0	10,2	1,0	30,7	1,2	41,3	2,0	10,0	1,7	20,3	3,3

Table 1: Average and standard deviation of quality traits



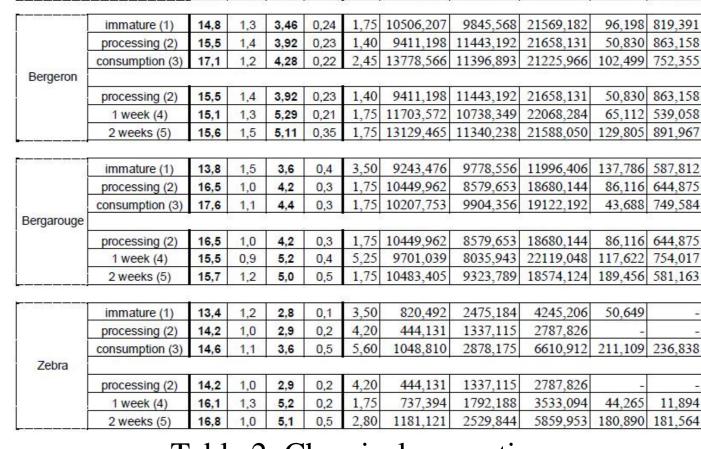


Fig.4: Bergeron, Bergarouge and Zebra cultivars

Table 2: Chemical properties



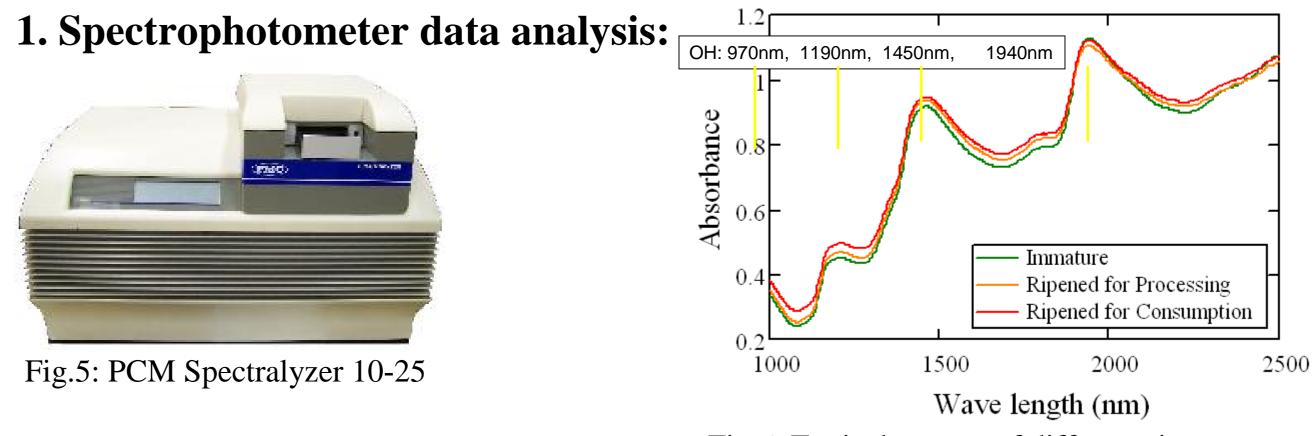
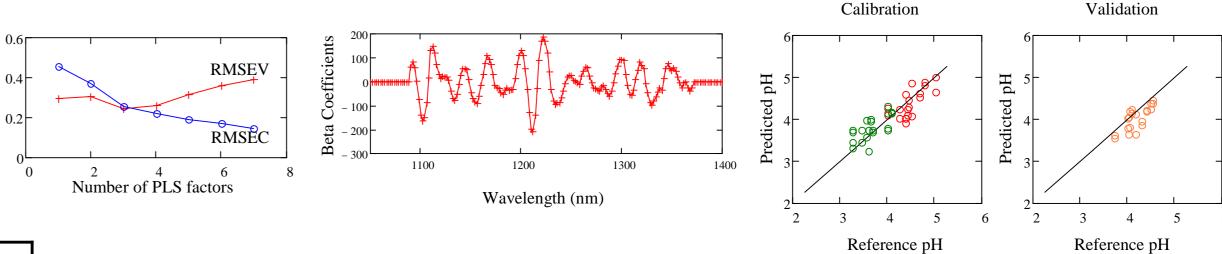


Fig.6. Typical spectra of different ripeness states (averages for all cultivars, items and sides)

Prediction of pH:



			Wavelength
LV=3	Cal	Val	
RPD	1.69	2.45	Eig 7 A) Determination of LV D) Ch
RMSE	0.26	0.24	Fig.7. A.) Determination of LV, B.) Ch
R ²	0.74	0.58	nH. The entimal number of fee

hecking overfitting, C.) Calibration and validation set

Prediction of Brix: LV=3 Cal

1.98

0.76

RPD

RMSE 0.97

pH: The optimal number of factors was found to be 3 (7.A). The diagram of B-coefficients appears smooth enough (7.B). The diagram of calibration and validation shows relationship (7.C). High RPD and small RMSEV signs, that this model is encouraging, despite of sample set contained both the spectra measured on blushed and un-blushed side.

Brix: RPD is less, RMSEV is higher, but it has the same correlation (R).

2. Multispectral data analysis:

0.58

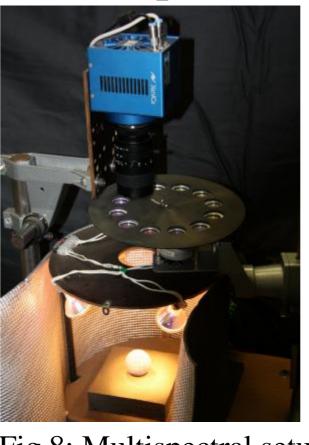


Fig.8: Multispectral setup

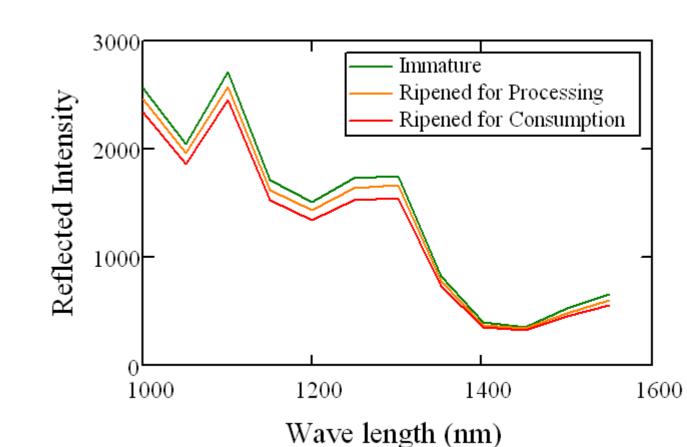
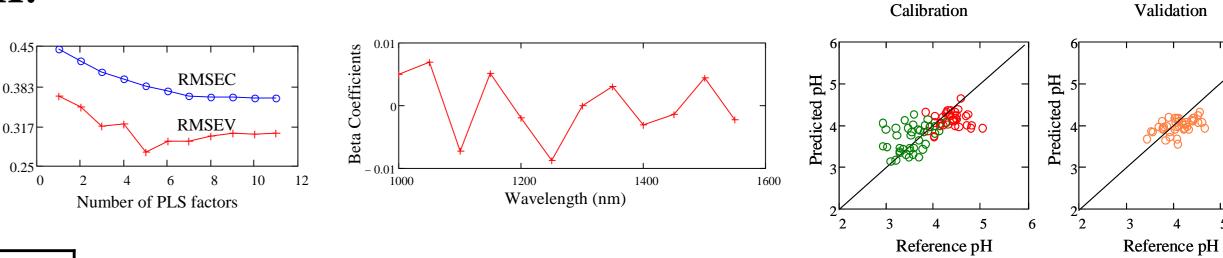


Fig.9. Typical spectra of different ripeness states (average for all cultivars, items and sides)

Prediction of pH:



			I
LV=5	Cal	Val	
RPD	1.19	1.73	
RMSE	0.38	0.27	
R ²	0.44	0.24	

Fig. 10. A.) Determination of LV, B.) Checking overfitting, C.) Calibration and validation set

Prediction of Brix: LV=3 Cal RPD 1.09 1.20 RMSE | 1.65 1.52 0.32 0.01

pH: Using multiplicative scatter correction, the RPD and the small RMSEV show acceptable relation. R2 is small, because the cultivars have different behaviour on these wavelengths. Building the model for given cultivar (e.i. Bergeron), the results were better. Without multiplicative scatter correction, only 2 factors resulted RPD=1.38 and RMSEV=0.37 values. The significant wavelengths were calculated by MLR method.

Brix: R² is almost zero, signing that these wavelengths are useless for predicting SSC. Significant wavelengths of this property must be studied.

Conclusion

The multispectral assessment of ingredients seems to be encouraging, but:

- •Set of samples must be selected for calibration, having wider range of properties.
- •All the noise, stray light should be especially excluded.
- •More chemical factors should be measured to explain the irregular changes of sugar and acid components (internal standard addition).
- •Significant wavelengths of properties will be studied by HSI method. •Image processing algorithm will be developed to segment blushed and un-blushed areas on multispectral images to improve efficiency.



