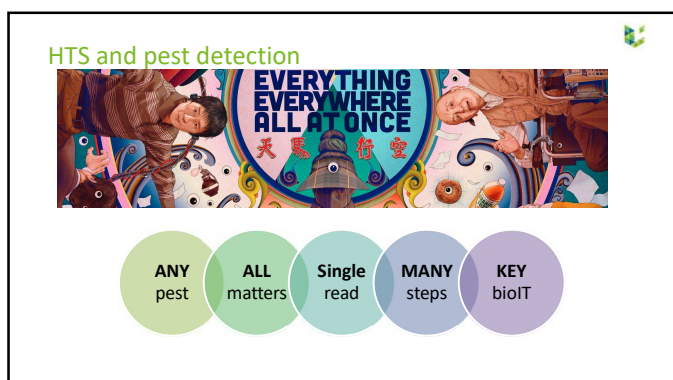




1



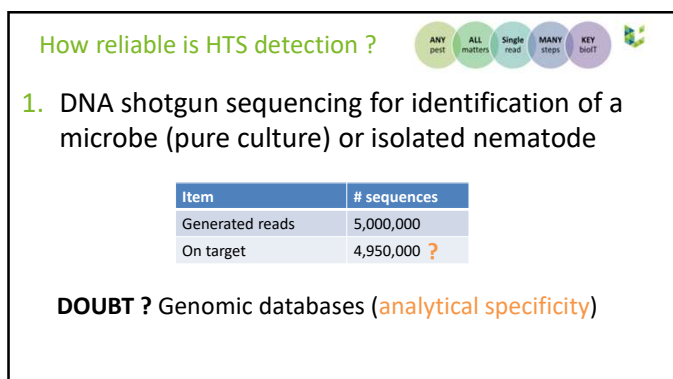
2



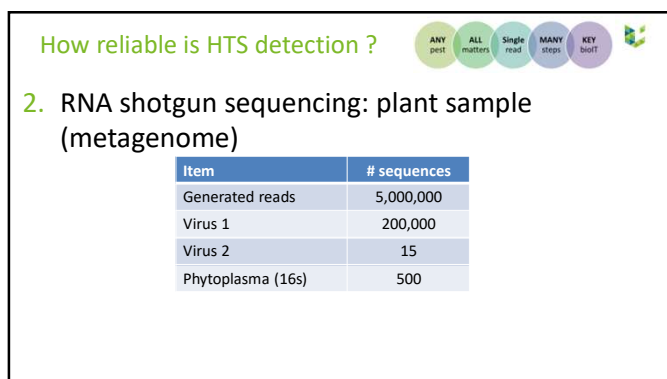
3



4



5



6

How reliable is HTS detection ?

ANY pest

2. RNA shotgun sequencing: plant sample (metagenome)

Item	# sequences
Generated reads	5,000,000
Virus 1	200,000
Virus 2	15
Phytoplasma (16s)	500

Looking beyond Virus Detection in RNA Sequencing Data: Lessons Learned from a Community-Based Effort to Detect Cellular Plant Pathogens and Pests

Amelien Hagerman¹, Yuka Kozumi², Xinyi Du Janghe¹, Thomas Condeelis³, Maher Al Bawabih^{2,4}, Neil Bonham⁵, Thierry Candresse^{6,7}, Yuya Z. A. Casati⁸, Oscar P. Hurtado-Gonzales^{9,10}, Zala Kogej Zwitter¹¹, Denis Kuznetsov¹², Jania Lamovska¹³, Marie Lefebvre¹⁴, Martha Malaga¹⁵, Irena Mavrot Pliska¹⁶, Serkan Oezler¹⁷, Jean-Sébastien Royward¹⁸, Ferran Salcedo Fanchinan¹⁹, Olivier Schupp²⁰, Kristian Stevens²¹, Chandan Pal²², Lucia Tamisier²³, Cigdem Ullabaz Serge²⁴, Inge van Duivenbode²⁵, David W. White^{26,27}, Xianjun Hu²⁸, Heiko Ziebell²⁹ and Sebastian Mansart³⁰

7

How reliable is HTS detection ?

ANY pest

8

How reliable is HTS detection ?

ANY pest

Considering the 37 selected datasets, in total 67 pathogens (39 different taxa) were detected (≥ 100 rpm) from 29 of the datasets (78%). Confirmation by PCR in >70% of the cases

Ok but really preliminary: many pending issues: what about sensitivity and selectivity ?

9

The hunting trophy bias in research

ALL matters

Virus discovery

First author PI

A new virus at high abundance (complete genome & high coverage)

Diagnostics (or virus ecology)

Other viruses/organisms ? If low abundance ?

10

How reliable is HTS detection ?

Single read

2. RNA shotgun sequencing: plant sample (metagenome)

Item	# sequences
Generated reads	5,000,000
Virus 1	200,000
Virus 2	15 ?
Phytoplasma (16s)	500 ?

DOUBT ? genomic databases (analytical specificity)
15 reads are enough ? (analytical sensitivity)

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How reliable is HTS detection ?

Single read

2. RNA shotgun sequencing: plant sample

Item	# sequences	# sequences - sample 2
Generated reads	5,000,000	5,000,000
Virus 1	200,000	0
Virus 2	15 ?	500,000
Phytoplasma (16s)	500	

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How reliable is HTS detection ?

MANY steps

**Bulletin OIEPP
EPPO Bulletin**

ORIGINAL ARTICLE | Free Access
Facilitating the adoption of high-throughput sequencing technologies as a plant pest diagnostic test in laboratories: A step-by-step description
 Benedicte Lebeau, Mehdi Al Rashidi, Steve Saeyen, Guillaume J. Biondini, Arnaud G. Blouin, Neil Broomham, Thierry Candresse, Anne Chaudrier, Kris De Jonghe, Asha Rai, Yanyu Z. A. Galet, Pascal Goffe, Anouk Heughebaert, Wouter Ho, Oscar Murcia-Gonzalez, Wilfried Jonkers, Jan Kovacs, Boris Kulnar, Blanca Landa, Mirjam Liu, François Maréchal, Martha Makai-Wright, Hana J. Morav, Françoise Nardoni, Natasa Nester, Angélique Minerva, Zuzanna Mielnic, Antonia Moreira, Mark Natcha, Françoise Peller, Alexander M. Pijper, Julien Pouchart, Bobba Raj, Benoit Romanat, Yuzmin Rivera, Brenden Rodon, Johannes W. Ruedenker, John Smith, Pasquale Salsenti, Juliana Santika, René Sauer-Richter, Davide Spadaro, David J. Stotholme, Stefanie Suhren, René van der Vlugt, Lucie Tansler, Charlotte Threlkoff, Hito Tazawa-Gottlieb, Claudio S. V. Vicente, Bart T. L. A. Wessenberg, Thierry Weis, Heiko Weis, Sebastian Weisheit | See fewer authors +

First published: 08 August 2022 | <https://doi.org/10.1111/app.12880>

13

How reliable is HTS detection ?

MANY steps

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First published: 08 August 2022 | <https://doi.org/10.1111/app.12880>

Error risk ↑
Process complexity →

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In results we trust: bioinformatics

- 12 viruses/viroids in 10 datasets
- Same data sent to 21 laboratories
- Each participant had to identify viruses present in the datasets

KEY bioIT

cost
EUROPEAN COOPERATION IN SCIENCE & TECHNOLOGY

Phytopathology - 2019 - 100-408-401 | <https://doi.org/10.1007/s12672-020-08-001-4>

bioRxiv e-Print archive

Virus Detection by High-Throughput Sequencing of Small RNAs: Large-Scale Performance Testing of Sequence Analysis Strategies
 Sébastien Mounier, Michel Choumert, Kim Do Jonghe, Rachel Ollery, Amelien Hegerman, Jan Kikvidze, Peter Krenn, Jan Kovacs, David Kozak, Caroline Lacroix, François Maréchal, Veronique Maréchal, René J. Natcha, Thibaut Ollery, Antonia Moreira, Mikael M. Pijper, Kees Scholten Krenn, Amy H. Ruiz Garcia, Dora Salazar, Pierre H. B. Schwabinger, Susi Saha, Shiva Suresh Kumar, Yanyu Z. A. Galet, Eric Vanhove, Eric Vanhove, Vincent Weisheit, Tracy Weisheit, and Thierry Candresse

15

In results we trust: bioinformatics

- Diagnostic sensitivity
 - 70%
 - Variable between labs
 - Depending on sequencing depth
- Bioinformatics really matters

KEY bioIT

labID	Sensitivity			Average
	50,000	250,000	2,500,000	
A	10%	53%	90%	51%
B	30%	20%	80%	46%
C	60%	71%	80%	70%
D	50%	82%	100%	78%
E	20%	82%	80%	68%
F	80%	88%	100%	89%
G	20%	53%	100%	57%
H	30%	65%	70%	57%
J	70%	98%	100%	89%
K	40%	71%	90%	68%
M	50%	98%	90%	81%
N	20%	82%	90%	70%
O	20%	41%	40%	35%
P	20%	59%	70%	51%
R	110%	100%	100%	100%
S	100%	100%	100%	86%
T	90%	100%	100%	97%
V	60%	88%	80%	78%
W1	40%	82%	90%	72%
W2	60%	82%	90%	78%
X	30%	71%	80%	62%
AVERAGE	46%	75%	86%	70%

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In results we trust: bioinformatics

Bioinformatic analysis was carried out
 « according to default parameters »

Will you do it in the laboratory ?

KEY bioIT

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In results we trust: bioinformatics

- « The holy trinity » from Roulev (XV century)

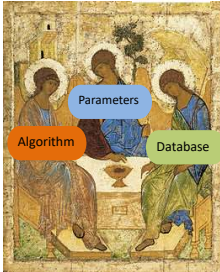
KEY bioIT

18

In results we trust: bioinformatics

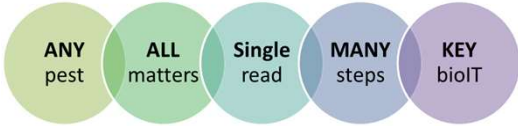
KEY bioIT

- > The « bioinformatics trinity »
- > Algorithms, their parameters and the database can create bias



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Summary of key elements



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LIÈGE université
Gembloux
Agro-Bio Tech


The contamination swamp

« Under the carpet » story



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Under the carpet the cross-contamination & the analytical sensitivity

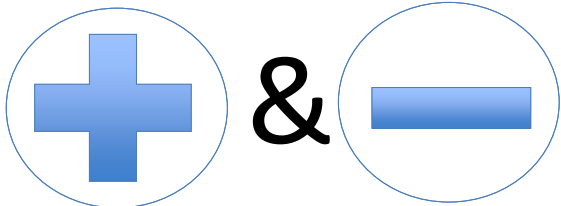


- > Analytical sensitivity
 - 10-100M reads
 - 10- 100 reads on target (1 ppm)
- > Low level infection or contamination ?

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4. Alien invasion

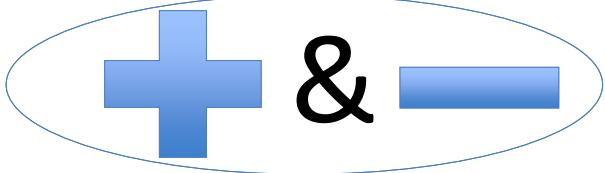
> About controls: before HTS



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4. Alien invasion

> With HTS, a positive control for a virus is also a negative control for any other virus



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Using an alien to monitor contamination

Library name	Batch	Reads	CMV	RT-PCR	HTS	CMV	RT-PCR	HTS
Alien control	1	1	1	1	1	1	1	1
110104	1	1	1	1	1	1	1	1
110105	1	1	1	1	1	1	1	1
110106	1	1	1	1	1	1	1	1
110107	1	1	1	1	1	1	1	1
110108	1	1	1	1	1	1	1	1
110109	1	1	1	1	1	1	1	1
110110	1	1	1	1	1	1	1	1
110111	1	1	1	1	1	1	1	1
110112	1	1	1	1	1	1	1	1
110113	1	1	1	1	1	1	1	1
110114	1	1	1	1	1	1	1	1
110115	1	1	1	1	1	1	1	1
110116	1	1	1	1	1	1	1	1
110117	1	1	1	1	1	1	1	1
110118	1	1	1	1	1	1	1	1
110119	1	1	1	1	1	1	1	1
110120	1	1	1	1	1	1	1	1
110121	1	1	1	1	1	1	1	1
110122	1	1	1	1	1	1	1	1
110123	1	1	1	1	1	1	1	1
110124	1	1	1	1	1	1	1	1
110125	1	1	1	1	1	1	1	1
110126	1	1	1	1	1	1	1	1
110127	1	1	1	1	1	1	1	1
110128	1	1	1	1	1	1	1	1
110129	1	1	1	1	1	1	1	1
110130	1	1	1	1	1	1	1	1
110131	1	1	1	1	1	1	1	1
110132	1	1	1	1	1	1	1	1
110133	1	1	1	1	1	1	1	1
110134	1	1	1	1	1	1	1	1
110135	1	1	1	1	1	1	1	1
110136	1	1	1	1	1	1	1	1
110137	1	1	1	1	1	1	1	1
110138	1	1	1	1	1	1	1	1
110139	1	1	1	1	1	1	1	1
110140	1	1	1	1	1	1	1	1
110141	1	1	1	1	1	1	1	1
110142	1	1	1	1	1	1	1	1
110143	1	1	1	1	1	1	1	1
110144	1	1	1	1	1	1	1	1
110145	1	1	1	1	1	1	1	1
110146	1	1	1	1	1	1	1	1
110147	1	1	1	1	1	1	1	1
110148	1	1	1	1	1	1	1	1
110149	1	1	1	1	1	1	1	1
110150	1	1	1	1	1	1	1	1
110151	1	1	1	1	1	1	1	1
110152	1	1	1	1	1	1	1	1
110153	1	1	1	1	1	1	1	1
110154	1	1	1	1	1	1	1	1
110155	1	1	1	1	1	1	1	1
110156	1	1	1	1	1	1	1	1
110157	1	1	1	1	1	1	1	1
110158	1	1	1	1	1	1	1	1
110159	1	1	1	1	1	1	1	1
110160	1	1	1	1	1	1	1	1
110161	1	1	1	1	1	1	1	1
110162	1	1	1	1	1	1	1	1
110163	1	1	1	1	1	1	1	1
110164	1	1	1	1	1	1	1	1
110165	1	1	1	1	1	1	1	1
110166	1	1	1	1	1	1	1	1
110167	1	1	1	1	1	1	1	1
110168	1	1	1	1	1	1	1	1
110169	1	1	1	1	1	1	1	1
110170	1	1	1	1	1	1	1	1
110171	1	1	1	1	1	1	1	1
110172	1	1	1	1	1	1	1	1
110173	1	1	1	1	1	1	1	1
110174	1	1	1	1	1	1	1	1
110175	1	1	1	1	1	1	1	1
110176	1	1	1	1	1	1	1	1
110177	1	1	1	1	1	1	1	1
110178	1	1	1	1	1	1	1	1
110179	1	1	1	1	1	1	1	1
110180	1	1	1	1	1	1	1	1
110181	1	1	1	1	1	1	1	1
110182	1	1	1	1	1	1	1	1
110183	1	1	1	1	1	1	1	1
110184	1	1	1	1	1	1	1	1
110185	1	1	1	1	1	1	1	1
110186	1	1	1	1	1	1	1	1
110187	1	1	1	1	1	1	1	1
110188	1	1	1	1	1	1	1	1
110189	1	1	1	1	1	1	1	1
110190	1	1	1	1	1	1	1	1
110191	1	1	1	1	1	1	1	1
110192	1	1	1	1	1	1	1	1
110193	1	1	1	1	1	1	1	1
110194	1	1	1	1	1	1	1	1
110195	1	1	1	1	1	1	1	1
110196	1	1	1	1	1	1	1	1
110197	1	1	1	1	1	1	1	1
110198	1	1	1	1	1	1	1	1
110199	1	1	1	1	1	1	1	1
110200	1	1	1	1	1	1	1	1

- CMV : infected or false positive ?
- RT-PCR & new HTS
- Check on other projects
- CMV+ samples (>500k reads)
- 100% identity

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So, contaminations are

Impossible to avoid

Unpredictable

Monitoring is needed (preferably by alien control)

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Fine, I monitor it but... which threshold to apply ?

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Fine, I monitor it but... which threshold to apply ?

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Fine, I monitor it but... which threshold to apply ?

The grey zone

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Fine, I monitor it but... which threshold to apply ?

How can I reduce my grey zone while maintaining reliability ?

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Fine, I monitor it but... which threshold to apply ?

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Fine, I monitor it but... which threshold to apply ?

rpm

1. A fixed absolute value
2. A fixed relative value (%) to the maximum in a sample

plants **biology**

Looking beyond Virus Detection in RNA Sequencing Data: Lessons Learned from a Community-Based Effort to Detect Cellular Plant Pathogens and Pests

Side-by-Side Comparison of Post-Entry Quarantine and High Throughput Sequencing Methods for Virus and Viroid Diagnosis

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Fine, I monitor it but... which threshold to apply ?

- Adaptative depending on run's contamination monitored by Alien control
- 3 levels evaluated :
 - no filter (fixed)
 - 10 reads filter (fixed)
 - Maximum of alien contamination in a sample: adaptative between batches: from 7 to 288

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Fine, I monitor it but... which threshold to apply ?

- Analytical sensitivity

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Fine, I monitor it but... which threshold to apply ?

- Analytical sensitivity -> accuracy

Sample	Filter	Accuracy	False positive	False negative
Batch 1	No filter	49%	51%	0%
	10 reads	93%	5%	2%
Batch 2	No filter	93%	5%	2%
	10 reads	29%	71%	0%
Batch 3	No filter	94%	5%	3%
	10 reads	43%	57%	0%
Batch 4 (with dilution)	No filter	86%	11%	2%
	10 reads	28%	3%	9%
Batch 5	No filter	80%	20%	0%
	10 reads	90%	10%	0%

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Fine, I monitor it but... which threshold to apply ?

- Absolute number of reads or rpm ?
- BBTV: low abundance virus (380 reads on average and maximum 2,397 reads per sample vs. 25,000 to 120,000 reads for alien virus)
- Batch with 282 alien contaminating reads -> many false negative for BBTV
- Relative value is needed

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Using an alien to monitor contamination

Cont-ID pipeline: automatic analysis with expert interpretation

SOFTWARE Open Access

Cont-ID: detection of sample cross-contamination in viral metagenomic data

Johan Rollin^{1,2*}, Wei Hong¹ and Sébastien Mascart^{1,2*}

BMC Biology

Bollu et al. BMC Biology (2023) 21:2177
https://doi.org/10.1186/s12915-023-01788-w

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Using an alien to monitor contamination

Parameter for Alien – background quantification (ELISA-like)

Normalization (Using Total read nb)

$$NB_{m,n} = \frac{\text{Average } (nb \text{ reads mapped}) \text{ for alien contamination (indexing = Subject)}}{(NB_{A1} + NB_{A2} + \dots + NB_{Am}) + (3 \times sd)}$$

Control 1

Cont-ID: detection of sample cross-contamination in viral metagenomic data

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Using an alien to monitor contamination

Parameter for Alien – number in Alien control

Control 2

$$\text{Max } (nb \text{ reads mapped}) \text{ for alien virus in alien control} \\ \text{max}(NB_{A1}, NB_{A2}, \dots, NB_{Am})$$

Cont-ID: detection of sample cross-contamination in viral metagenomic data

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Using an alien to monitor contamination

Parameter for Alien – number in Alien control

Control 3

$$\text{Average } (nb \text{ reads mapped}) \text{ for alien} \\ (D_{A1} + D_{A2} + \dots + D_{Am})$$

Cont-ID: detection of sample cross-contamination in viral metagenomic data

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Using an alien to monitor contamination

Reads mapped on each target (virus) in each sample:

RATIO

Mapping ratio

$$\frac{nb \text{ reads mapped on a virus } m \text{ relatively to the most abundant sample for that virus}}{NB_{m,n}}$$

Normalization (Using Total read nb)

$$NB_{m,n} = \frac{\text{Average } (nb \text{ reads mapped}) \text{ for alien contamination (indexing = Subject)}}{(NB_{A1} + NB_{A2} + \dots + NB_{Am}) + (3 \times sd)}$$

Control 1

Cont-ID: detection of sample cross-contamination in viral metagenomic data

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Using an alien to monitor contamination

Reads mapped on each target (virus) in each sample:

NUMBER

NB read

$$\frac{nb \text{ reads mapped on } m \text{ virus}}{NB_{m,n}}$$

Control 2

$$\text{Max } (nb \text{ reads mapped}) \text{ for alien virus in alien control} \\ \text{max}(NB_{A1}, NB_{A2}, \dots, NB_{Am})$$

Cont-ID: detection of sample cross-contamination in viral metagenomic data

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Using an alien to monitor contamination

Cont-ID: detection of sample cross-contamination in viral metagenomic data

Reads mapped on each target (virus) in each sample:
DEDUPLICATION

Deduplication ratio
Correspond to Deduplication₃ (no of reads identical to reads from the most abundant sample for that virus) / Z_{max}

Data Metrics

For virus n in sample m

Control 3
Average (Deduplication₃) for alien
($D_{3m} = \sum_{i=1}^n D_{3m,i} / n$)

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Using an alien to monitor contamination

Cont-ID: detection of sample cross-contamination in viral metagenomic data

Rule 1:
Mapping_ratio > (Control_1 / X)

Rule 2:
NB_read > (Control_2 / Y)

Rule 3:
Deduplication_ratio <= (Control_3 / Z)

Adaptability Metrics

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But still a long way to go... April 2022

ARTICLE
The pollen virome of wild plants and its association with variation in floral traits and land use

Supplementary material

Virus family ^a	Virus genus ^a	Known virus	Region ^b	Plant species ^c	No. segments identified ^d	Percent sequence coverage ^e	No. alignment ^f	NCBI accession nos.
Bromoviridae	Brome mosaic virus	EDM1	CA	Albizia leonardum	25	25.1	22-28	NC_032029.1
				Sida acuta	33	17.81 - 32.81	16 - 20	NC_060307.1, NC_052038.2
Columbicovirus	Peanut rosette virus	EDM1	CA	Vernonia pycnantha	19	31.91	18	NC_052048.1
				Impatiens capensis	23	8.88 - 22.11	2889 - 2924	NC_052127.1, NC_052128.1
Asteraceae	Aster virus	EDM1	CA	Phlox paniculata	33	8.68 - 28.76	88 - 112	NC_052129.1
				Phlox paniculata	23	20.81 - 61.58	56 - 70	NC_052481.1, NC_052482.1
Rhabdoviridae	Impatiens virus	EDM1	CA	Convolvulaceae sp.	23	20.24 - 63.33	48 - 72	NC_051633.1
				Impatiens capensis	18	42 - 106.11	1781 - 2000	NC_051634.1
Rhabdoviridae	Impatiens virus	EDM1	CA	Convolvulaceae sp.	23	20.49 - 79.52	168 - 248	NC_051634.1
				Impatiens capensis	22	20.49 - 62.51	72 - 247	NC_051634.1
Rhabdoviridae	Impatiens virus	EDM1	CA	Subsp. sp.	25	15.39 - 71.89	47 - 216	NC_051634.1
				Impatiens capensis	23	20.24 - 57.39	47 - 216	NC_051634.1
Rhabdoviridae	Impatiens virus	EDM1	CA	Impatiens capensis	23	8.18 - 22.69	42 - 88	NC_052481.1, NC_052482.1

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Investigating contamination from Impatiens

Genome identity between samples ?

RNA 1	KX834010 Impatiens	KX834010 Oenothera	KX834010 Lotus	Consensus	KX834010 Solidago
KX834010_BCRV_contig_6 Impatiens (reversed)		0	100	100	99
KX834010_BCRV_contig_722 Oenothera (reversed)	0		100	100	0
KX834010_BCRV_contig_154 Lotus (reversed)	100	100		100	100
Consensus	100	100	100		100
KX834010_BCRV_contig_1293 Solidago (reverse-d)	99	0	100	100	

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Investigating contamination from Impatiens

SNP analysis ?

Lotus RNA1 of BCRV has 20 mutations absent in Impatiens

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Investigating contamination from Impatiens

SNP analysis ?

All the SNPs in Oenothera (73), Convolvus (9) and Vernonia (7) are present in Impatiens

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Impact on regulation ?

SCIENTIFIC OPINION

ADOPTED: 21 November 2020
doi: 10.2903/j.efsa.2020.5928

Pest categorisation of non-EU viruses of *Rubus L.*

EFSA Panel on Plant Health (PH),
Claude Bragard, Katharina Dehnen-Schmütz, Paolo González, Marie-Agnès Jacques,
Josep Anton Jaques Miret, Annemarie Fejer Justesen, Alan MacLeod, Christer Sven
Magnusson, Panagiotis Milonas, Juan A Navas-Cortes, Stephen Parnelli, Rolf Potting,
Philippe Lucien Raouf, Hans Herrmann Thulke, Wopke Van der Werf, Antonio Vicent Civera,
Jonathan Yuen, Lucia Zappalà, Thierry Candresse, Elisabet Chaitovskisliu, Franco Freni,
Stephan Winter, Domenico Bosco, Michela Chilumetti, Francesco Di Serio, Franco Ferrali,
Tomaz Kraljic, Angelantonio Minifra and Luisa Rubinio

For BCRV, due to the insufficient information available the Panel was unable to conclude on the potential consequences in the EU territory.

However, meet all the other criteria evaluated by EFSA to qualify as Union quarantine pests.

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Agro-Bio Tech

The path toward diagnostics use

Guidelines for reliable HTS diagnostics

56

Valitest project: guidelines for any pest

- Writing guidelines for using HTS in diagnostics setting
- Two scientific publications and one EPPO standard

55 co-authors

>1,500 revisions

Any pest

Any technology

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Valitest project: a follow-up for any pest

- First publication: building the foundations

Facilitating the adoption of high-throughput sequencing technologies as a plant pest diagnostic test in laboratories: A step-by-step description

Bulletin OEPP
EPP0 Bulletin

Original article: 08 August 2022 | <https://doi.org/10.1111/epp.12868>

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Valitest project: a follow-up for any pest

- Running it: from adoption to validation

Peer Community Journal
Section: Infections

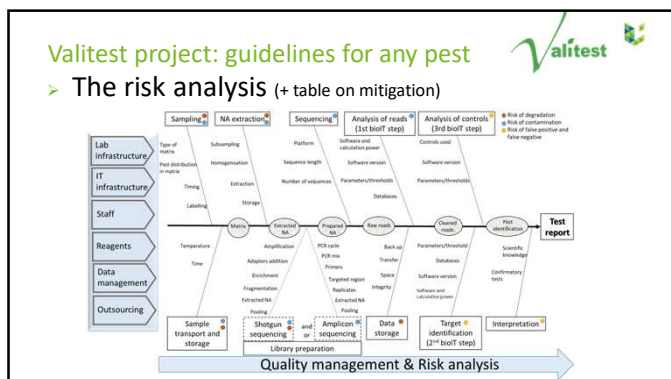
Guidelines for the reliable use of high throughput sequencing technologies to detect plant pathogens and pests

16 authors: Stéphane Huet, Ian Adams, Maher Al Bakhrah, Steve Bacon, Guillaume Bédouin, Armand S. Boinin, Neil Bounieau, Thierry Candresse, Anne-Claire de Jonghe, Adnan H. Yaqub, Z. S. Gader, Anouk Gauthier, Amel Hachimi, Willem Ho, Oscar Hurtado-Gonzales, Wilfried Jankari, Jan Knežek, Denis Kudva, Shweta Laksh, Meng Lu, François Leclerc, Steffi Altmann-Wright, Sami Luoma, Francesco Mariani, Natasa Mihic, Angelantonio Minifra, Dimitre Motov, Adriana Mónica, Mark Natchez, Françoise Petit, Alexander M. Pijet, Jean Potbury, Balázs Rákos, Bernd Reinhardt, Samira Sereika, Brenton Roberts, Johannes W. Rumberg, John Rubin, Priscilla Sathiraj, Johannes Sautter, Rose Soule-Richard, Davide Spadaro, David J. Stachowicz, Stefanie Summers, René van der Vlugt, Lucie Tamber, Charlotte Tronin, Erik Vazquez-Sunyer, Claudio S. Vitoria, Bert T. L. H. Vossenberg, Thierry Yoccoz, Hedio Zedler, Sebastian Zvereva

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Valitest project: guidelines for any pest

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Adaptation toward official standard

DOI: 10.1111/epi.12884

EPPO STANDARD ON DIAGNOSTICS

PM 7/151 (1) Considerations for the use of high throughput sequencing in plant health diagnostics¹

Specific scope: This Standard describes elements to take into consideration for the use of high throughput sequencing (HTS) tests, including validation, quality control measures and interpretation and reporting of results to ensure HTS test results are robust and accurate, have biological significance in a phytosanitary context, and are implemented in a harmonized way. This Standard applies to all plant pest groups and HTS technologies. This Standard should be used in conjunction with PM 7/76 Use of EPPO diagnostic protocols. Specific approval and amendment: Approved in 2022–09. Authors and contributors are given in the Acknowledgements section.

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Conclusion: technical challenges

> Super hype and high potential but ...

« All that glitters is not gold »

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Conclusion: technical challenges

> Each step can create bias

> Double check results of any publication

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Conclusion: my experience

... the supplementary material of any publication !

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Conclusion: toward larger scale application

- > More complex process than previous tests
- > Guidelines and official standard

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Conclusion : toward larger scale application

- > New pests (viruses) uncharacterized
- > New hosts (and pathways)
- > New countries
- > Biology and risk management as rising epicenters of HTS earthquake...

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Thanks for the support

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A team work in the lab

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Special thanks to hundreds of plant pathologists collaborating on disseminating HTS and making it more reliable

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Thanks for the invitation & for your attention

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 @Be_Phytopath

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