

# ***Berry class synchronization: a prospective to understand berry ripening ?***



***Laurent Deluc***

***Horticulture Department (OSU) - August 25th 2011***

Oregon *Wine* Research Institute



**Two questions:**

**Do we have uniform clusters at harvest ?**



**How does it work?**



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**Do we have uniform clusters at harvest ?**

*If not it may contribute to the wine balance*



**How does it work?**



## Two questions:

**Do we have uniform clusters at harvest ?**

*If not it may contribute to the wine balance*



**How does it work?**

*it could be useful to improve fruit quality*



# Consequences of ripeness heterogeneity at harvest

- Wine Quality :

- presence of a non-negligible proportion of unripe berries can increase the appearance of bitter and astringent characters in wine and undesirable flavors



# Contributions of different berry classes to the overall quality

Harvest date based on ripening rate of berry class # 1

Harvest date based on ripening rate of berry class #1

Berry Class

1: 100%

5: 100%

2: 90%

4,6: 92%

3: 80%

3,7: 82%

4: 67%

2,8: 68%

5: 55%

1,9: 55%

6: 43%

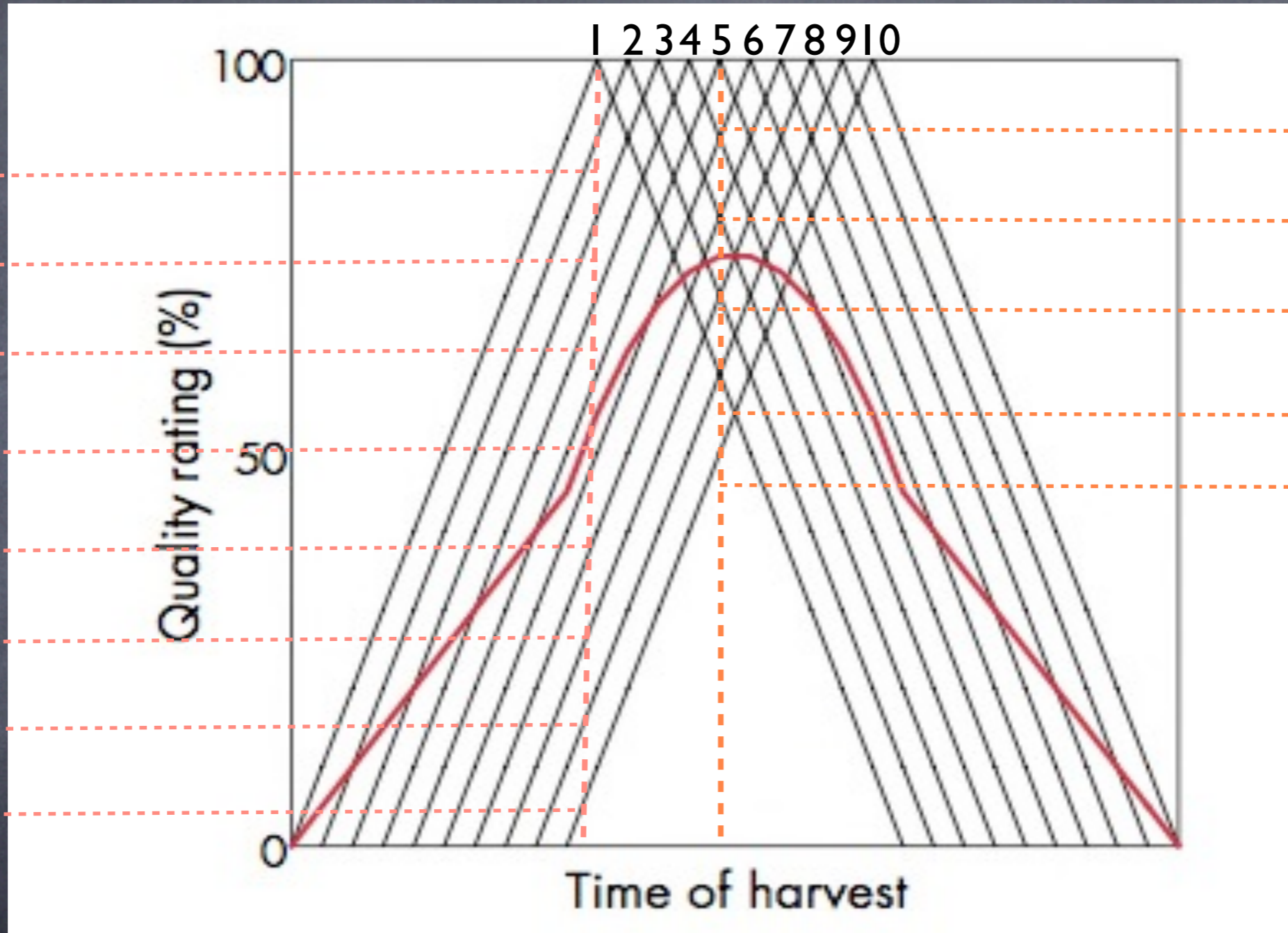
10: 50%

7: 35%

8: 20%

9: 10%

10: 0%



Overall Quality = 50%

Gray 2006

Overall Quality = 74%



# Consequences of ripeness heterogeneity at harvest

- Wine Balance :

- presence of a non-negligible proportion of unripe berries can increase the appearance of bitter and astringent characters in wine and undesirable flavors

- Heterogeneity in phenolic maturity observed between different berry classes at harvest in Cabernet Sauvignon (Kontoudakis et al., 2011):

- Higher sugar content is correlated to higher anthocyanin, total phenolic index, proanthocyanidin concentrations

- Few data are available...



# Ripeness heterogeneity among berries may depend upon the genetic background



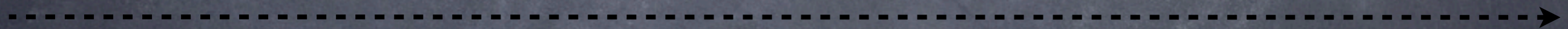
“Pinot Noir Berries”



“Zinfandel, Cabernet Sauvignon, Shiraz”



“Merlot Berries”



Low

Moderate

Extreme

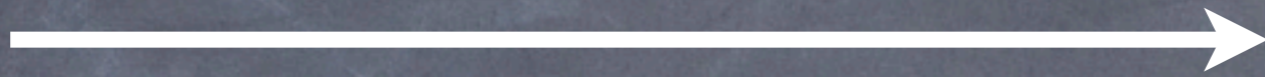




# Is Pinot Noir able to alleviate berry variability at harvest?



?

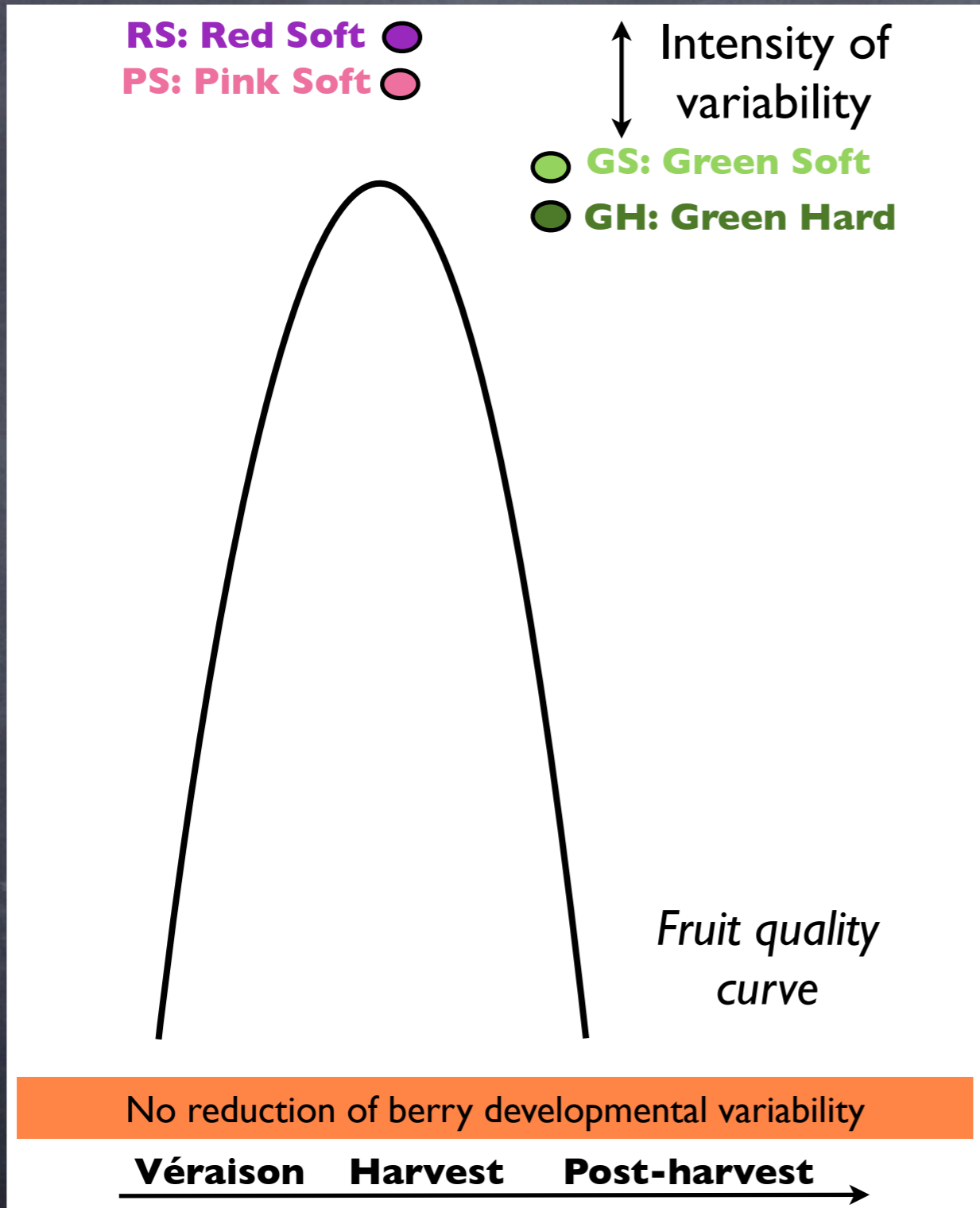


***Are under ripe berries at mid-*véraison* able to catch up more advanced berries and how?***

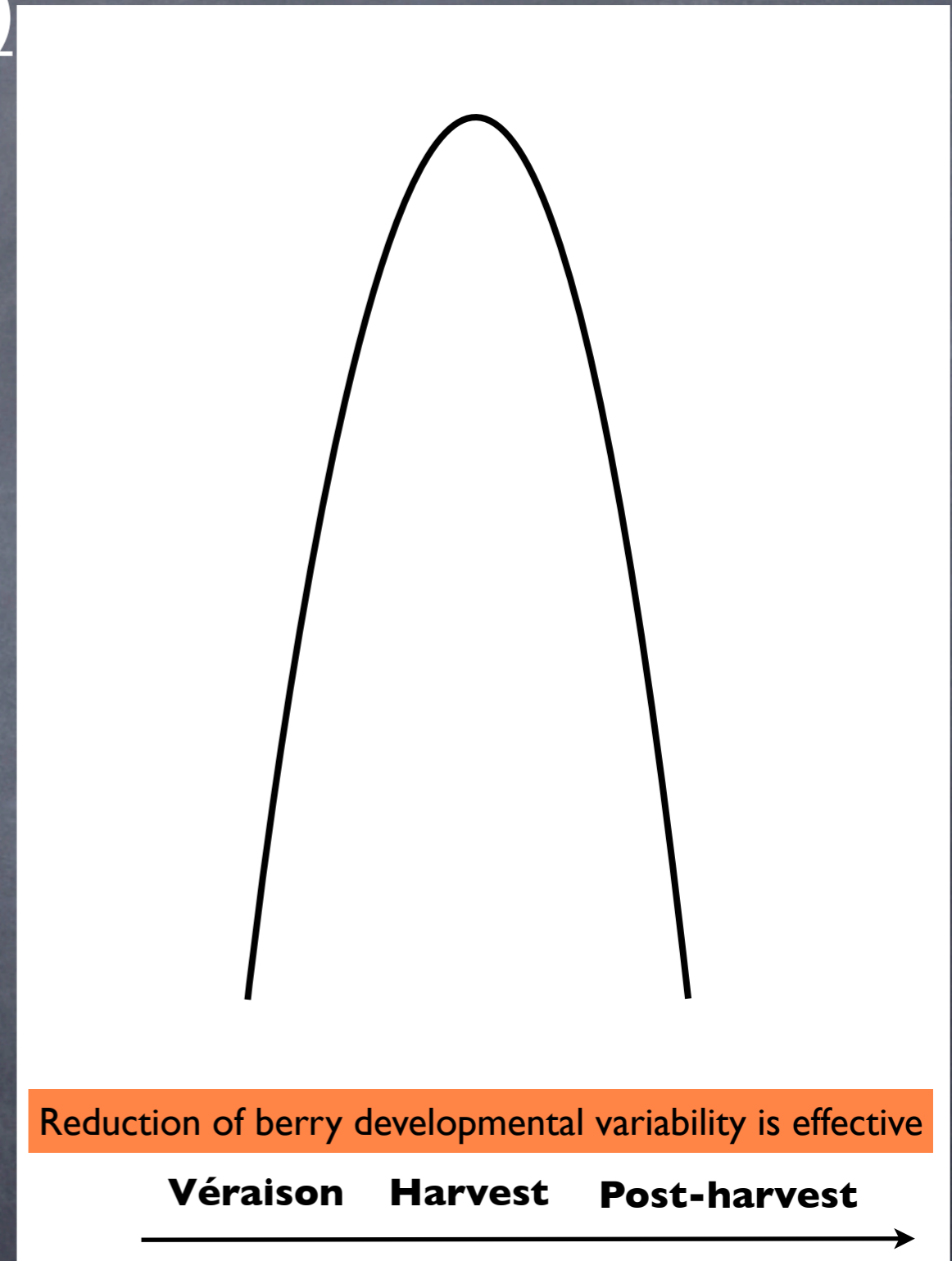


# Reduction of berry class variability

1)



2)



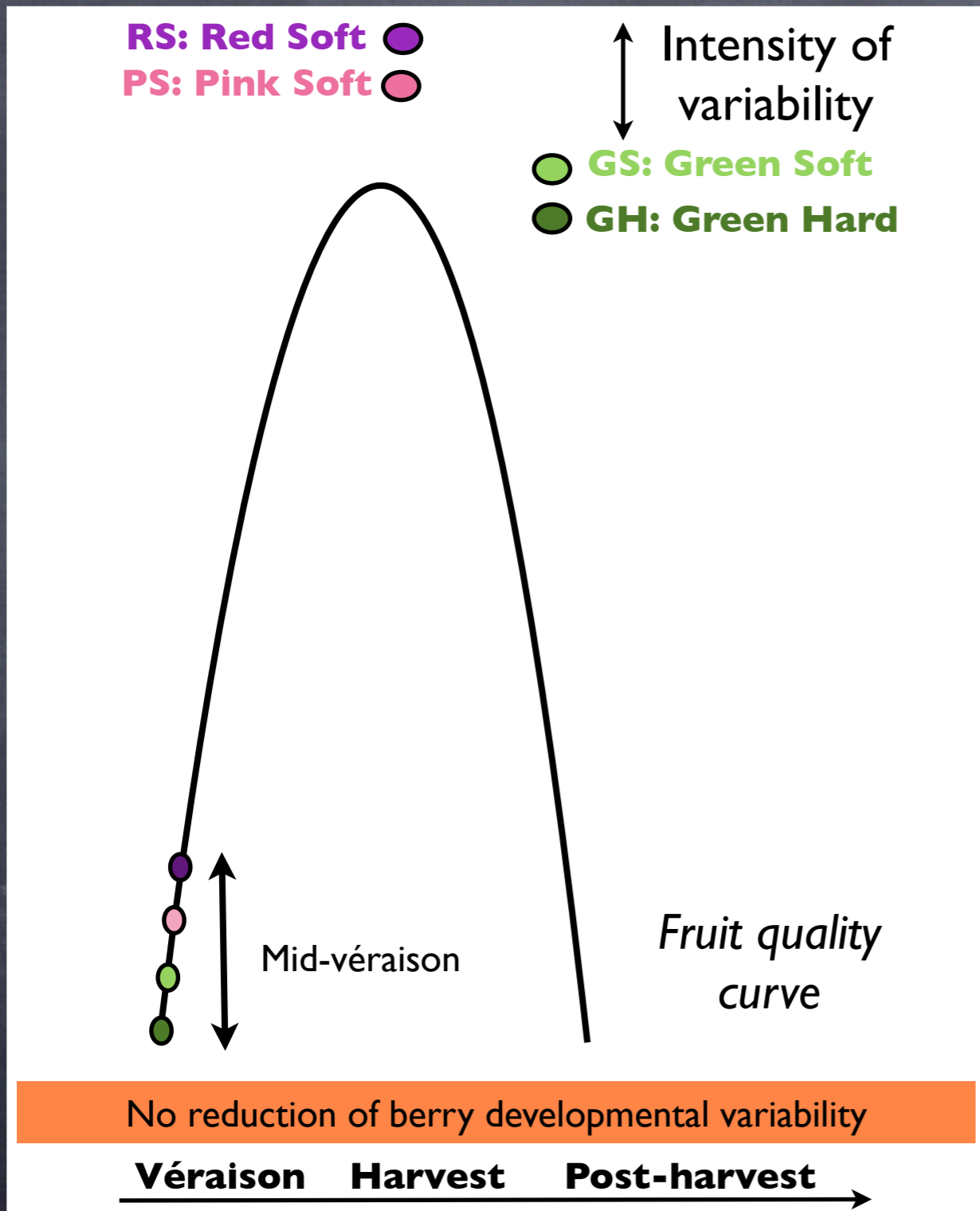
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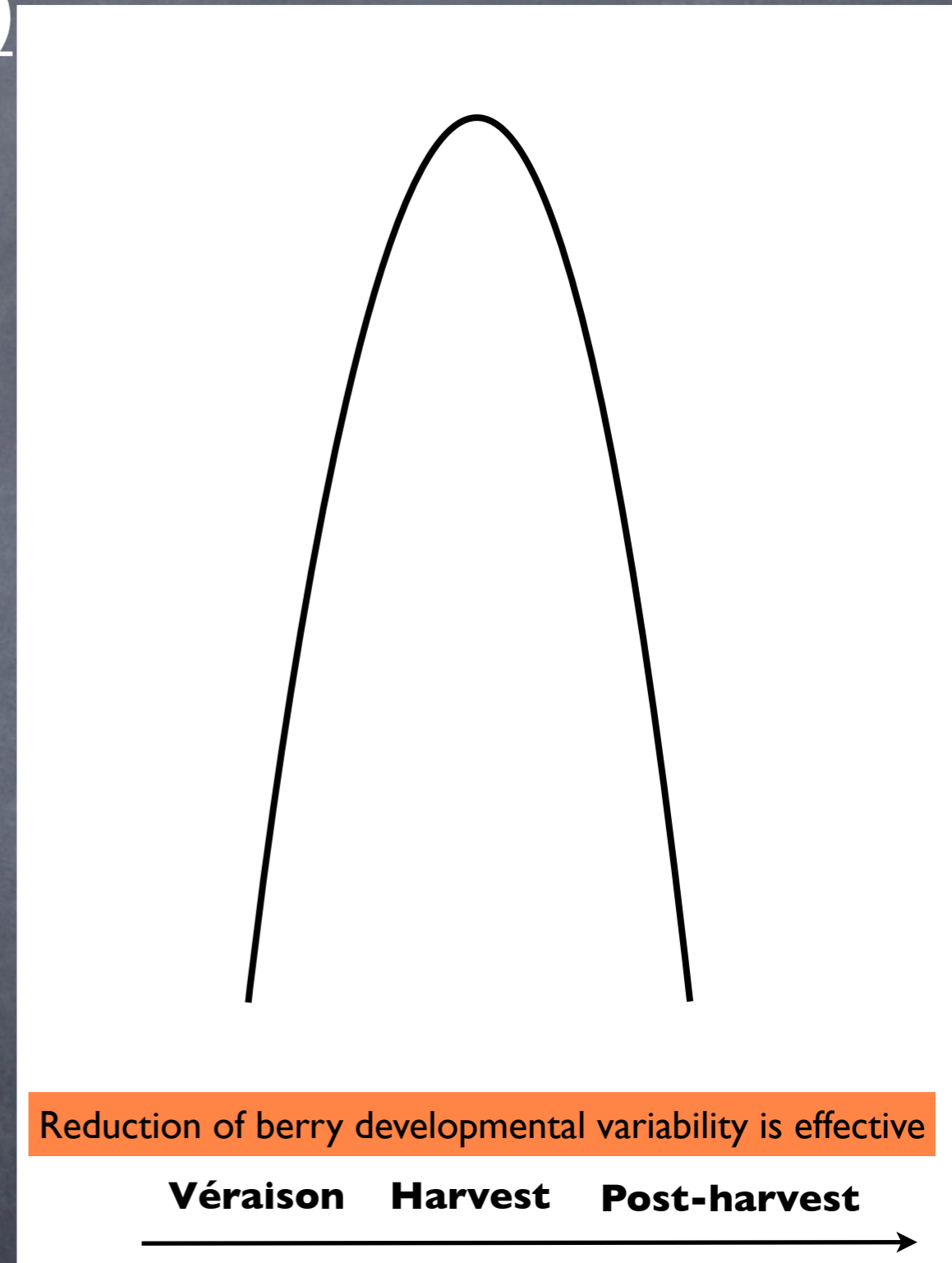


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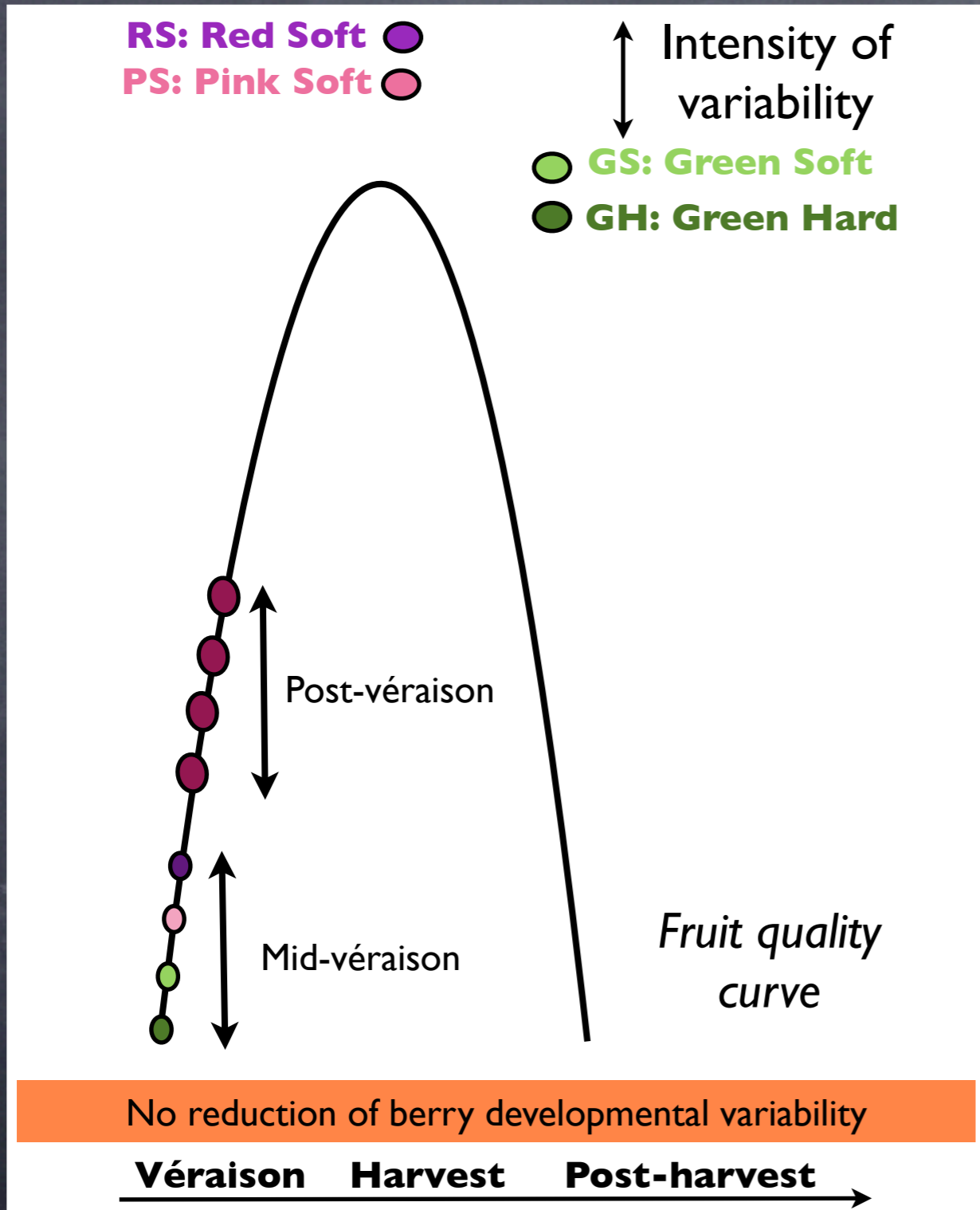
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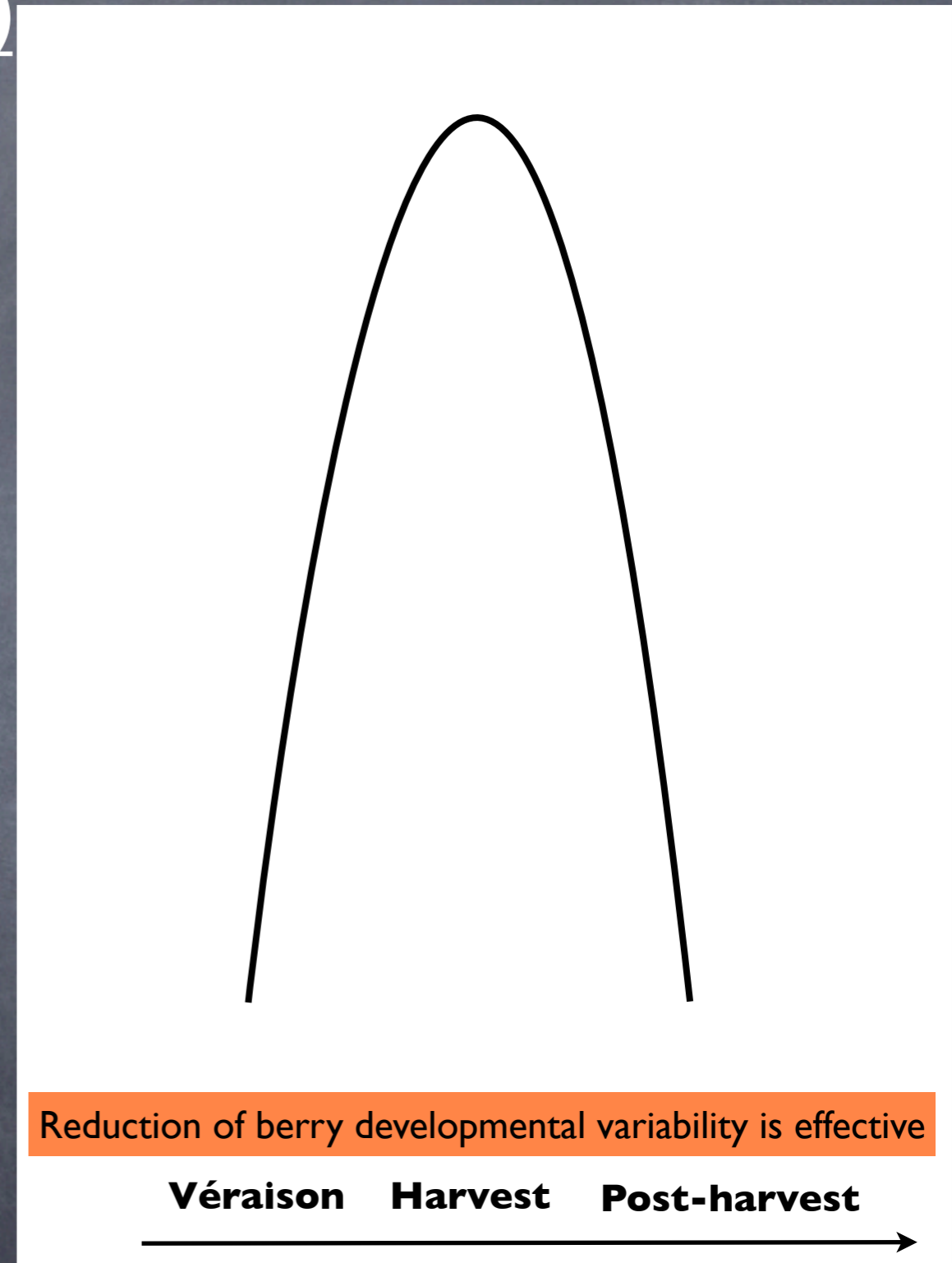


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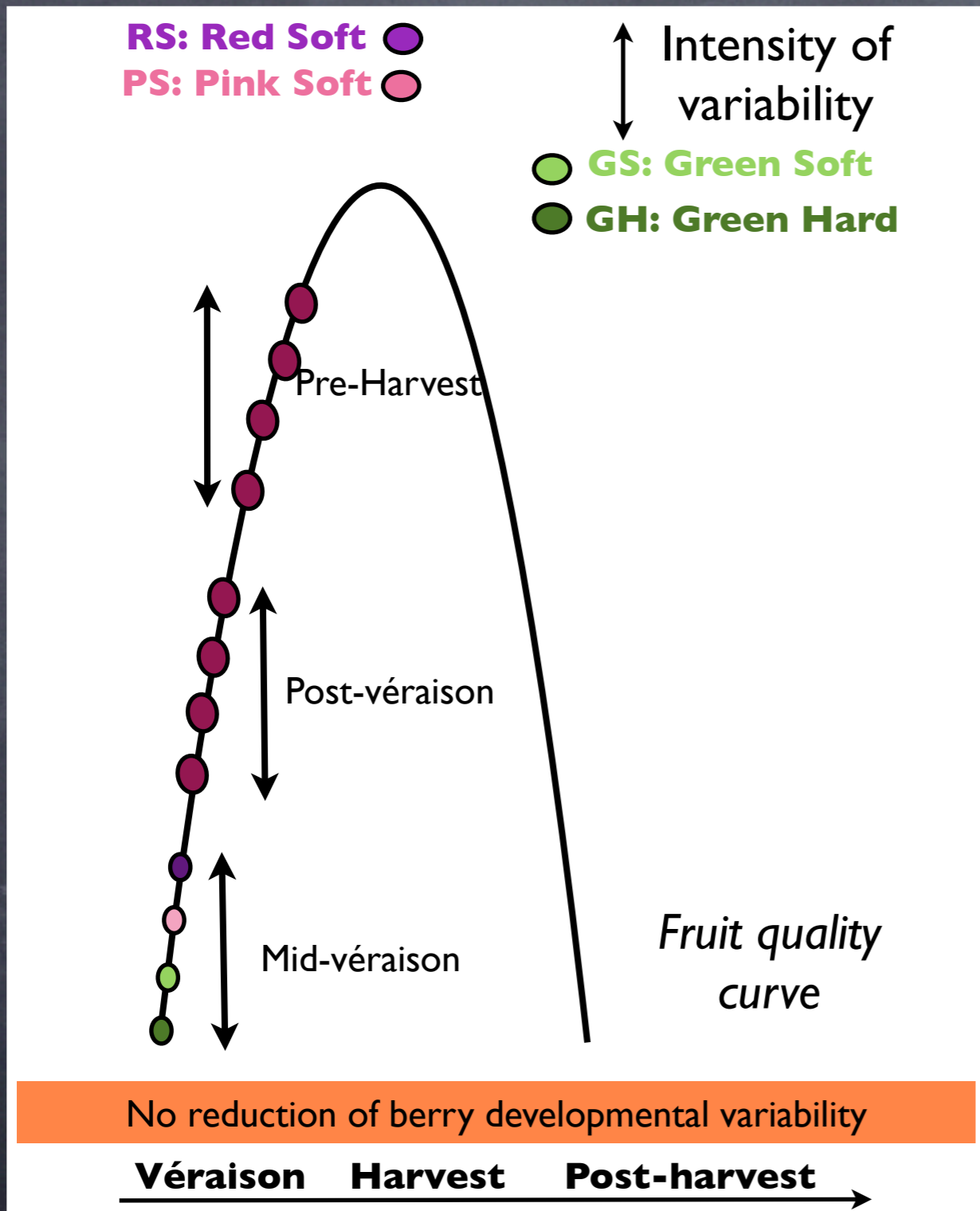
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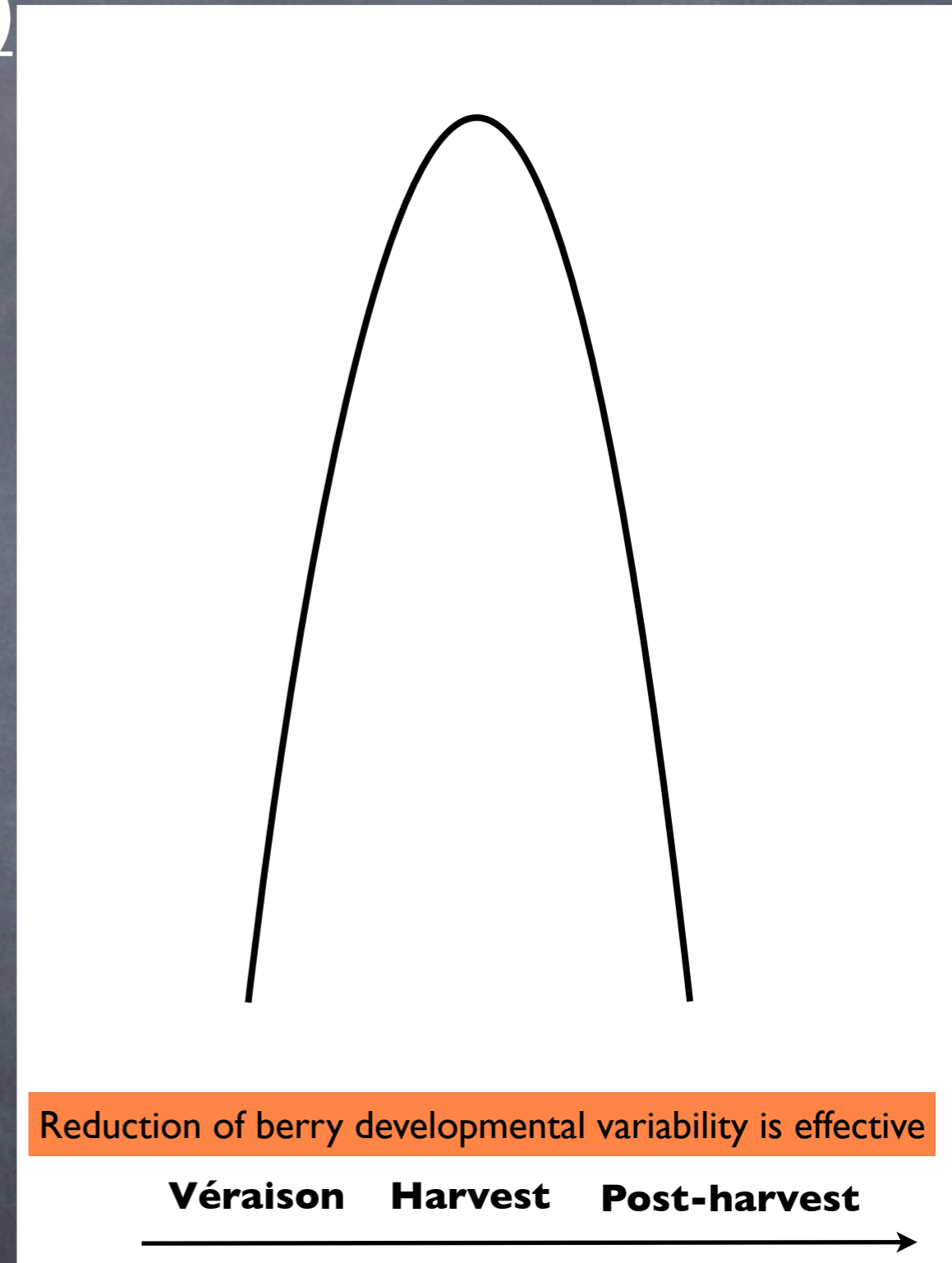


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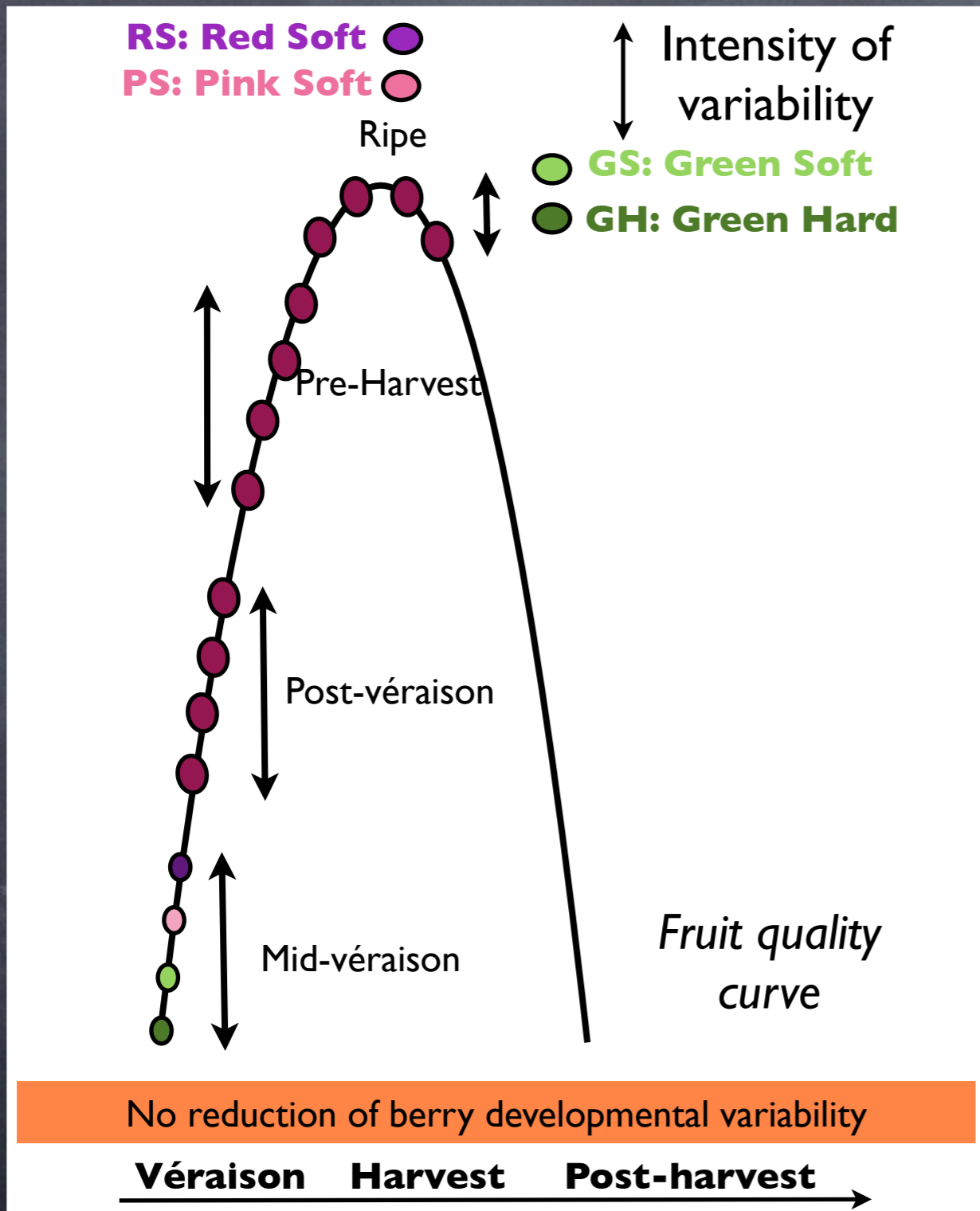
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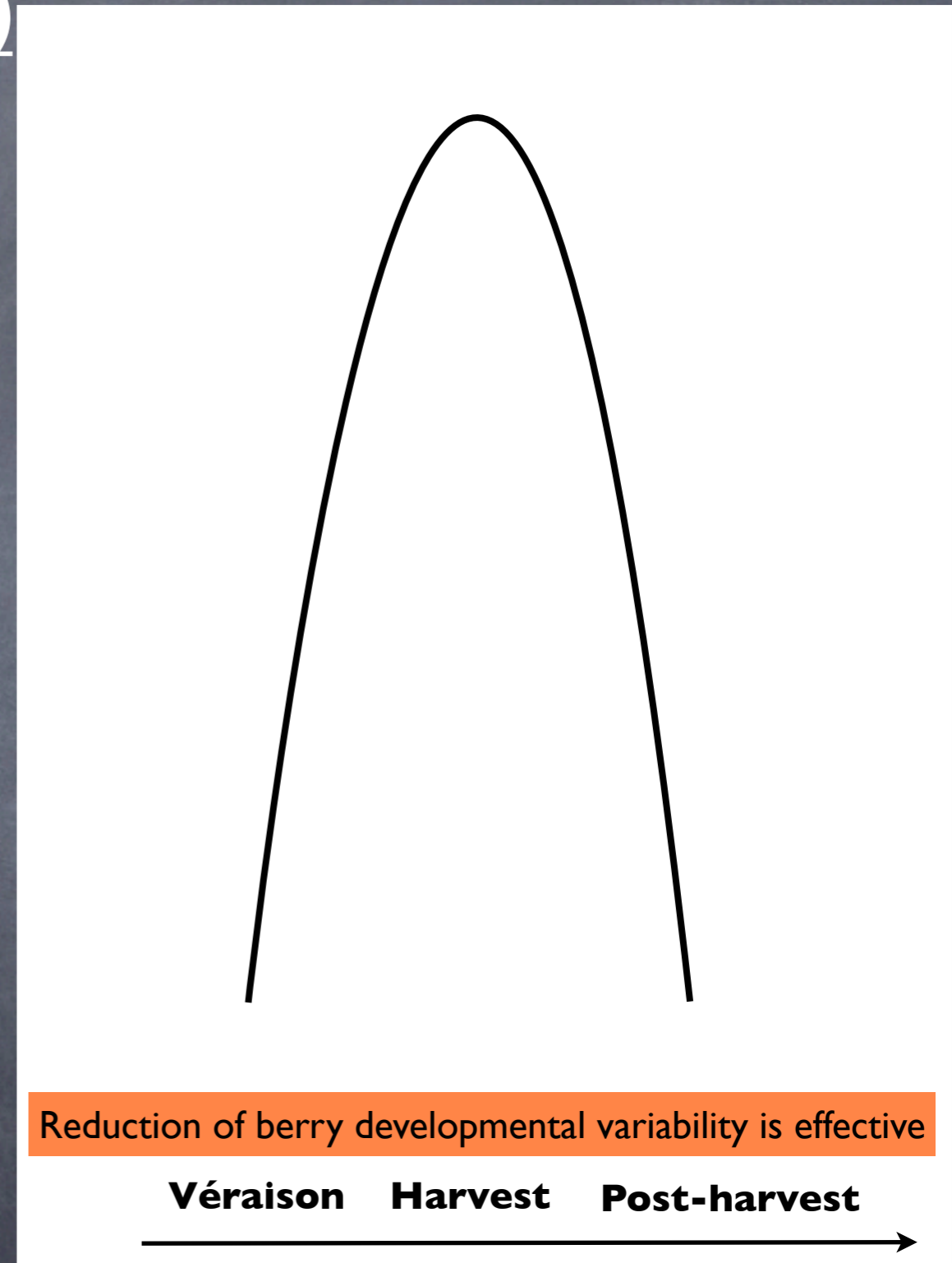


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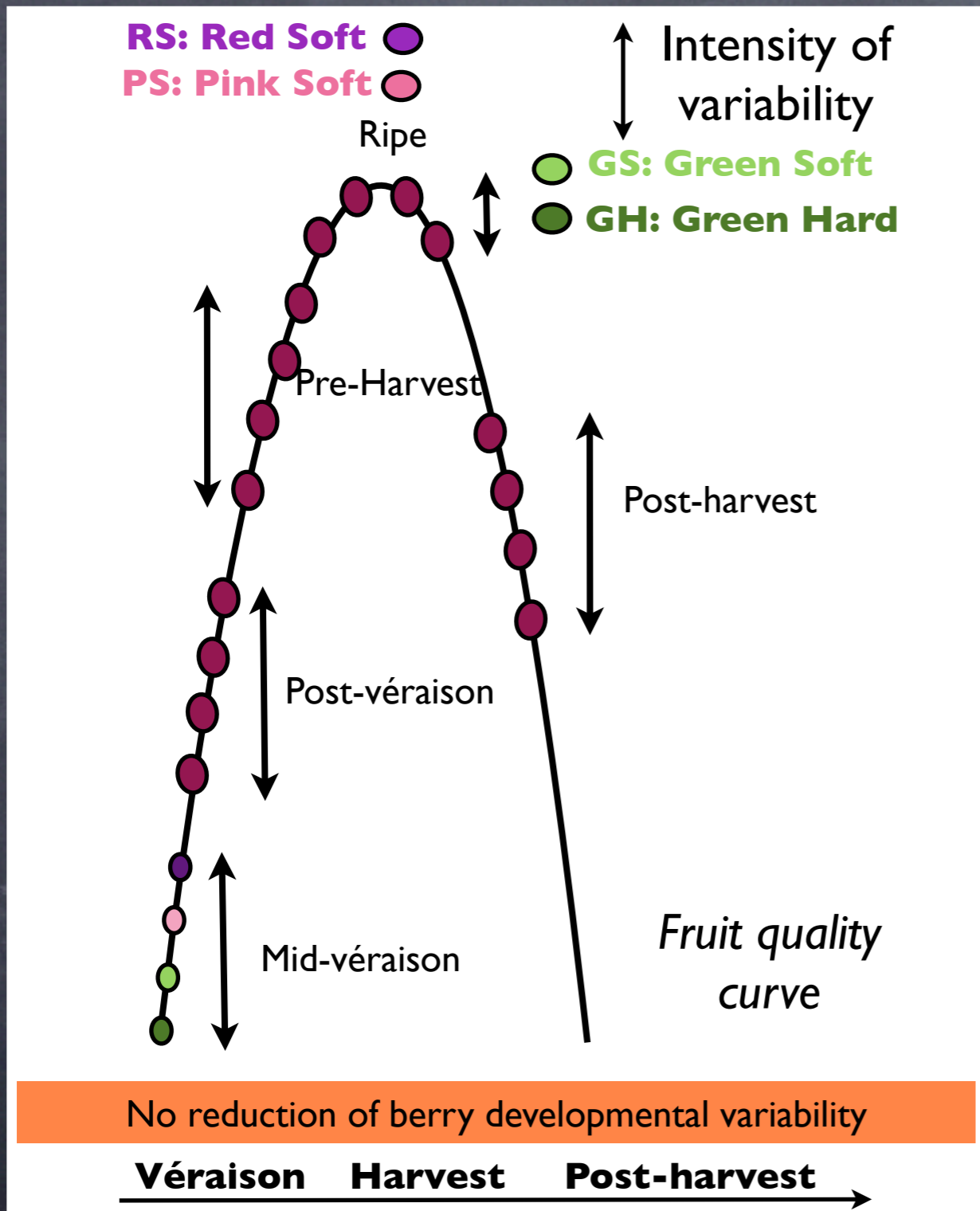
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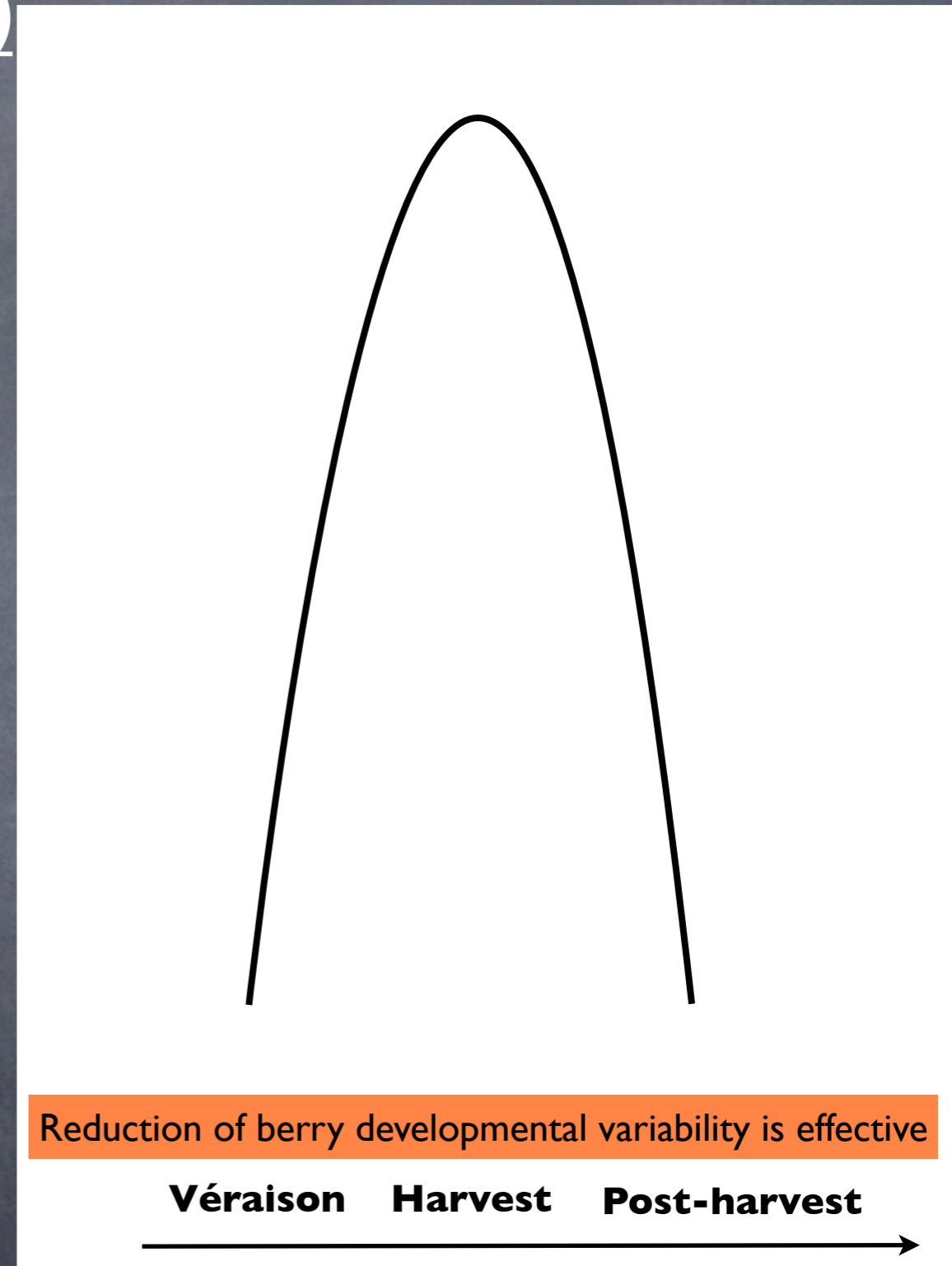


# Reduction of berry class variability

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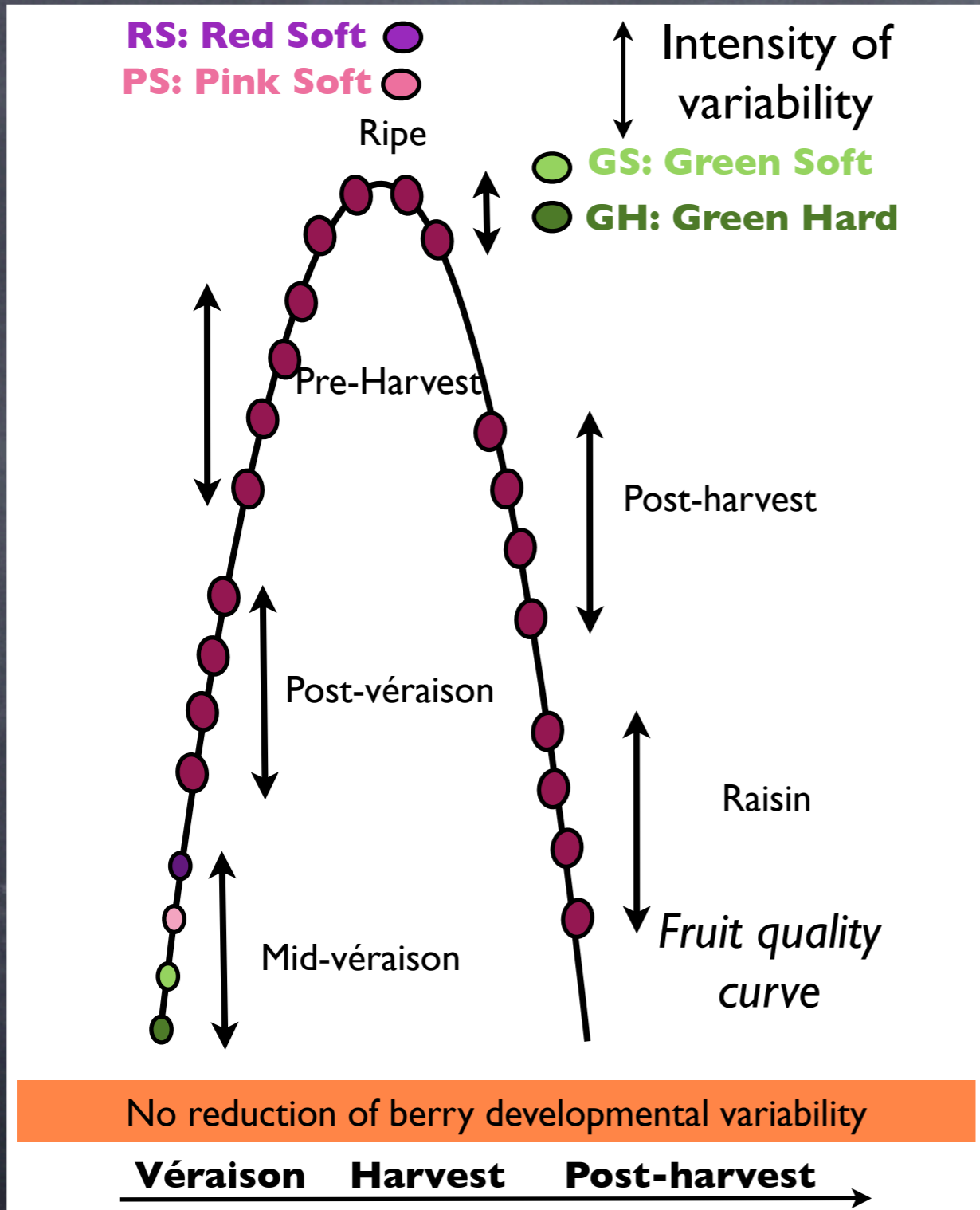
2)



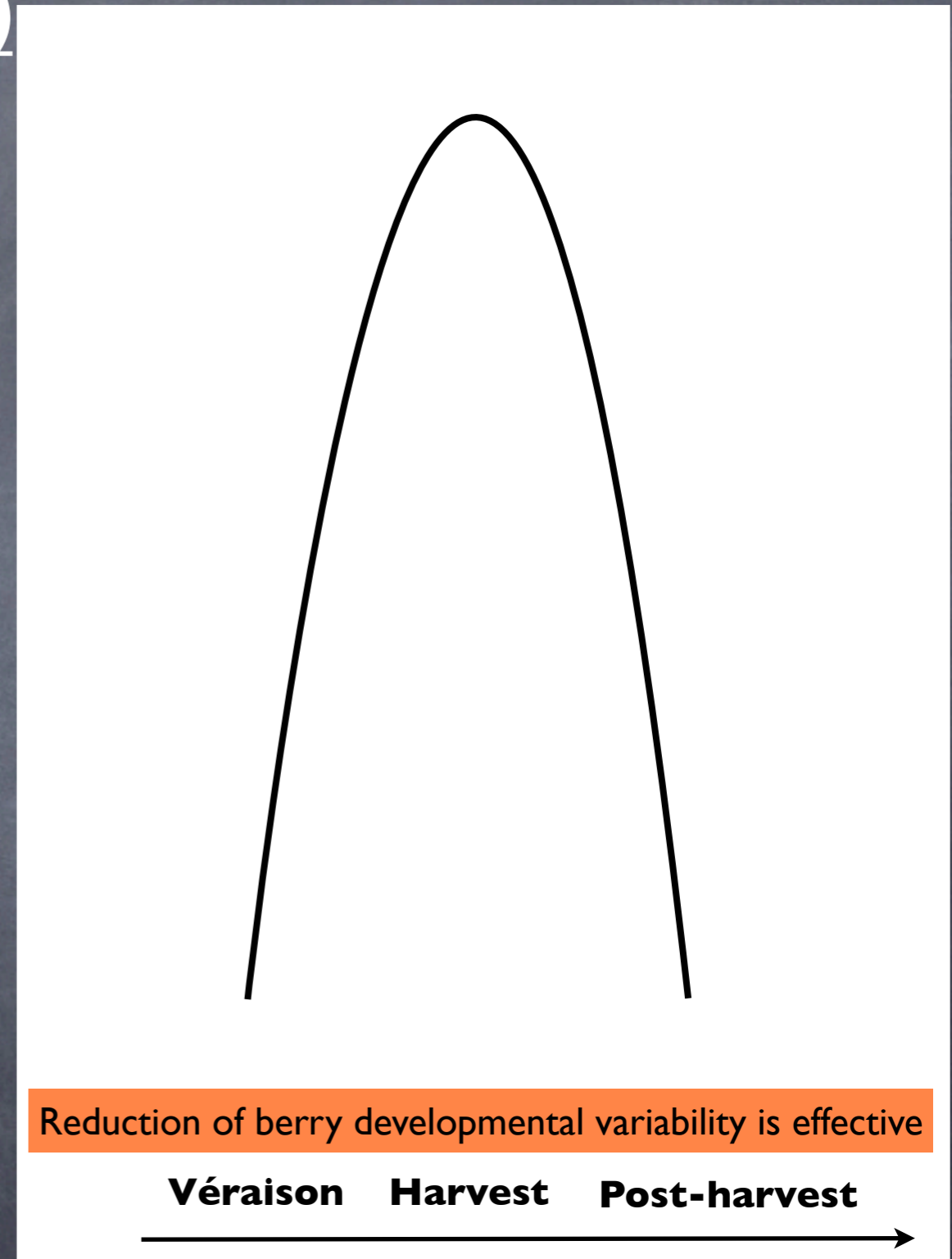
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# Reduction of berry class variability

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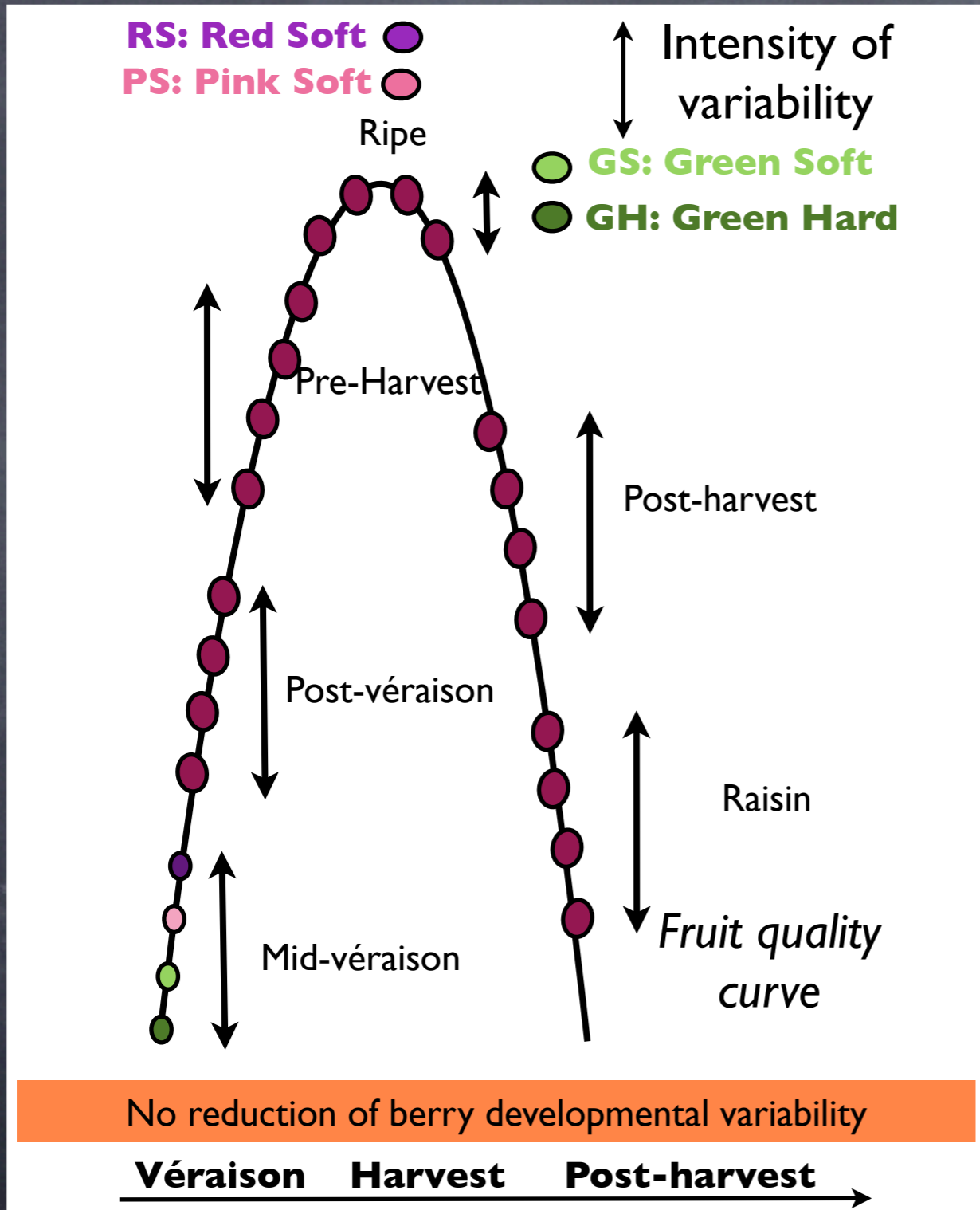
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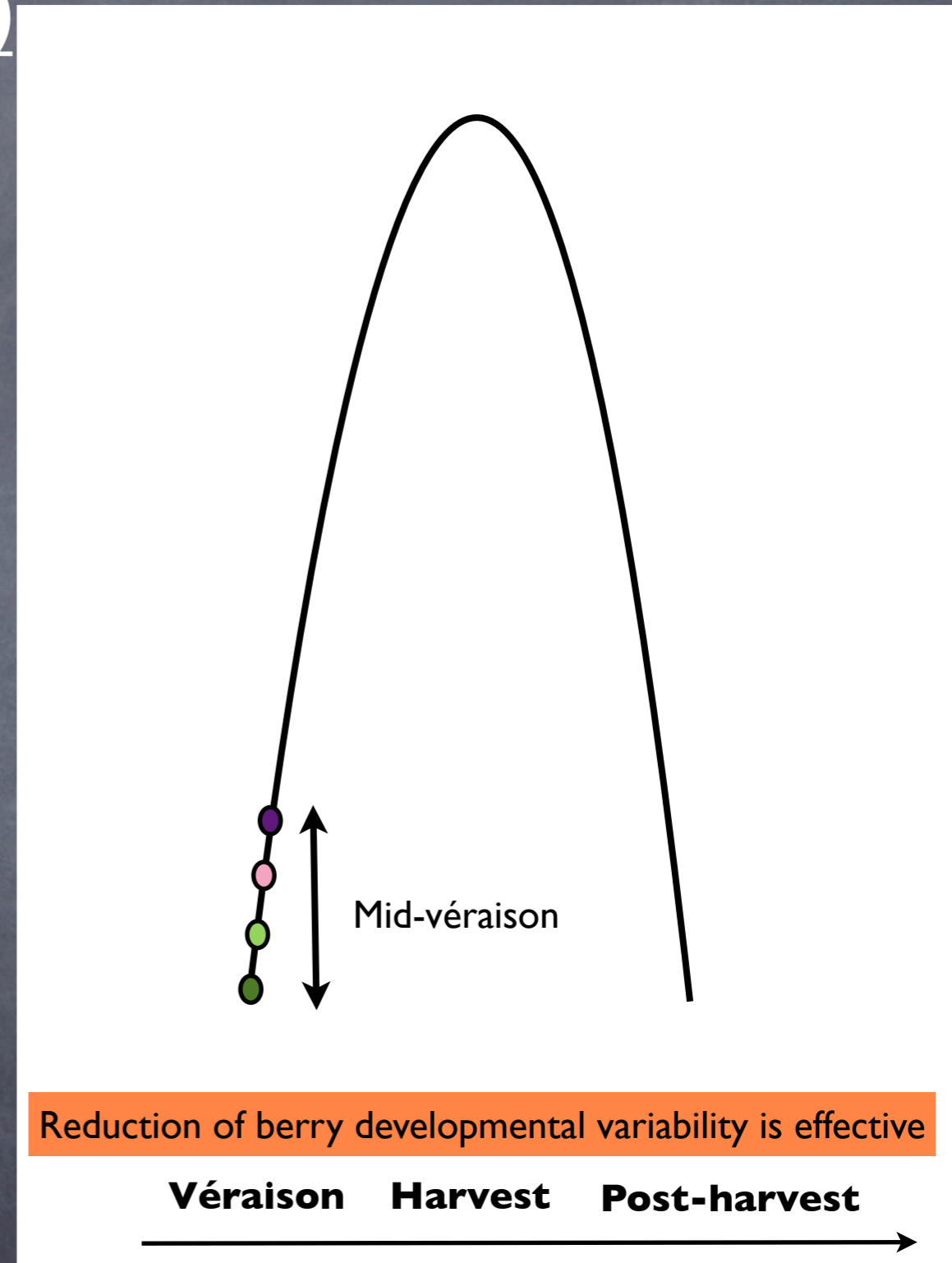


# Reduction of berry class variability

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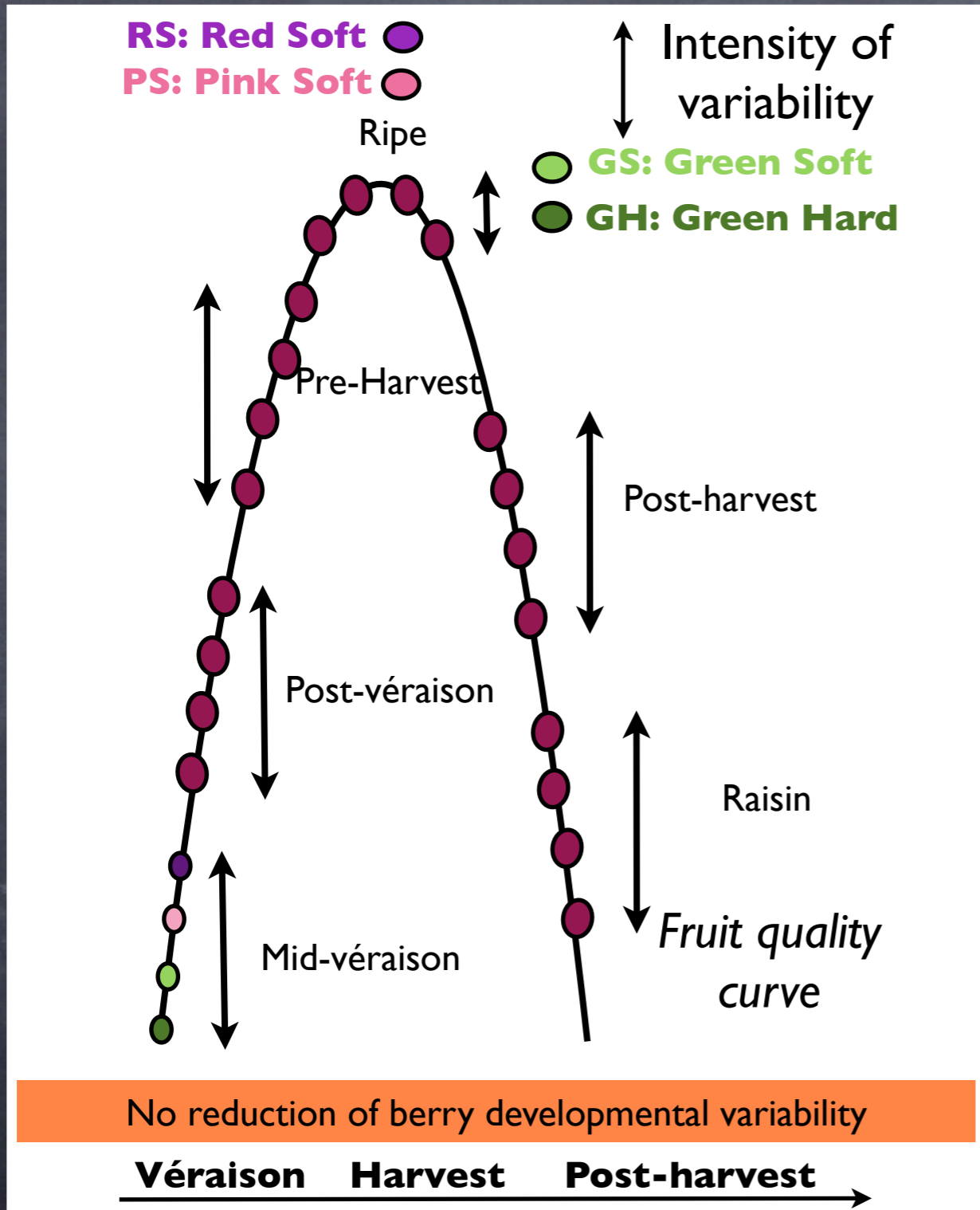
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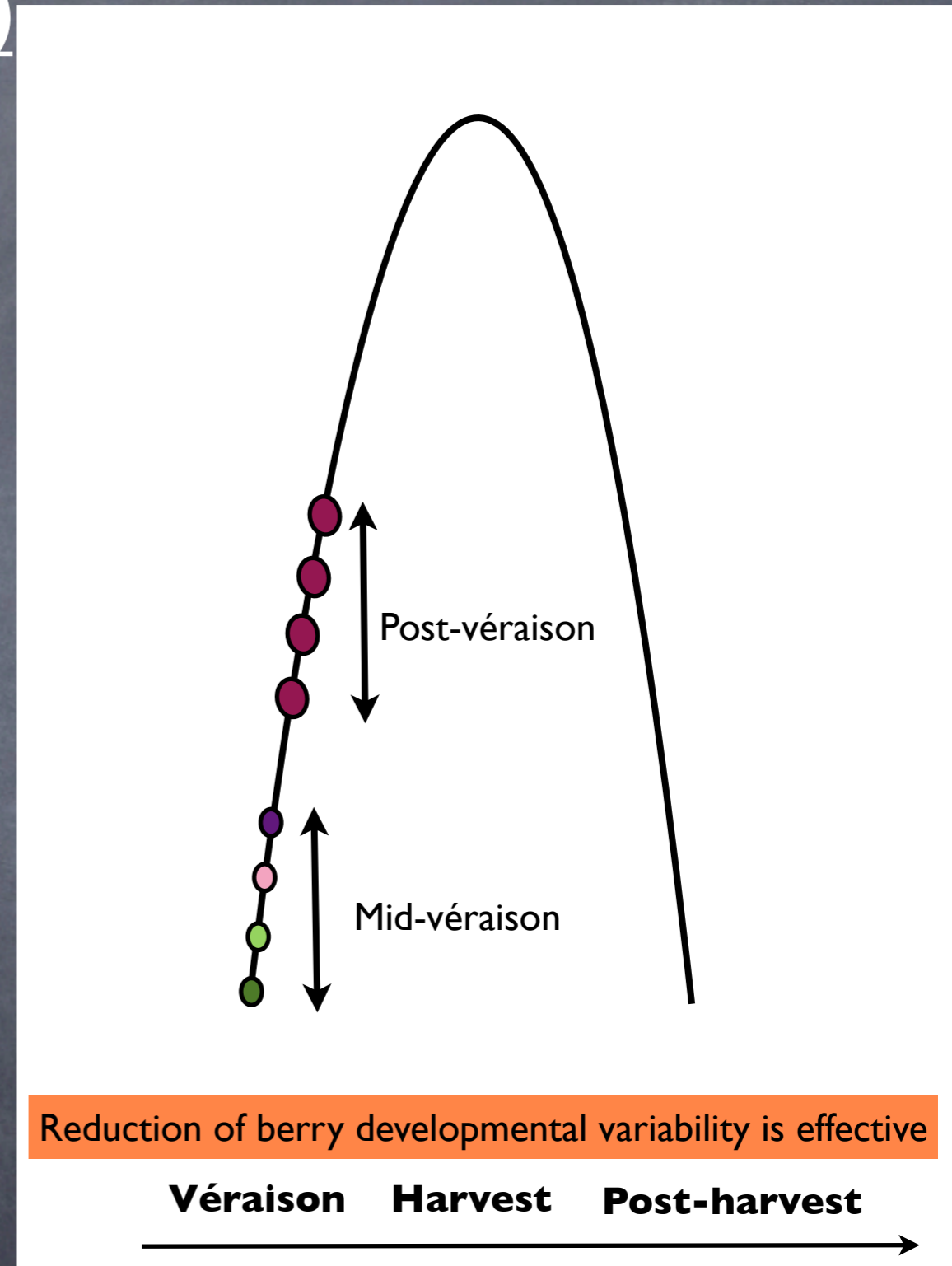


# Reduction of berry class variability

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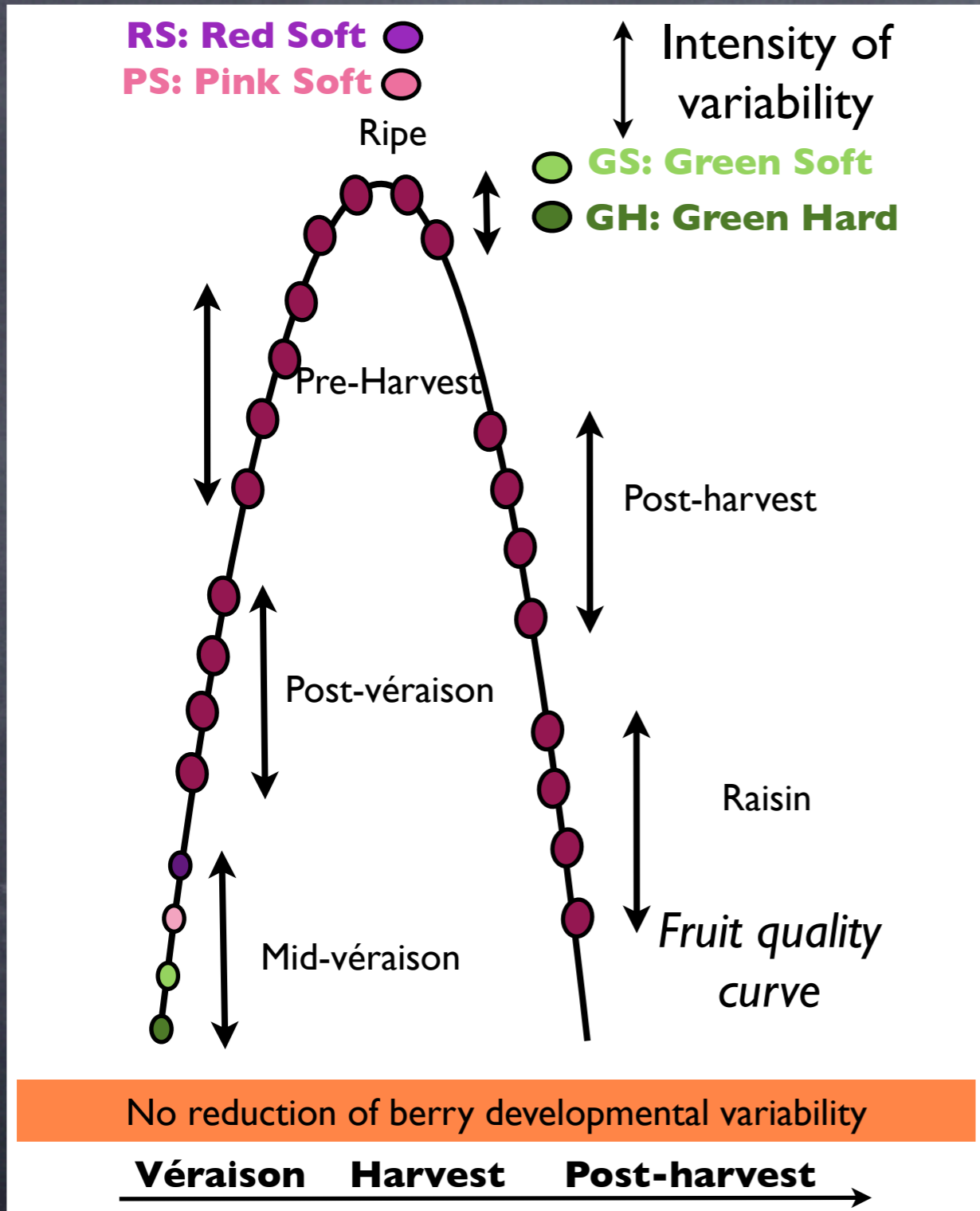
Reduction of berry developmental variability is effective

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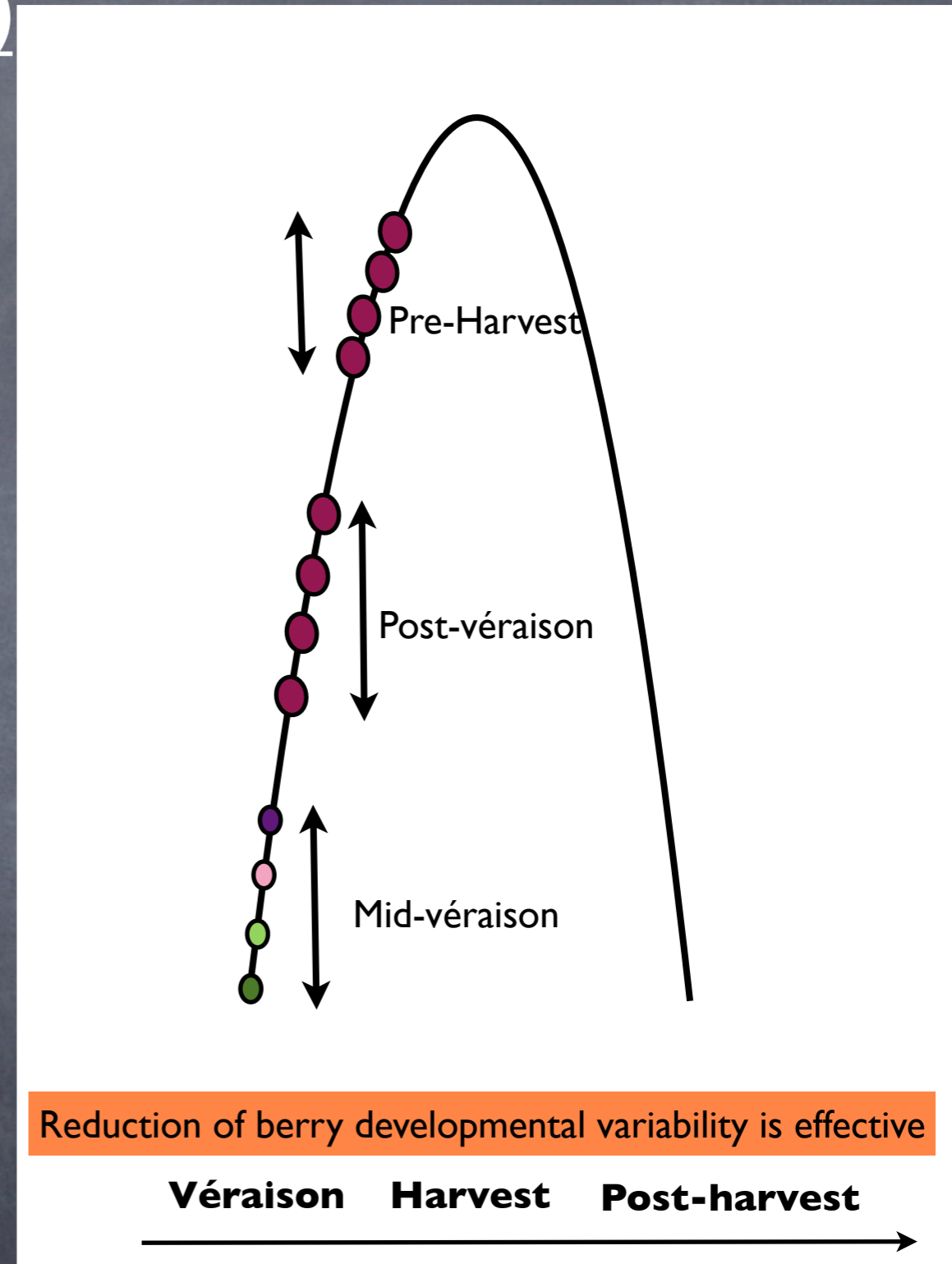


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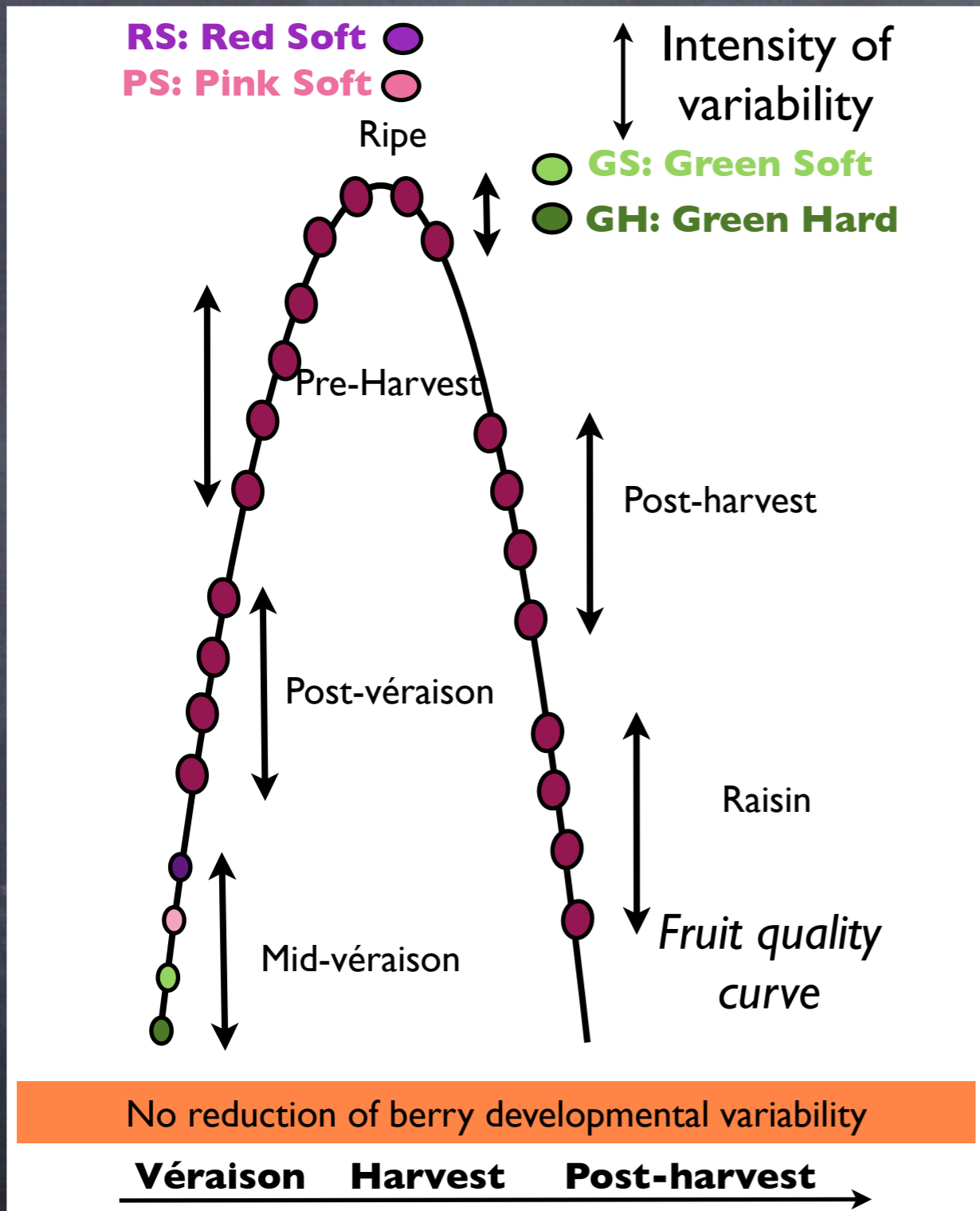
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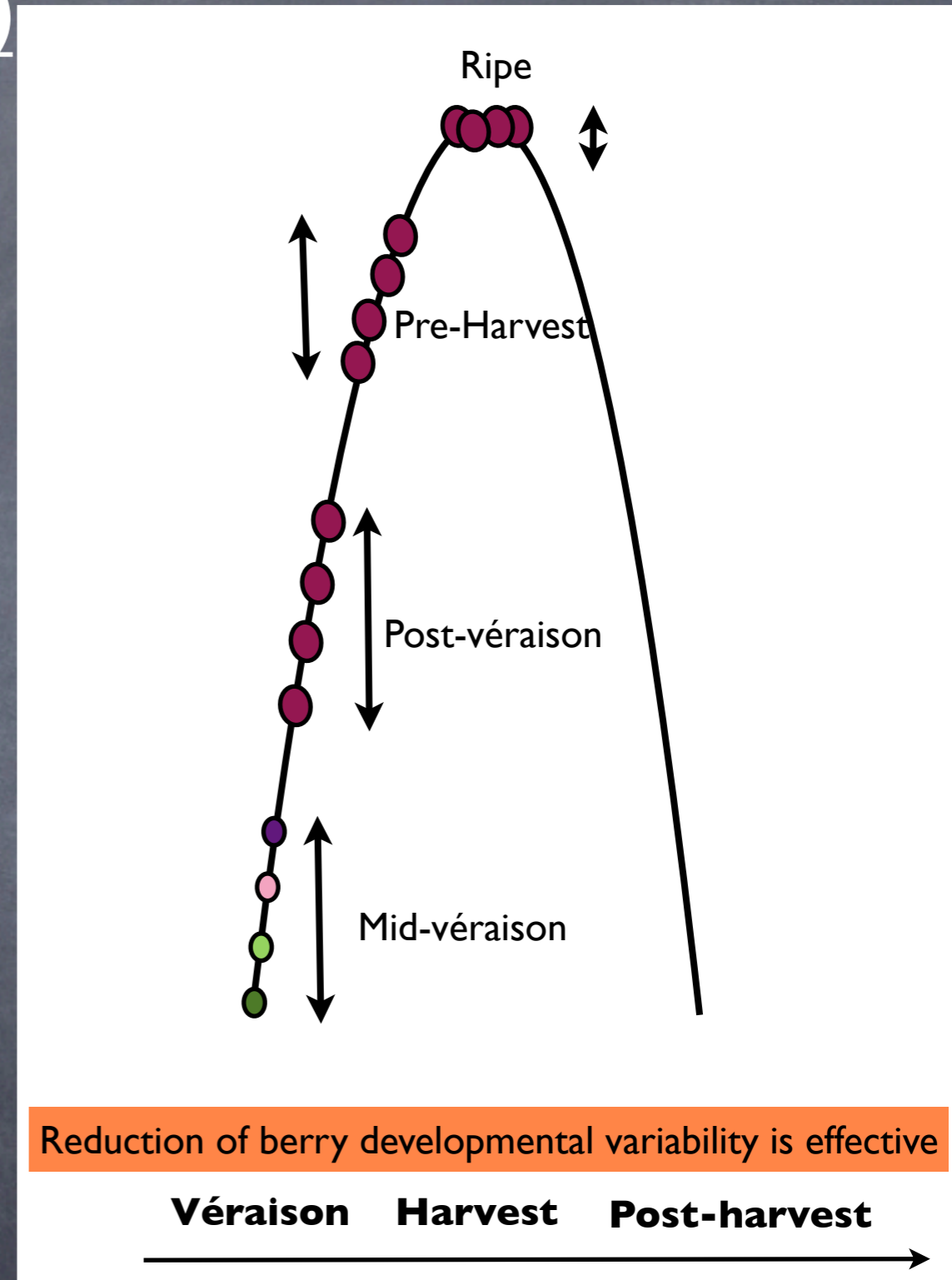


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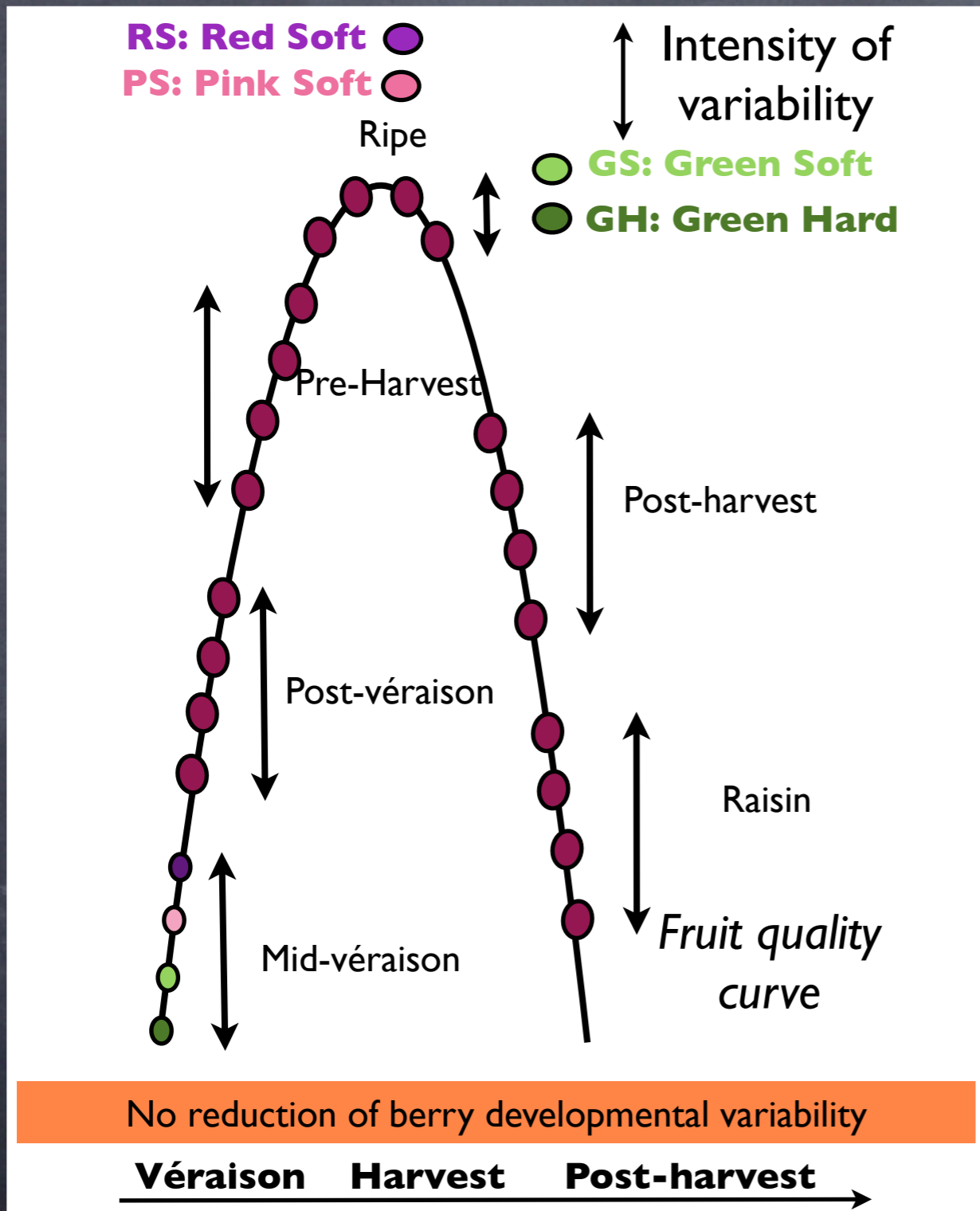
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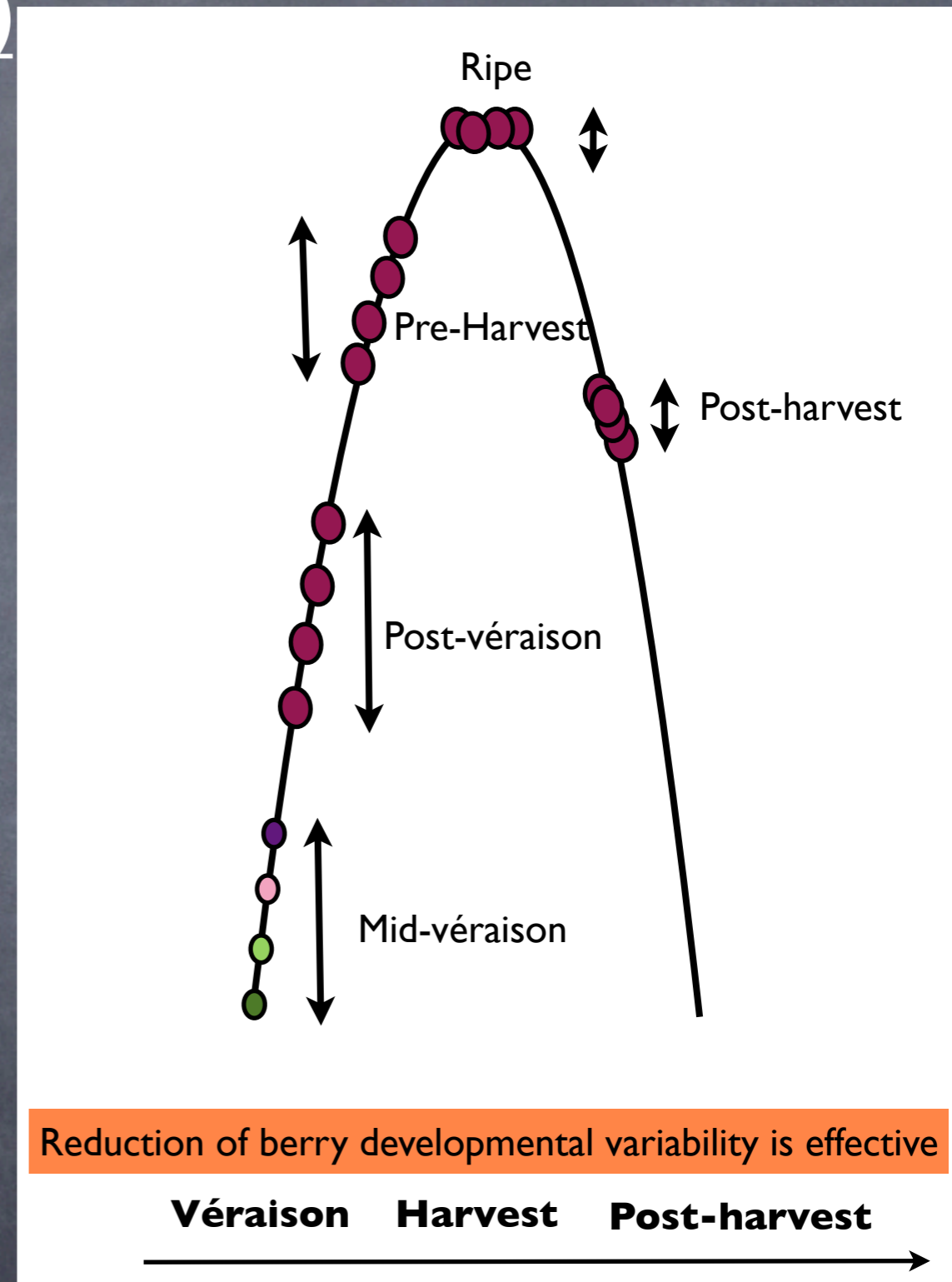


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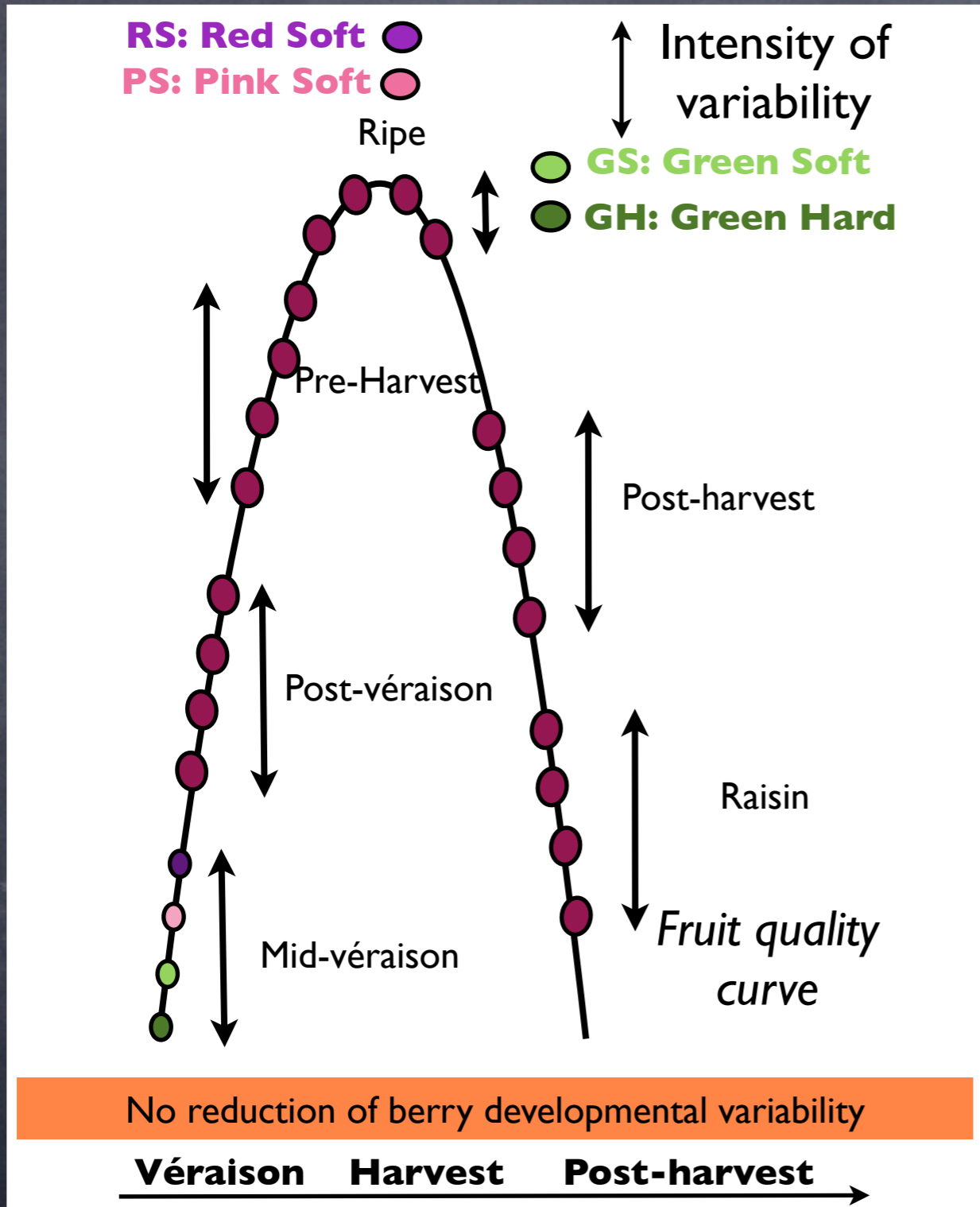
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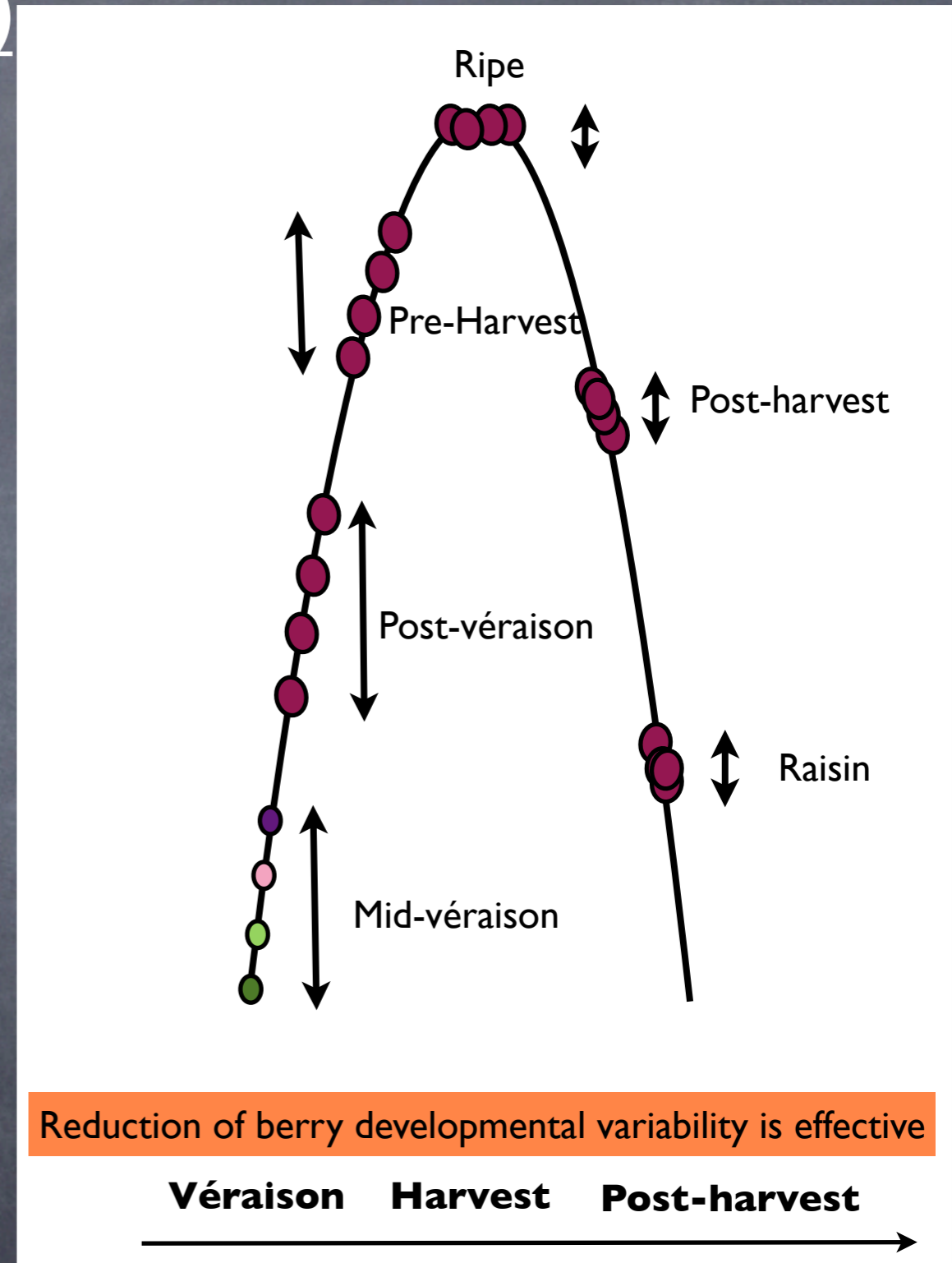


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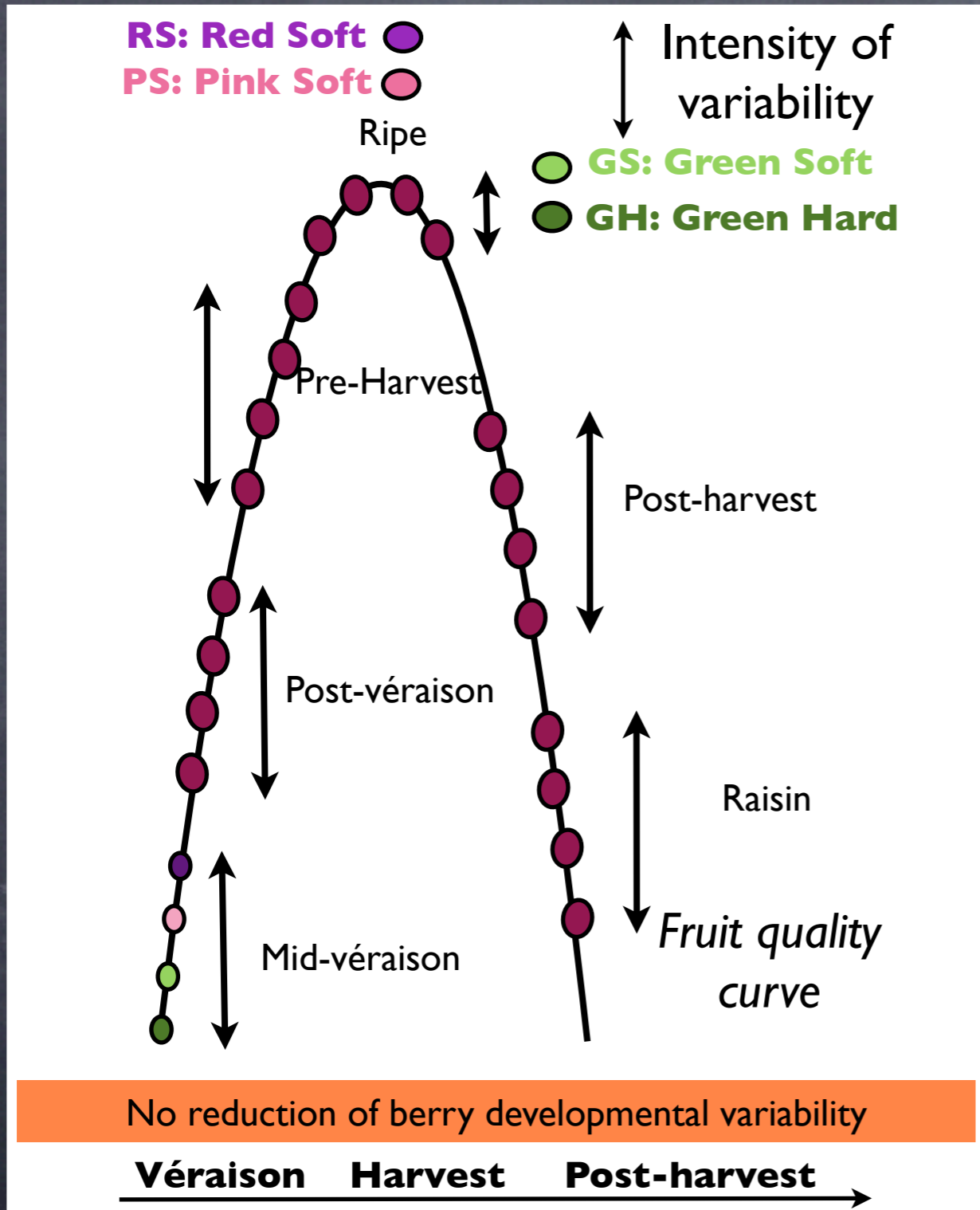
2)



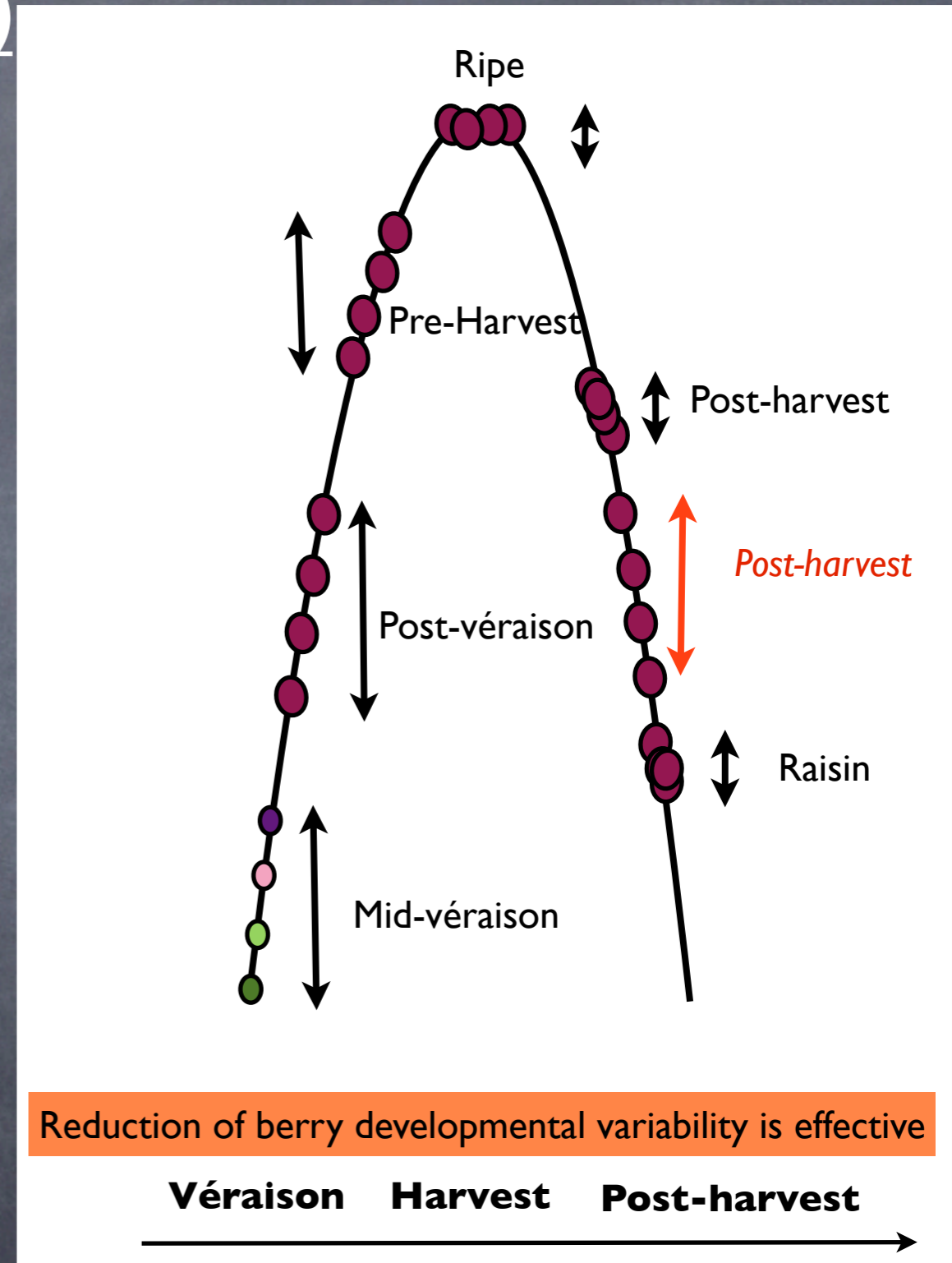
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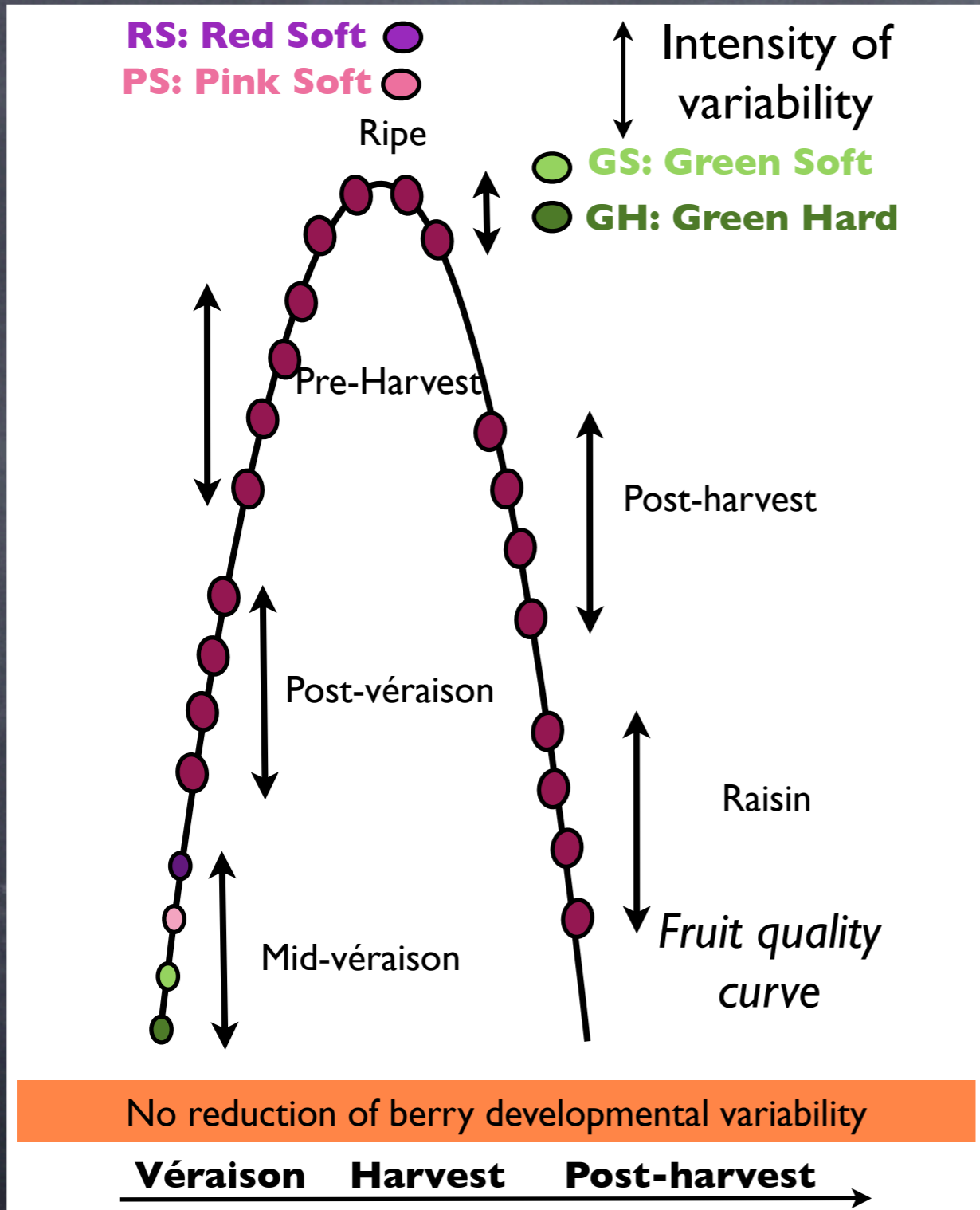
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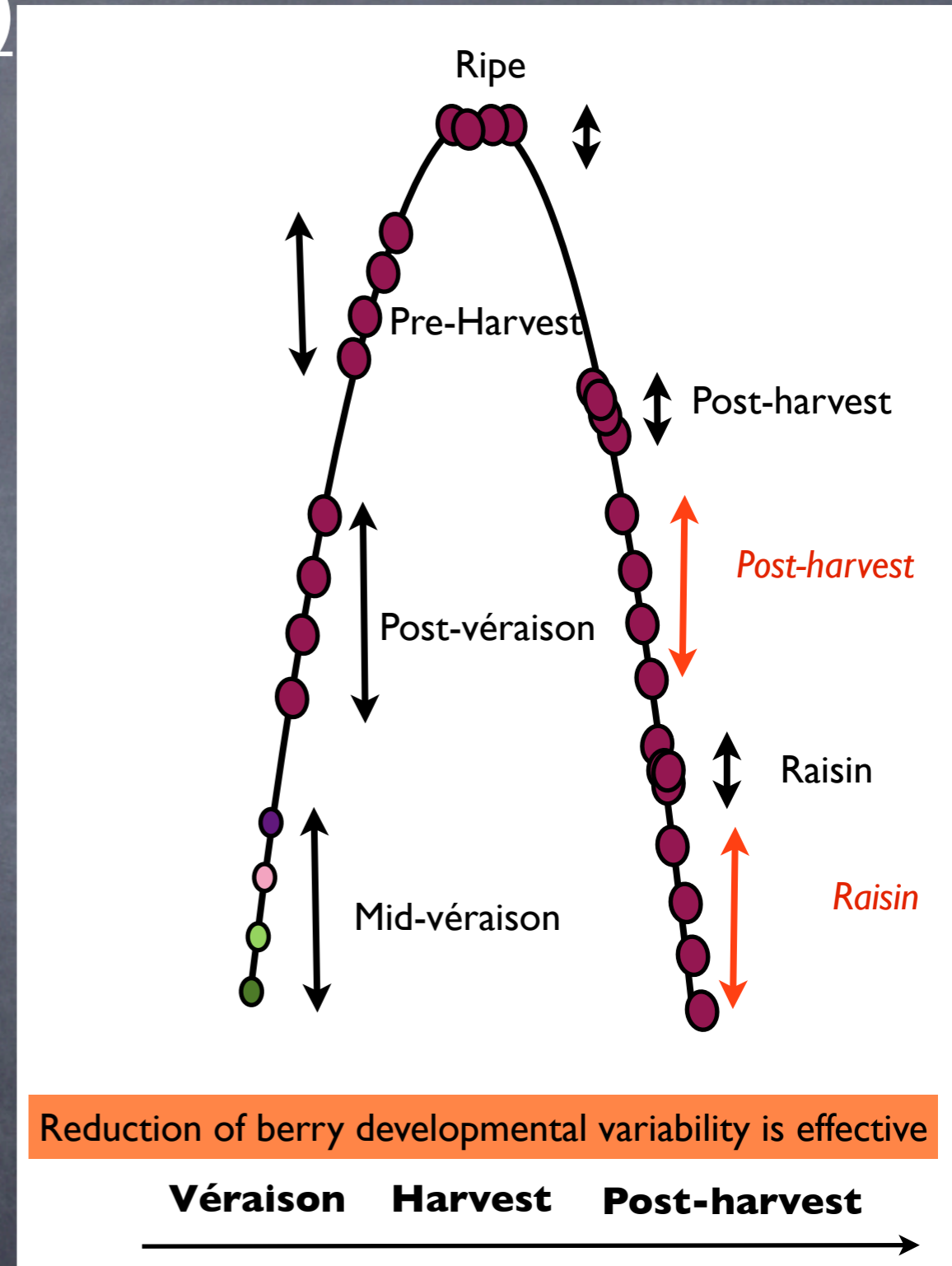


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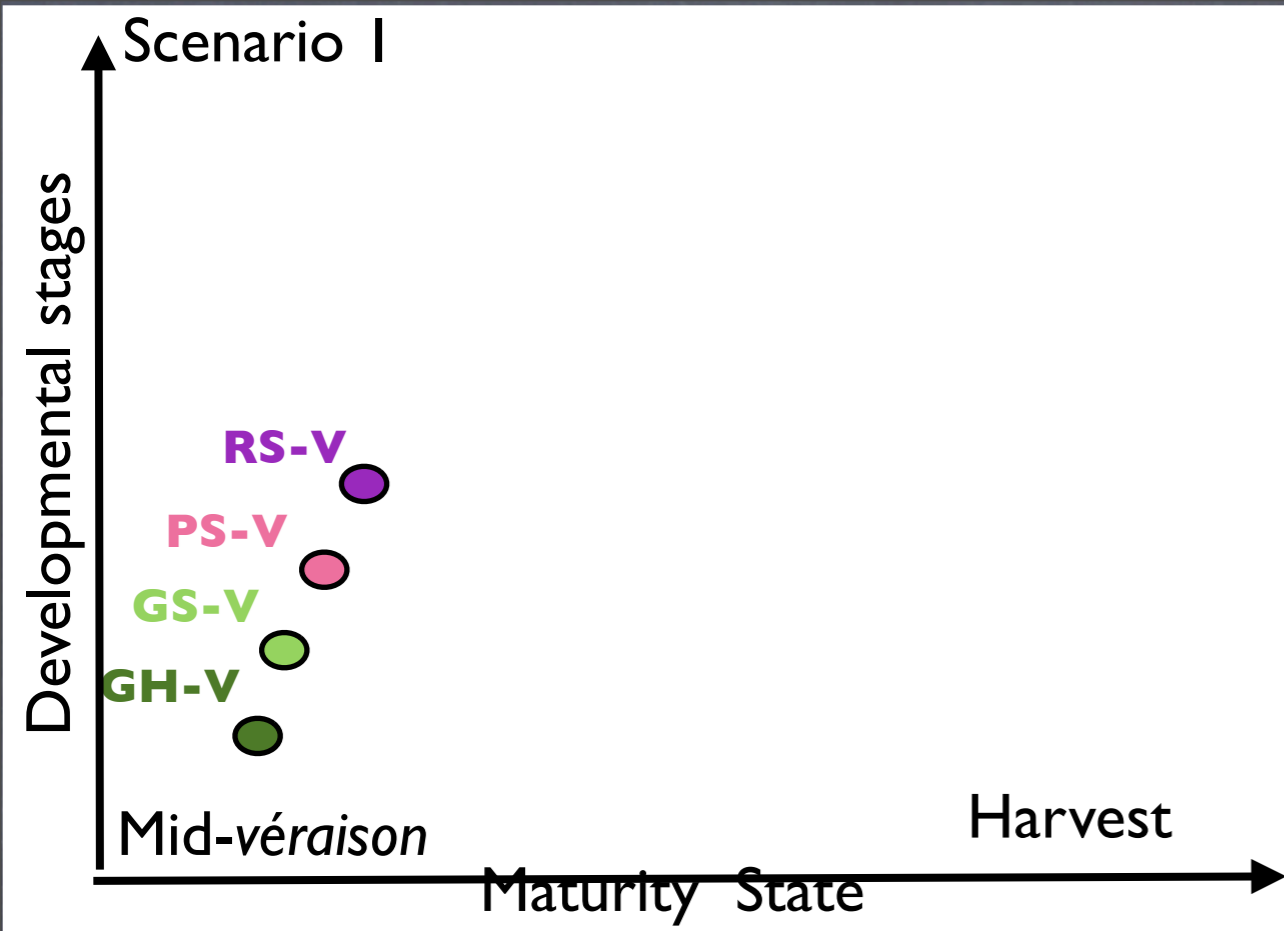


2)

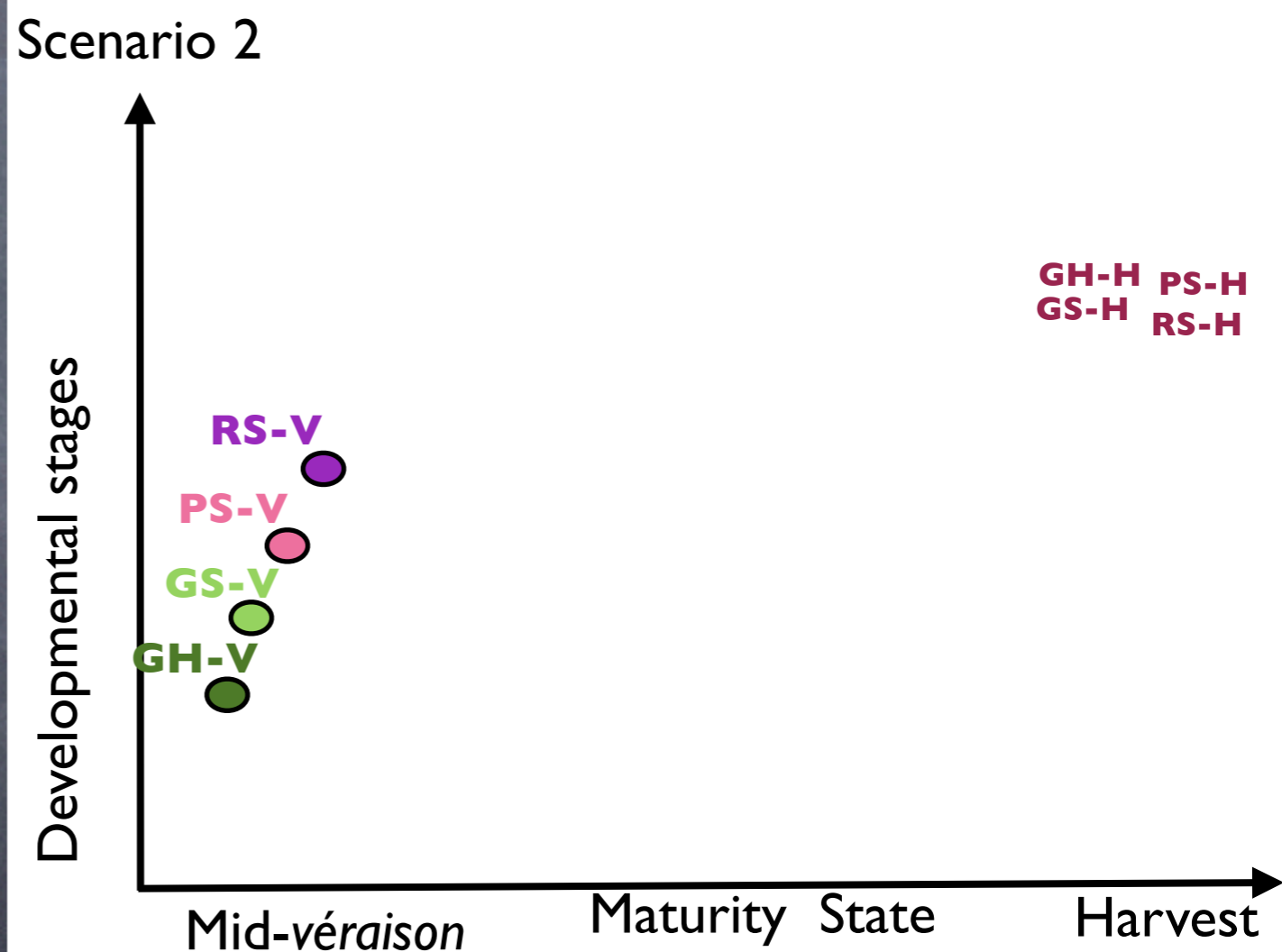


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If we assume differences at mid-*véraison*, we must still see differences near harvest at the developmental level

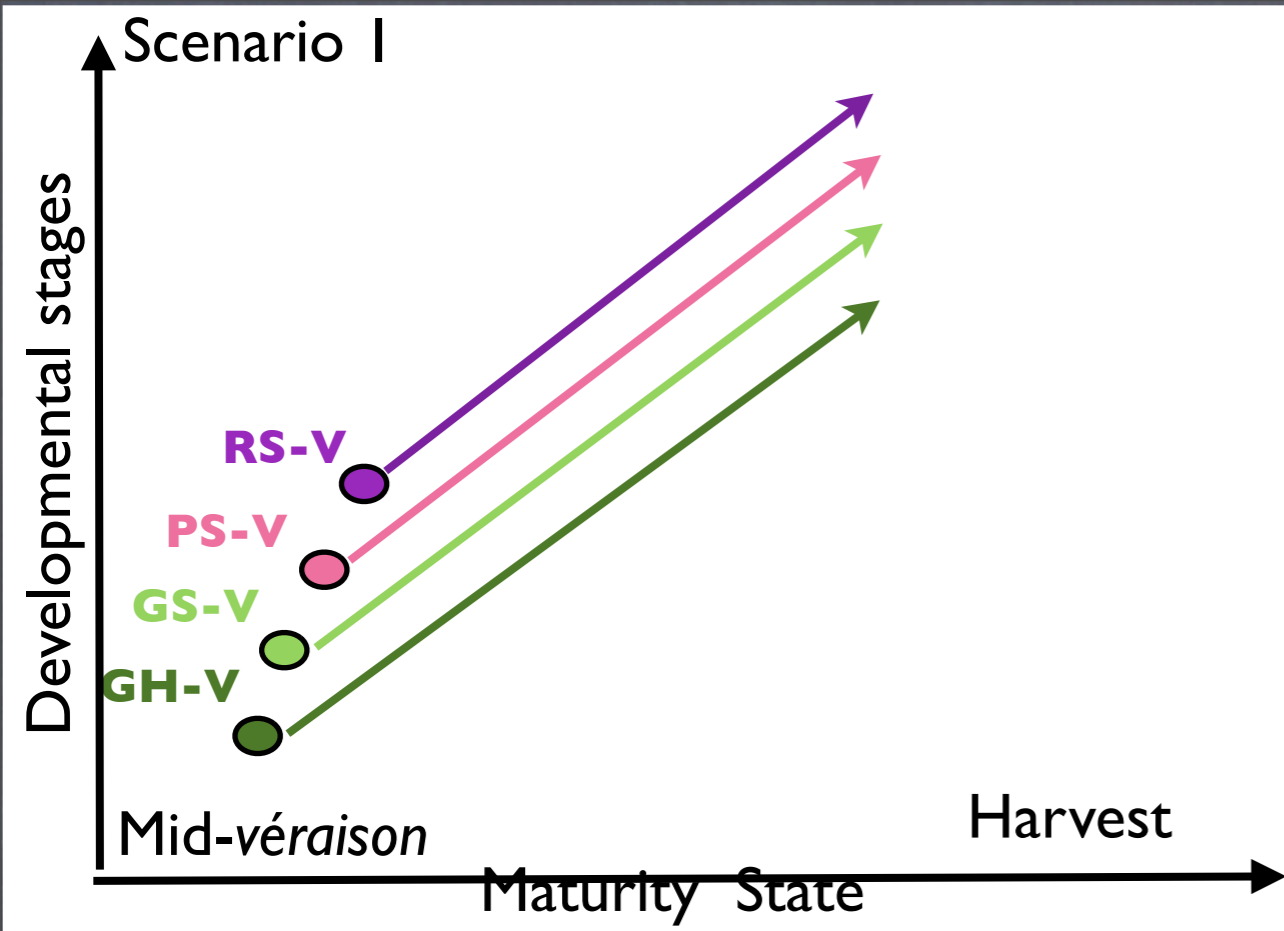


If under ripe berries are catching up we do not know when that happens....

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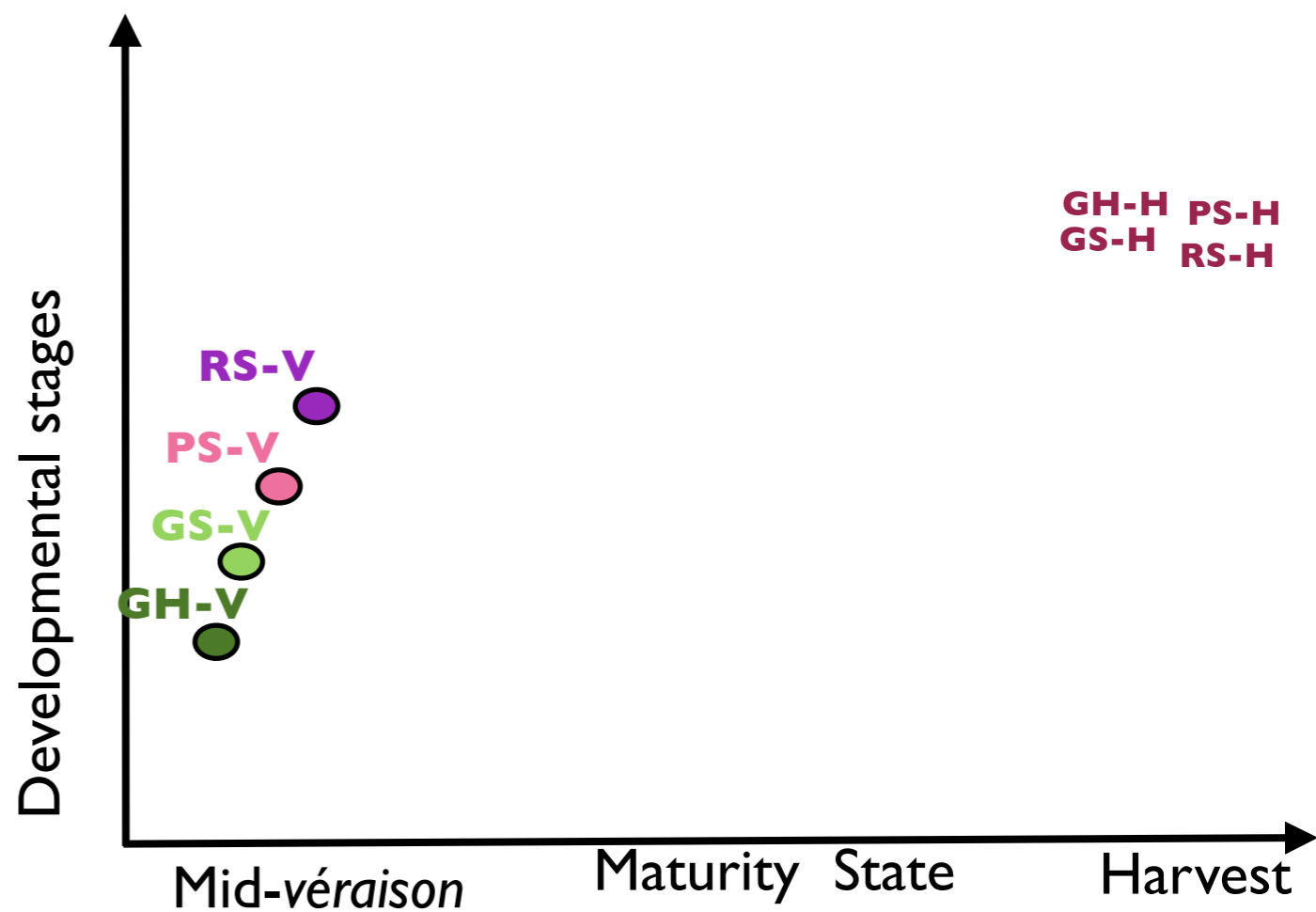
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### Scenario 2

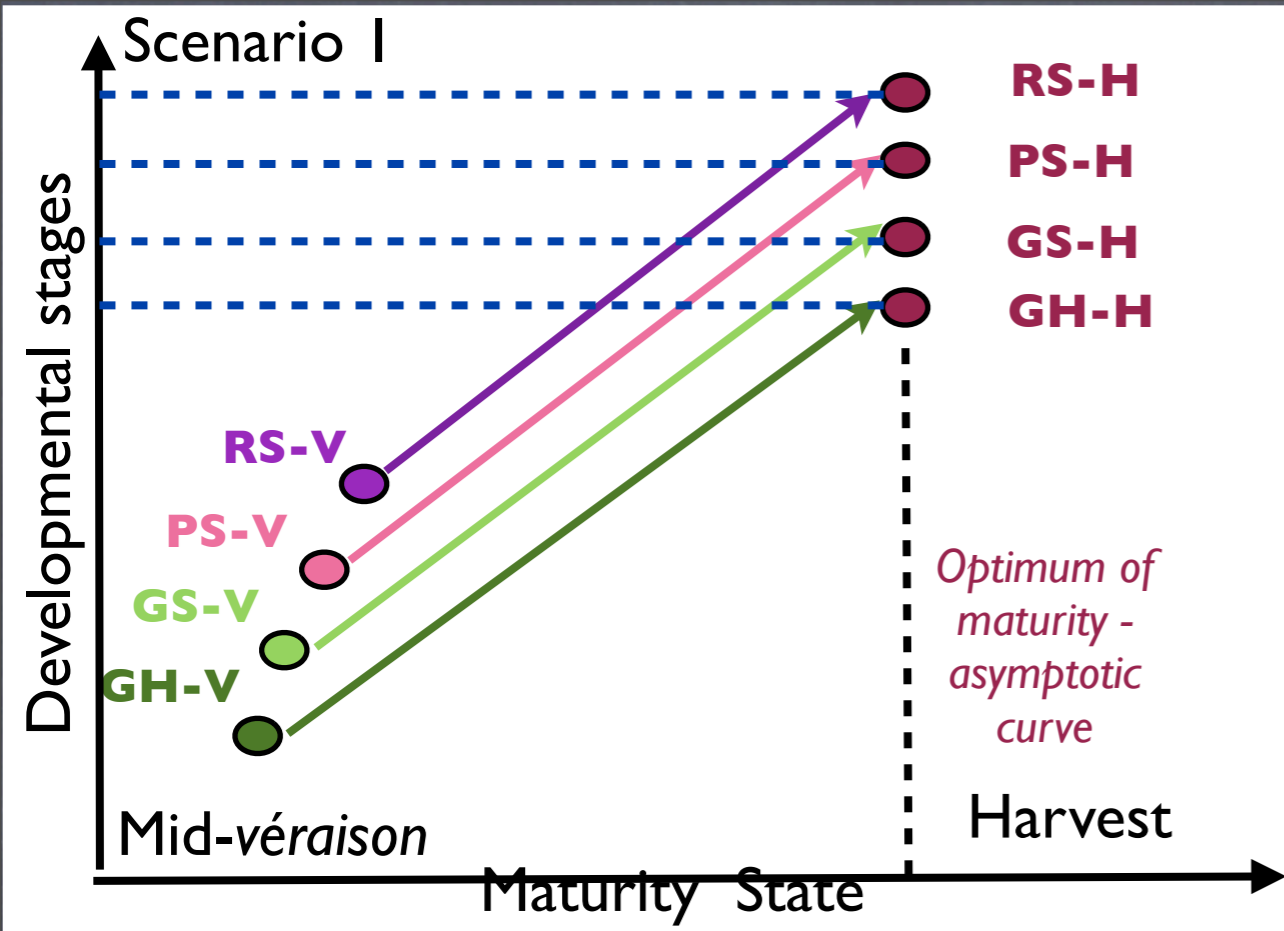


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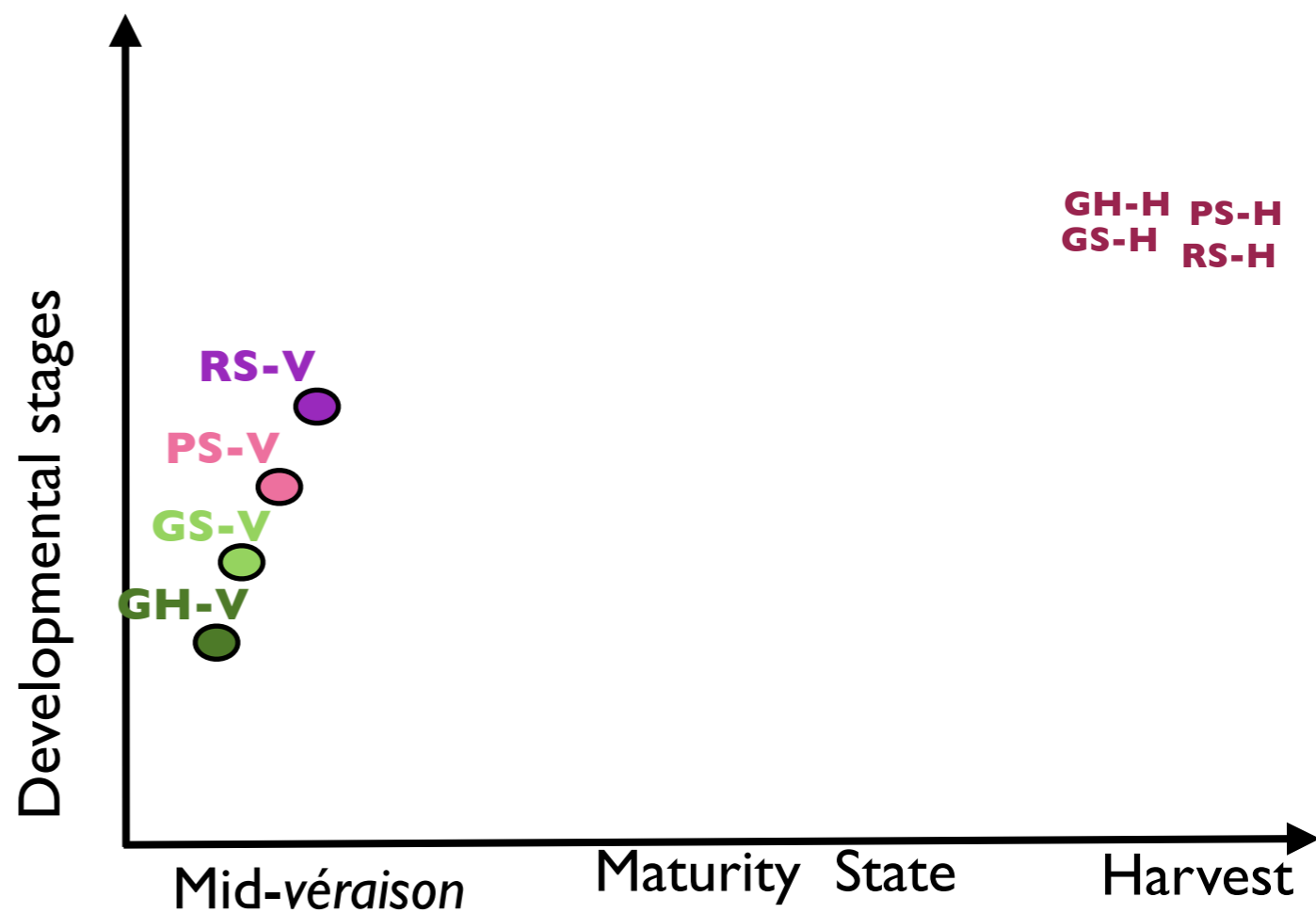
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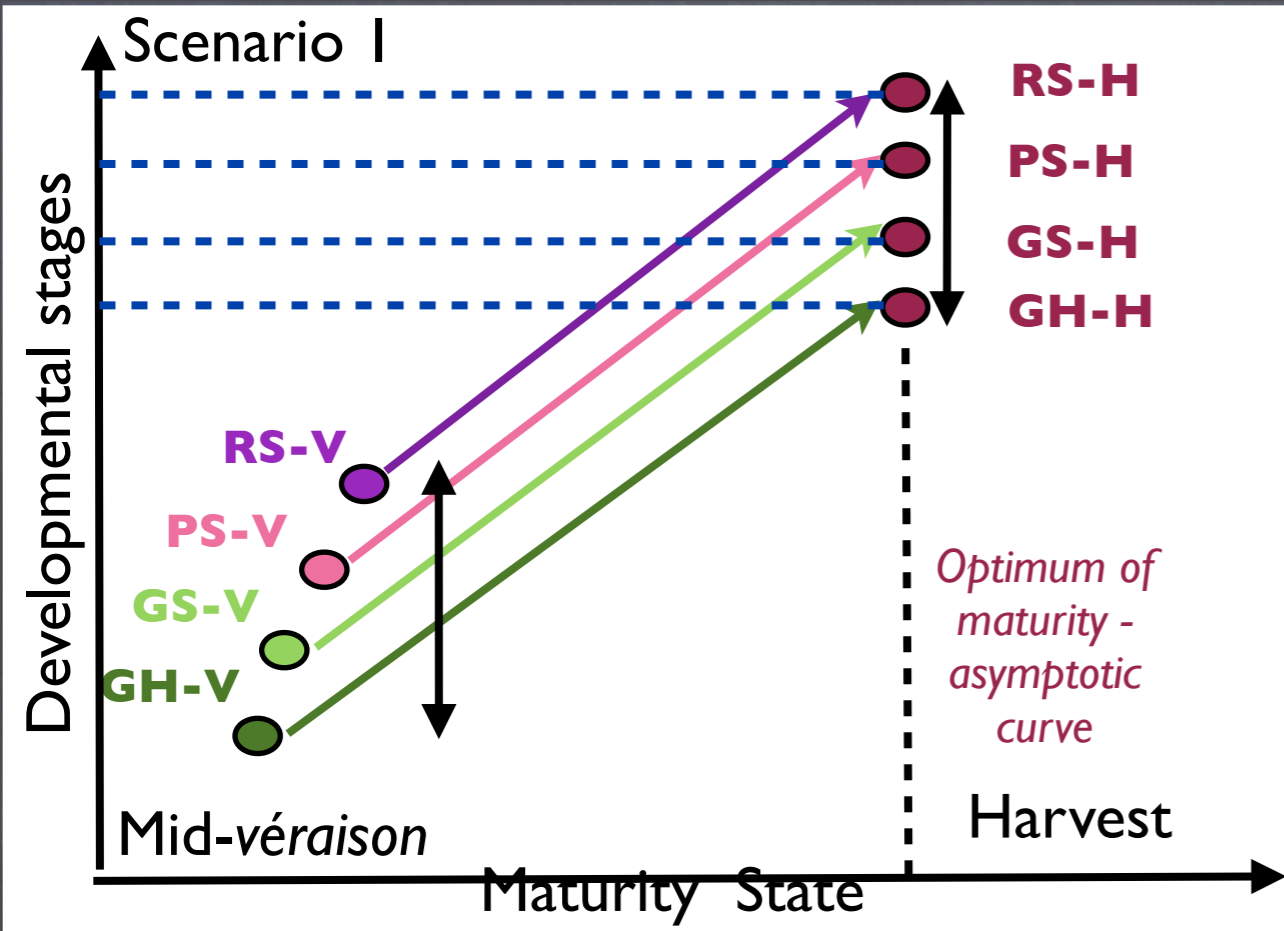


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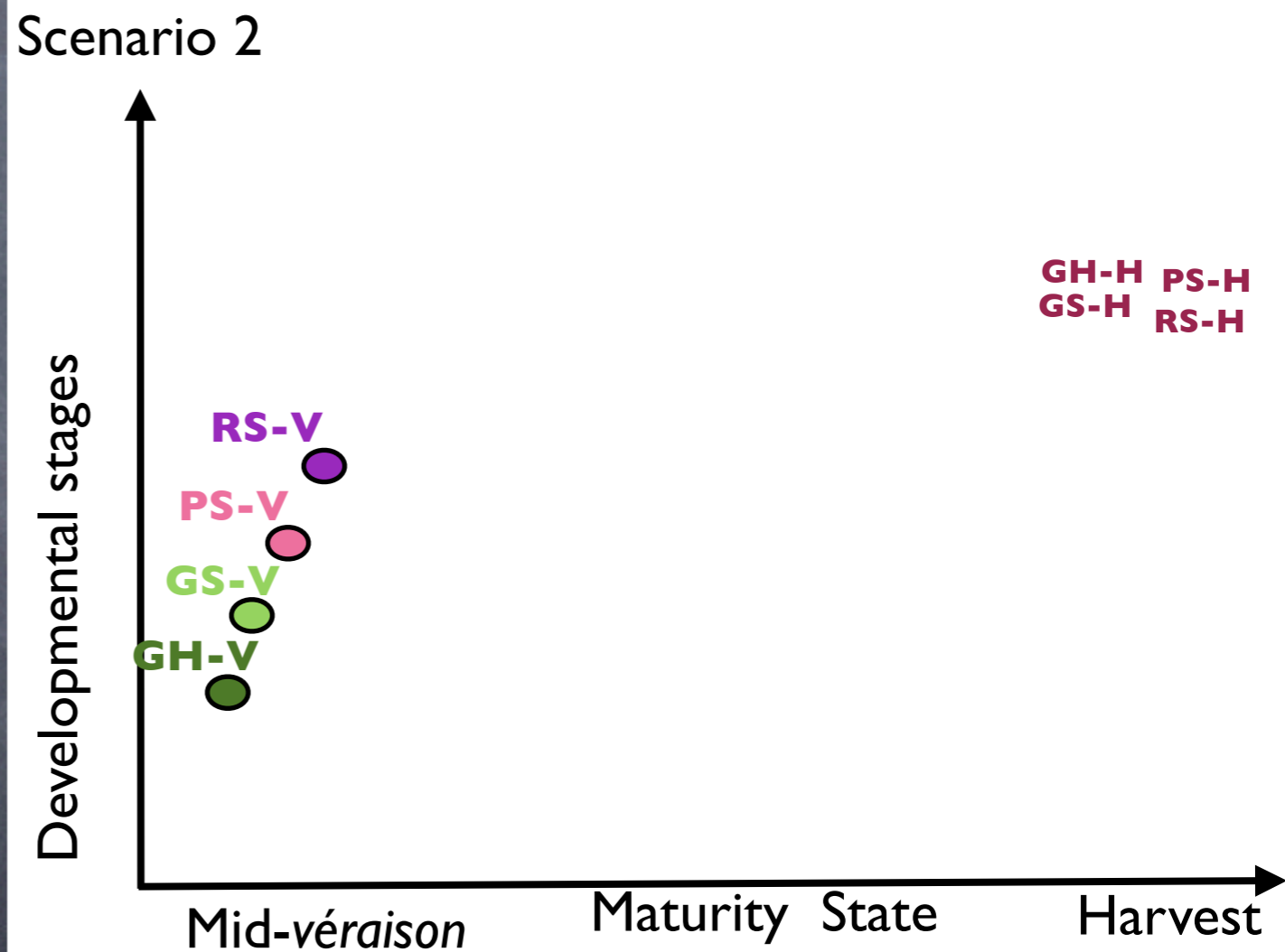
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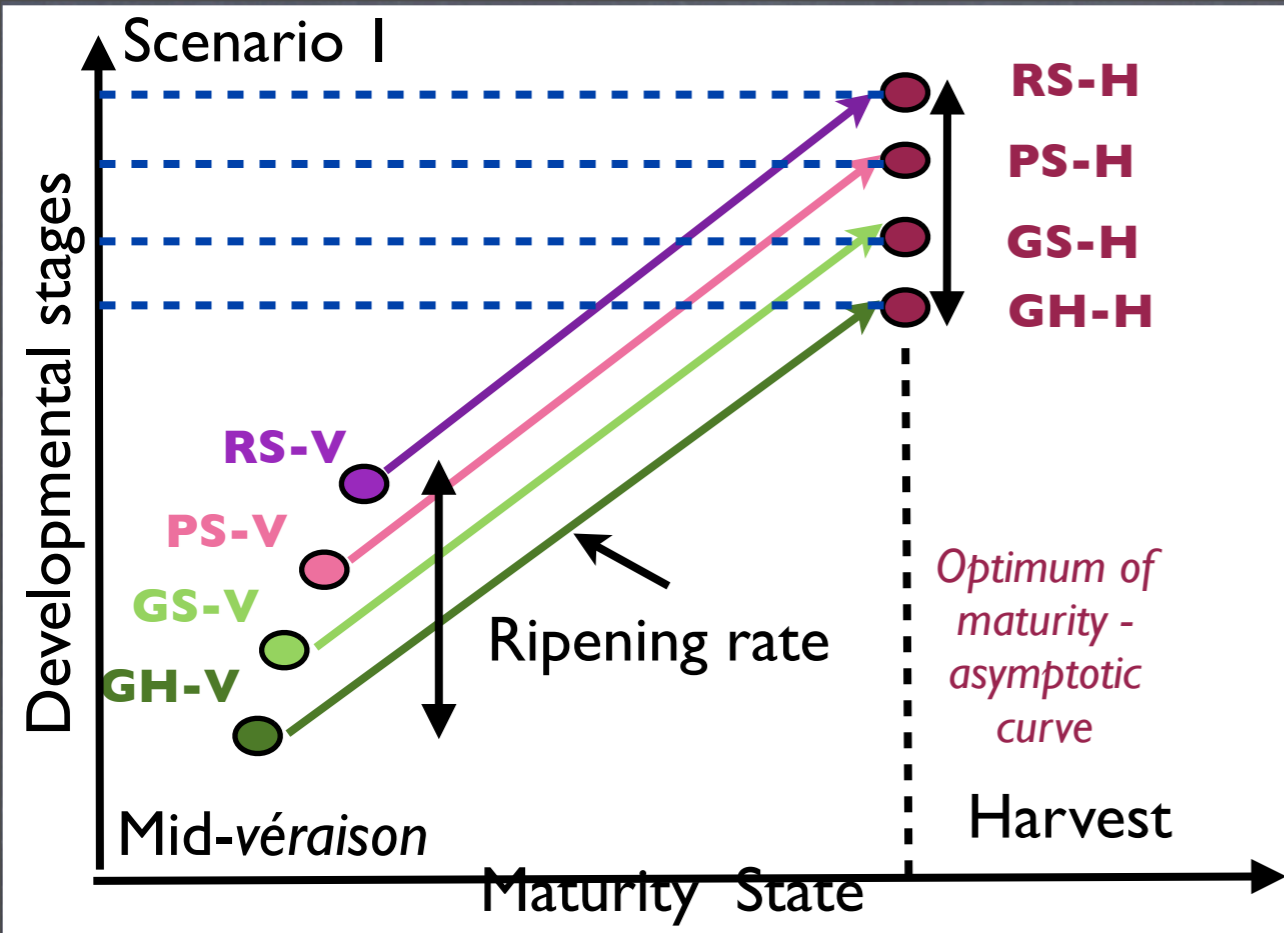
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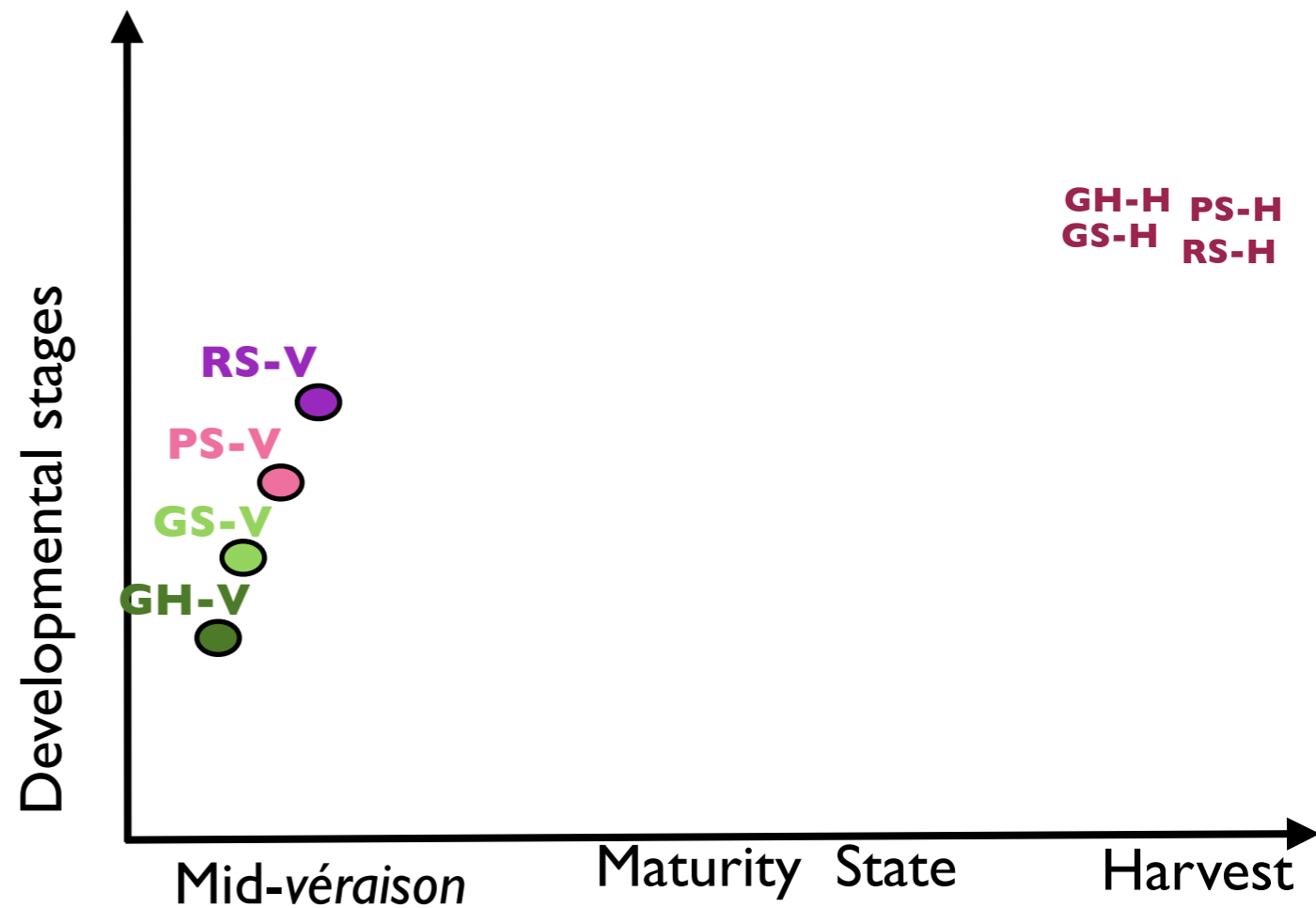




If we assume differences at mid-*véraison*, we must still see differences near harvest at the developmental level

If under ripe berries are catching up we do not know when that happens....

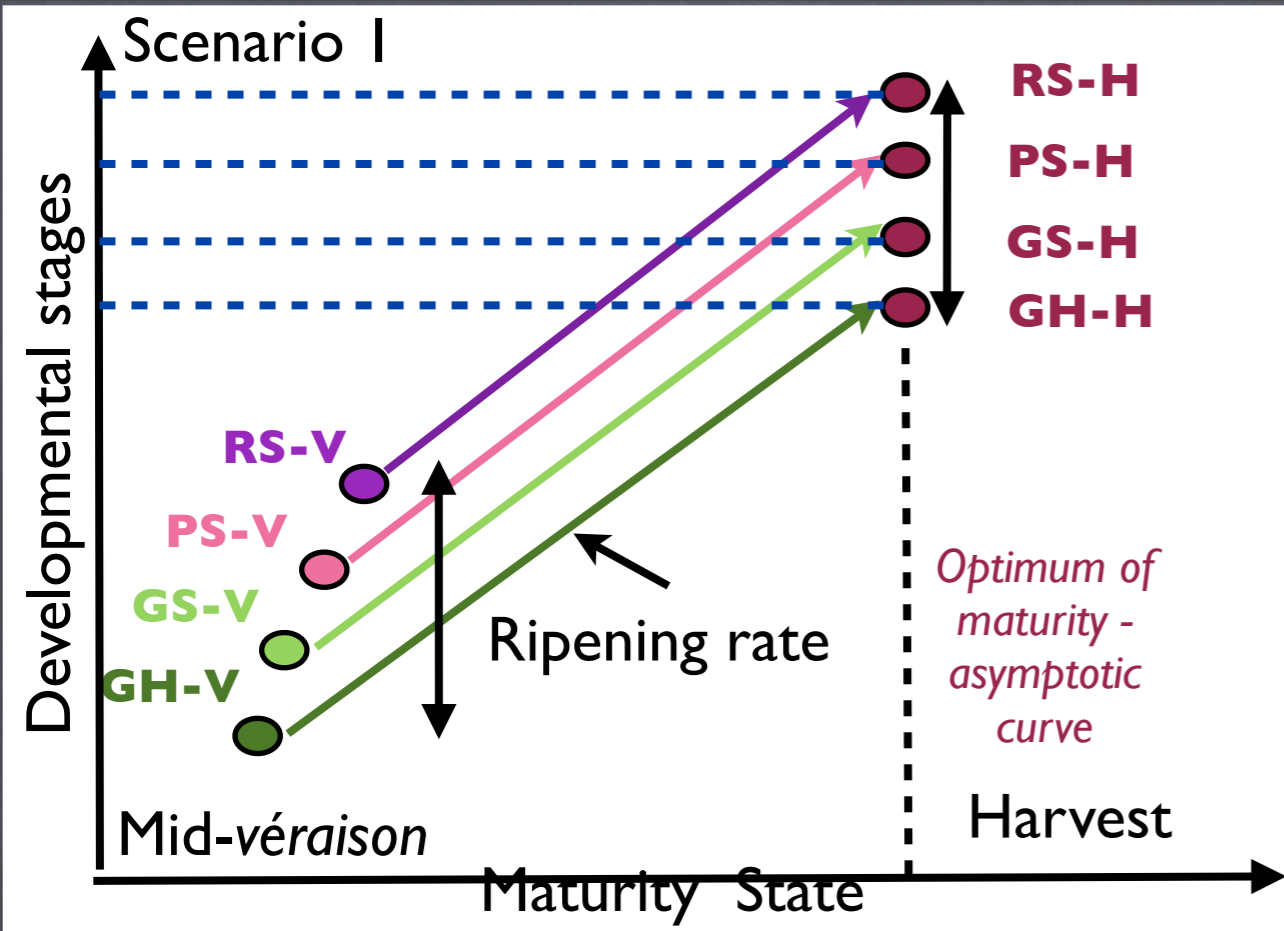
**Scenario 2**



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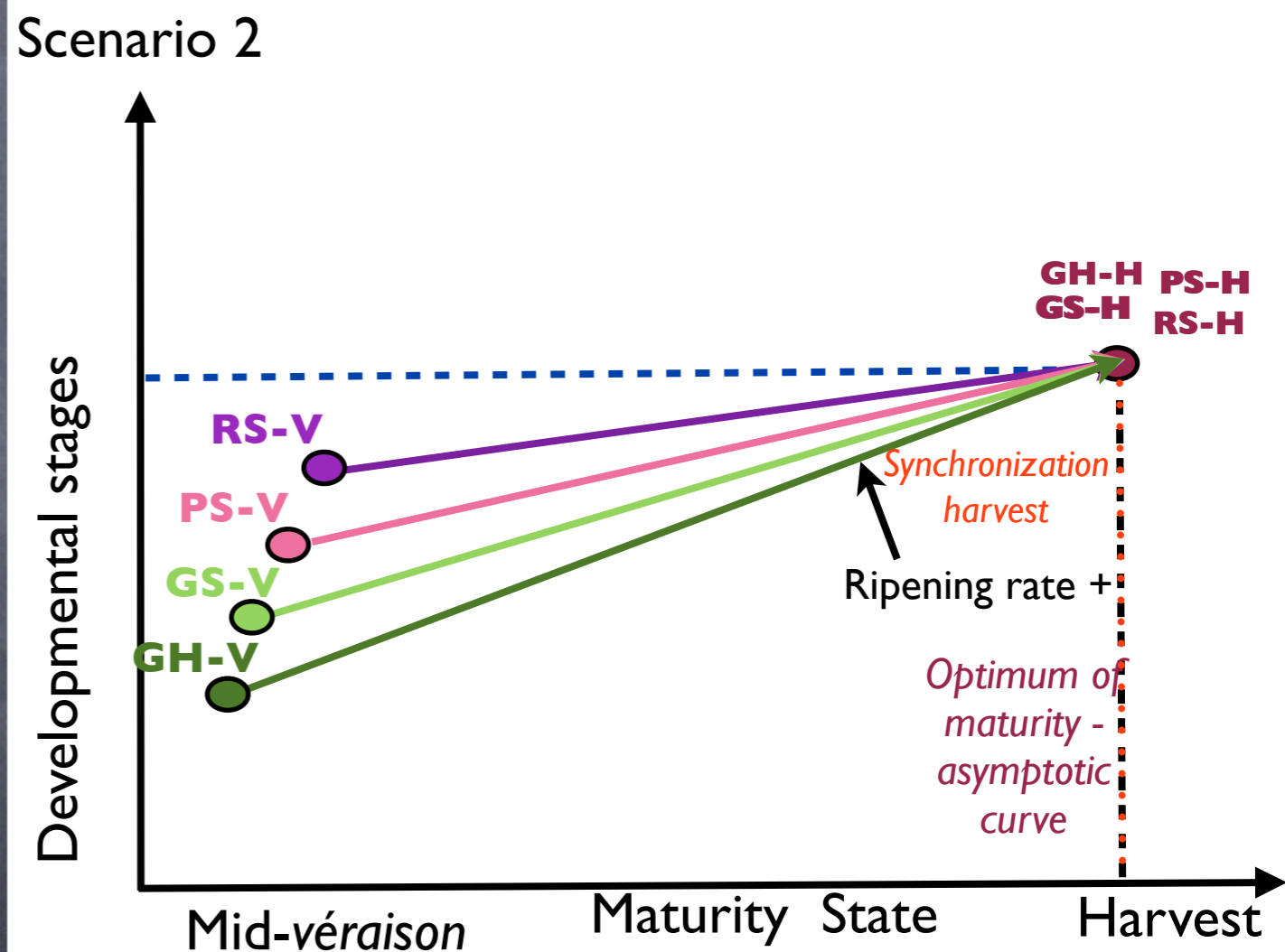
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If we assume differences at mid-*véraison*, we must still see differences near harvest at the developmental level

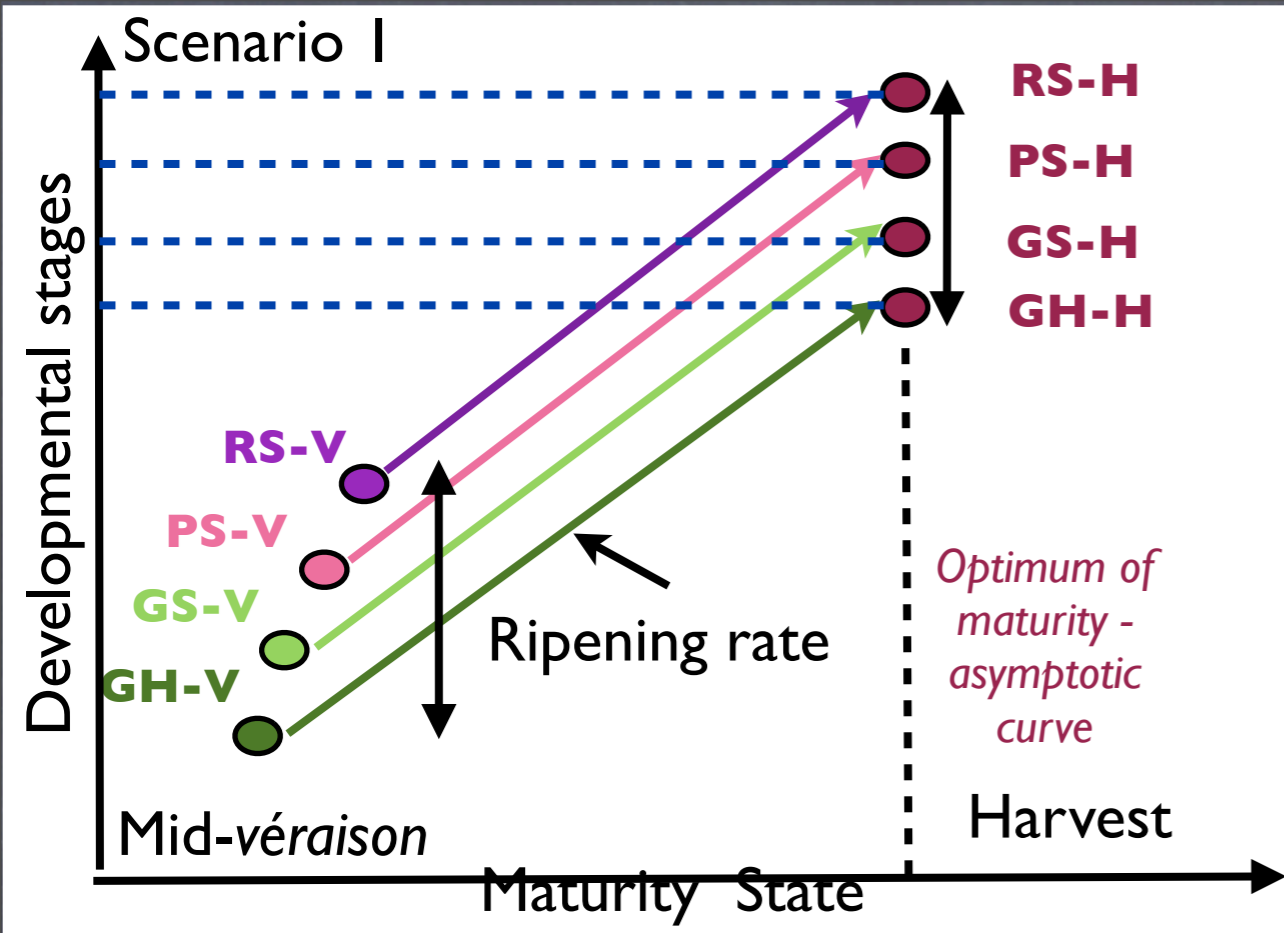
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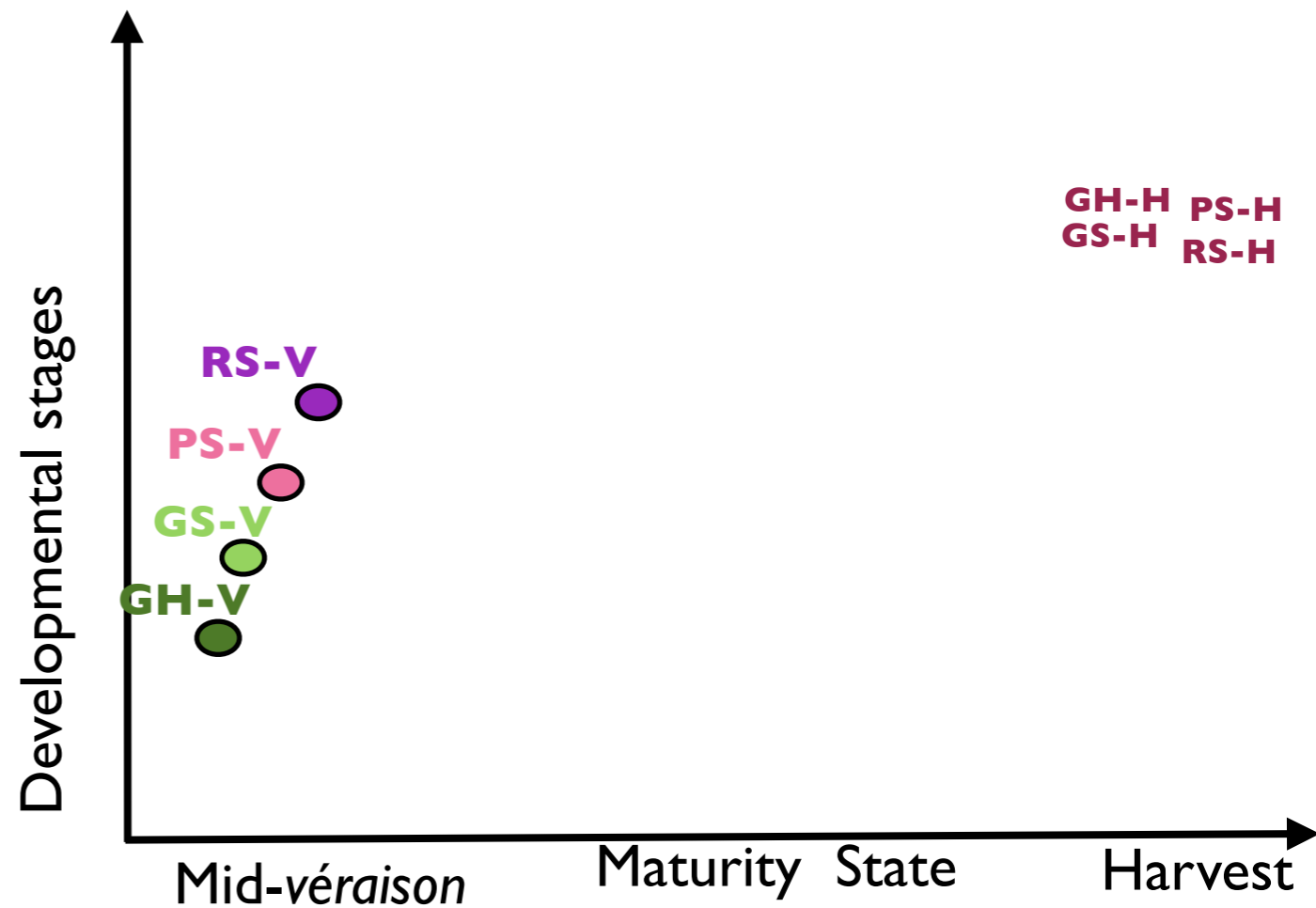
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If we assume differences at mid-*véraison*, we must still see differences near harvest at the developmental level

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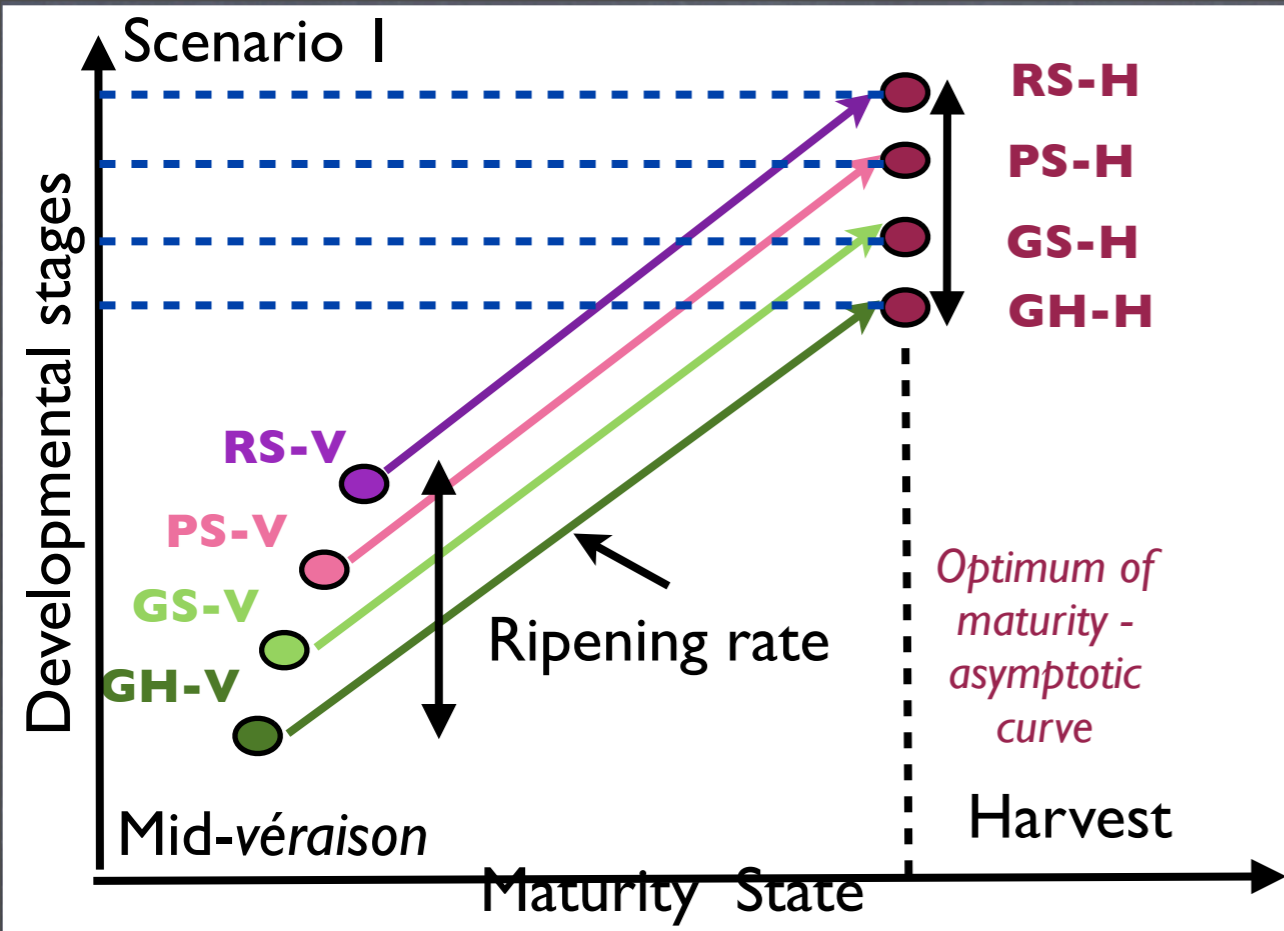


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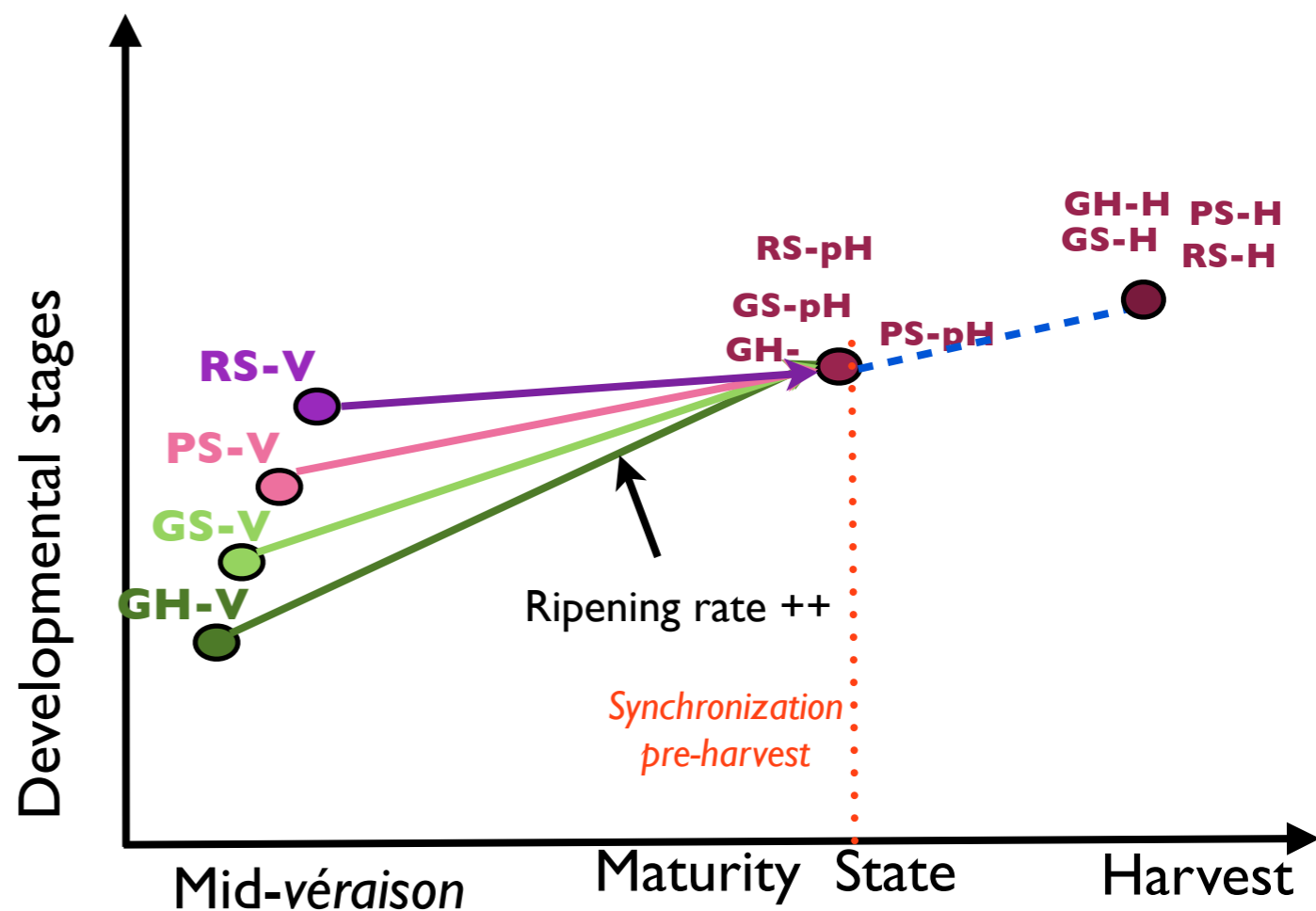




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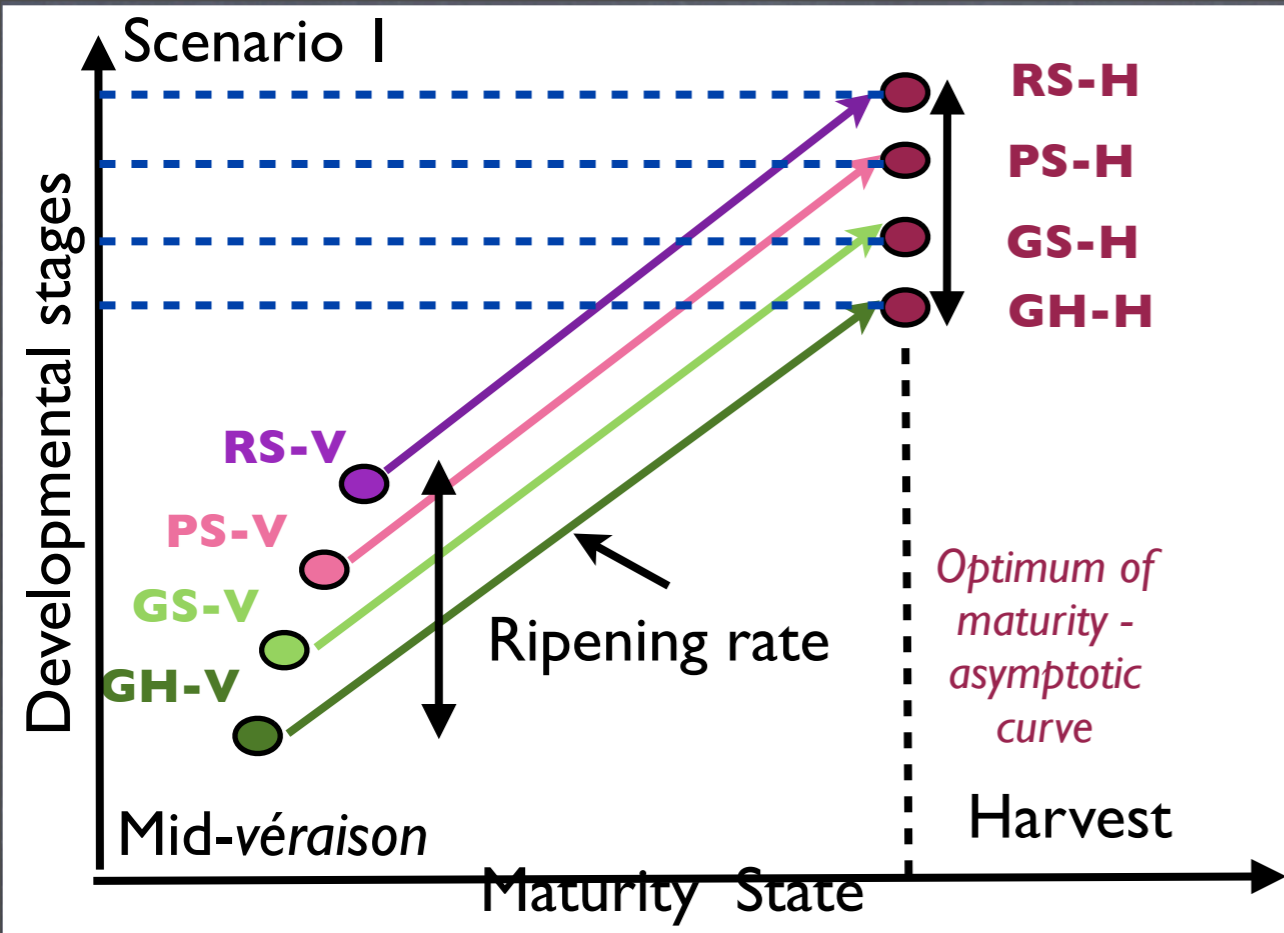


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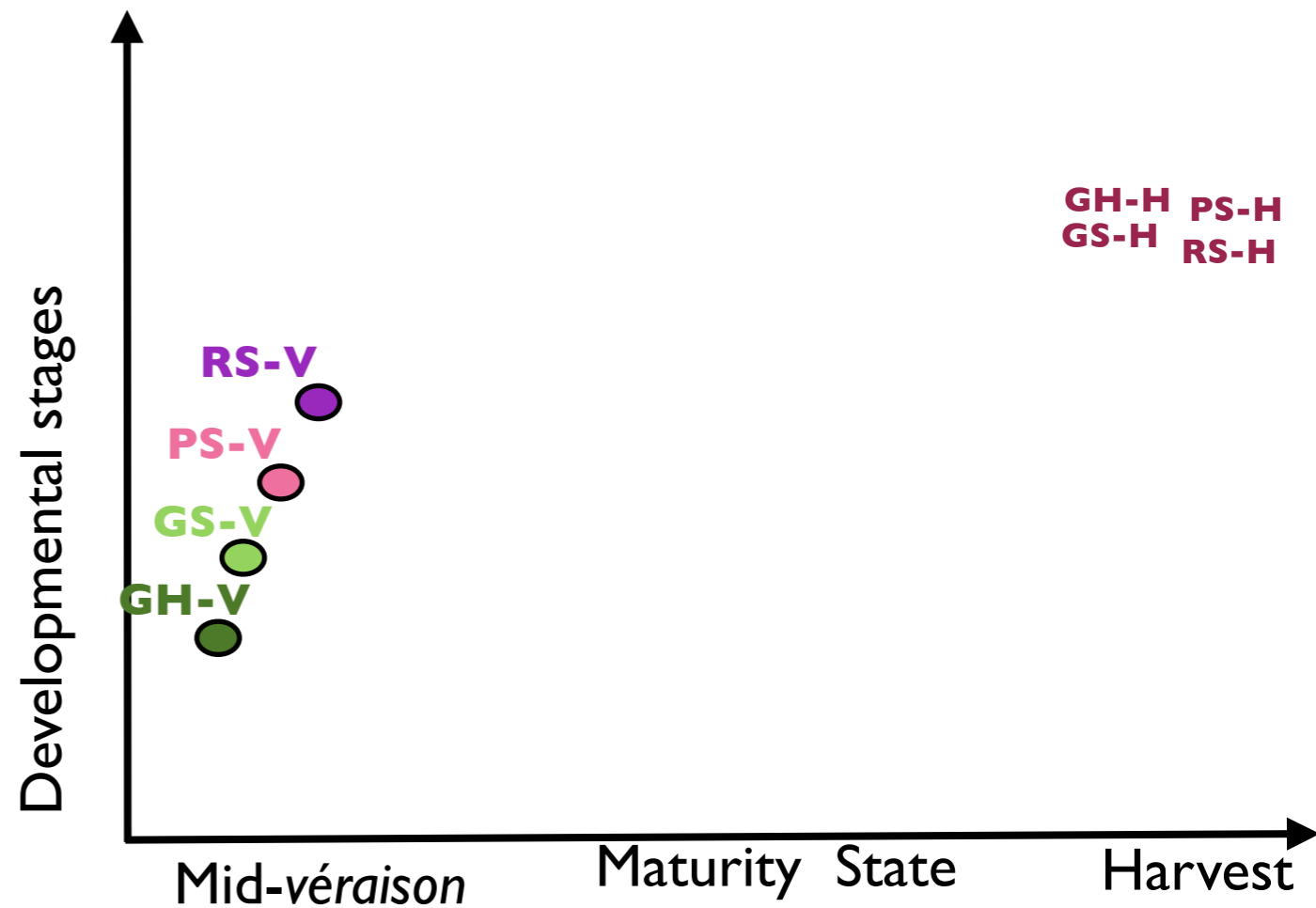






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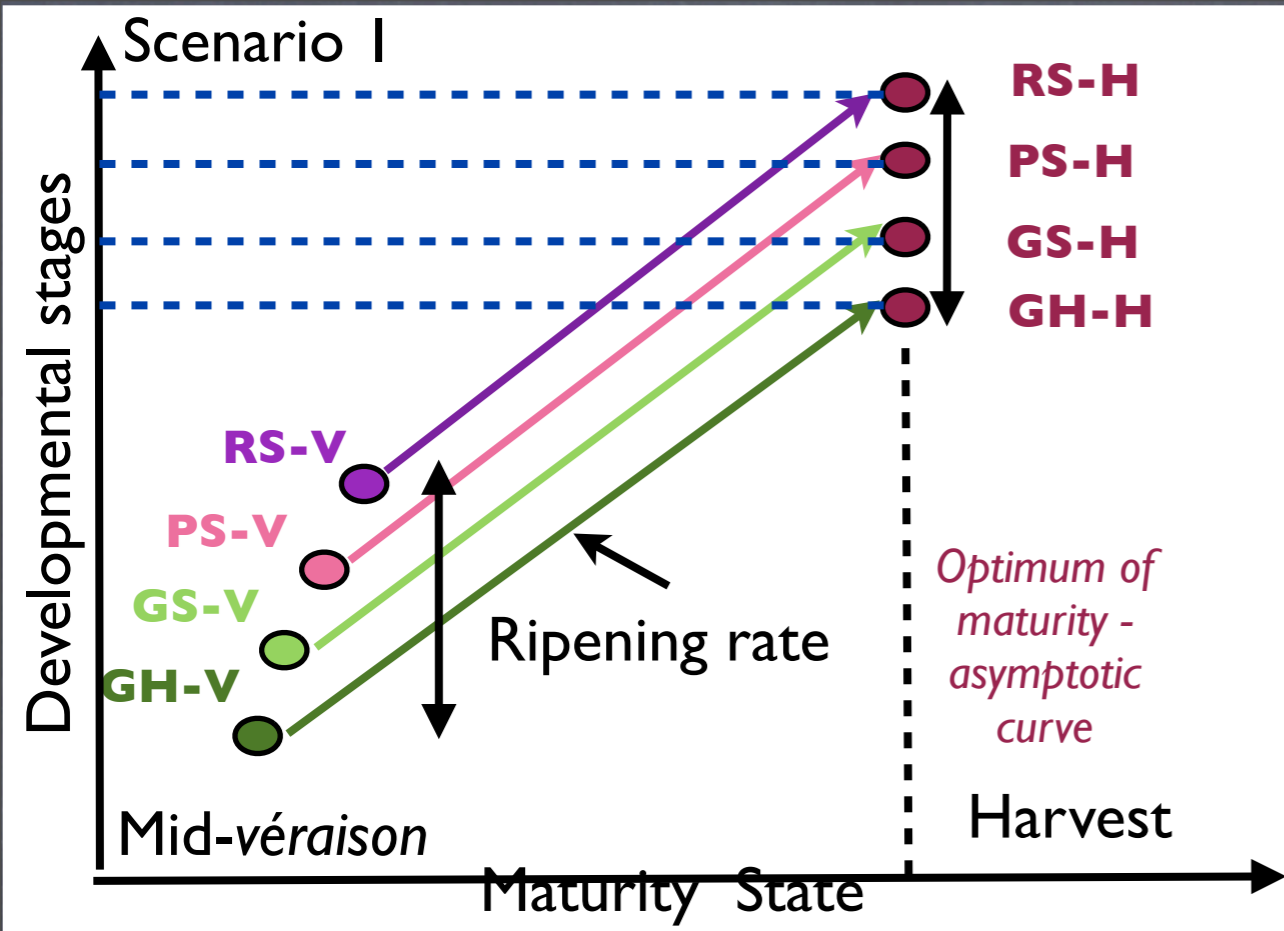


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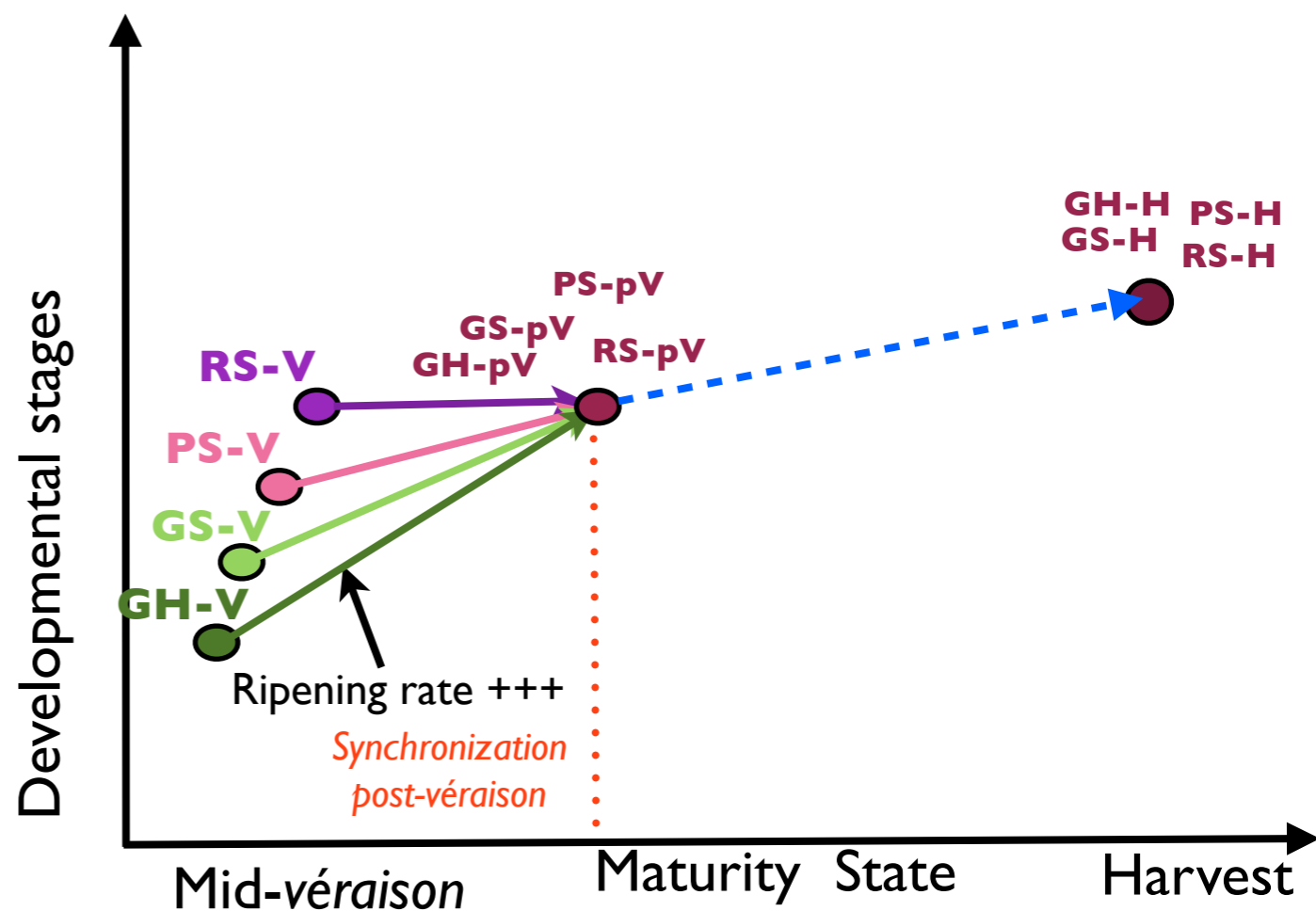




If we assume differences at mid-*véraison*, we must still see differences near harvest at the developmental level

If under ripe berries are catching up we do not know when that happens....

**Scenario 2**



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# Questions

***Is there a reduction of berry developmental variability at harvest in Pinot Noir?***

***What is the contribution of each berry class to the overall fruit quality?***

***What is the mechanism ?***

---

***Appropriate way to correlate berry classes and a berry developmental stage?***

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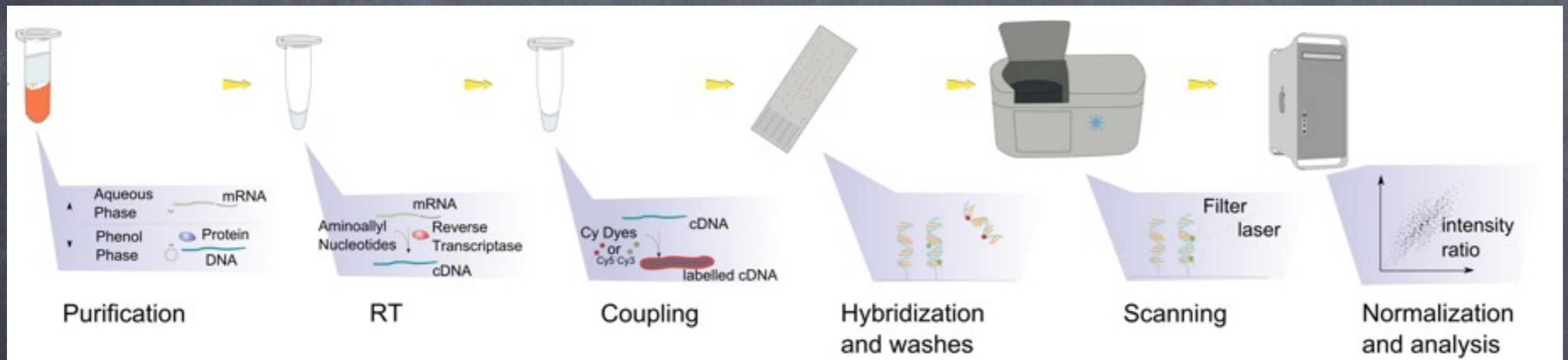


Thursday, August 25, 2011

# How describe a developmental stage?

Using genomic hybridization or sequencing technology as fingerprint to describe a given developmental stage or a tissue

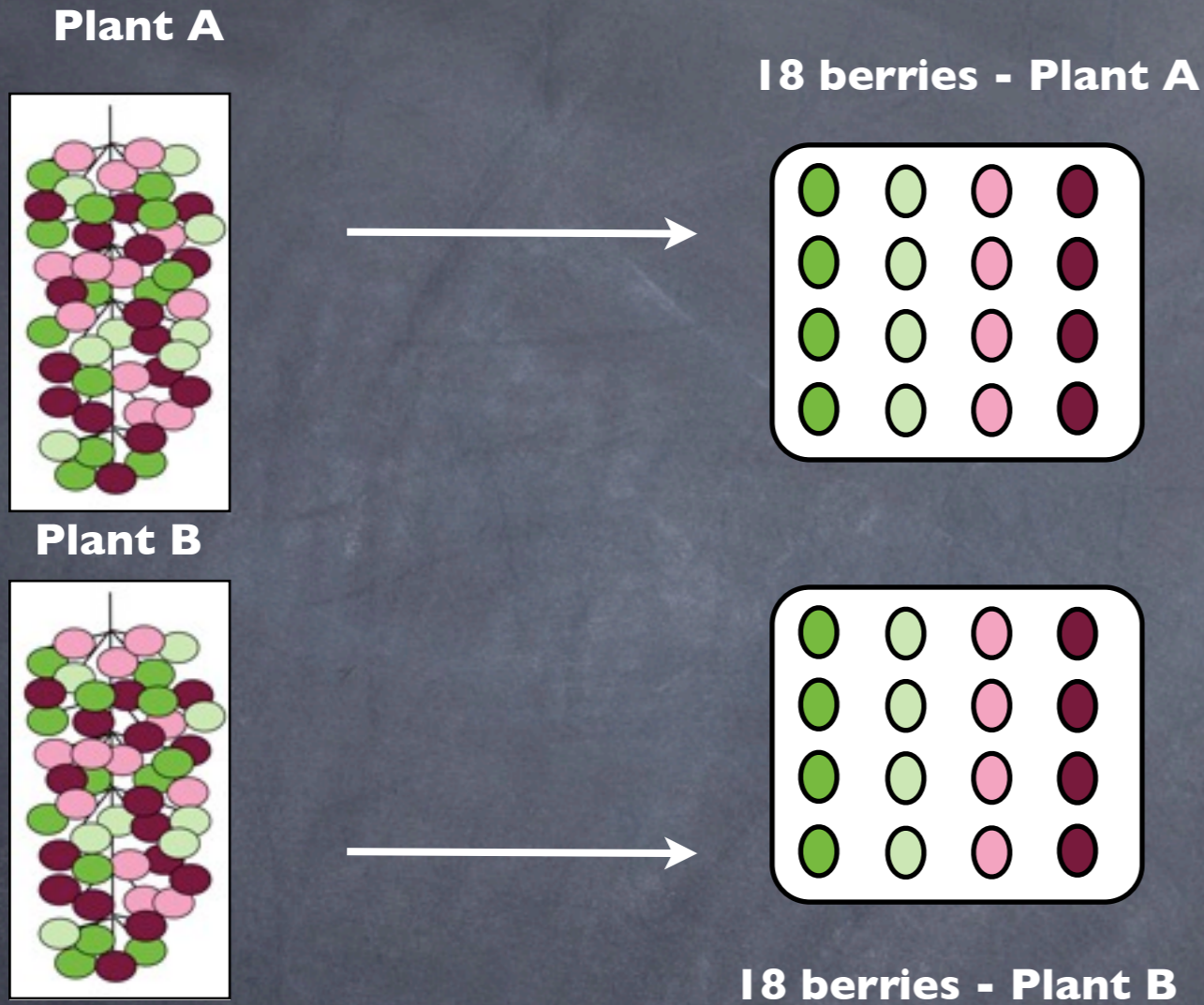
## DNA microarray experiment



Generate the most comprehensive snapshot at the gene expression level to describe a particular stage



# A global gene expression profile reflects a developmental stage (Lund et al., 2008)



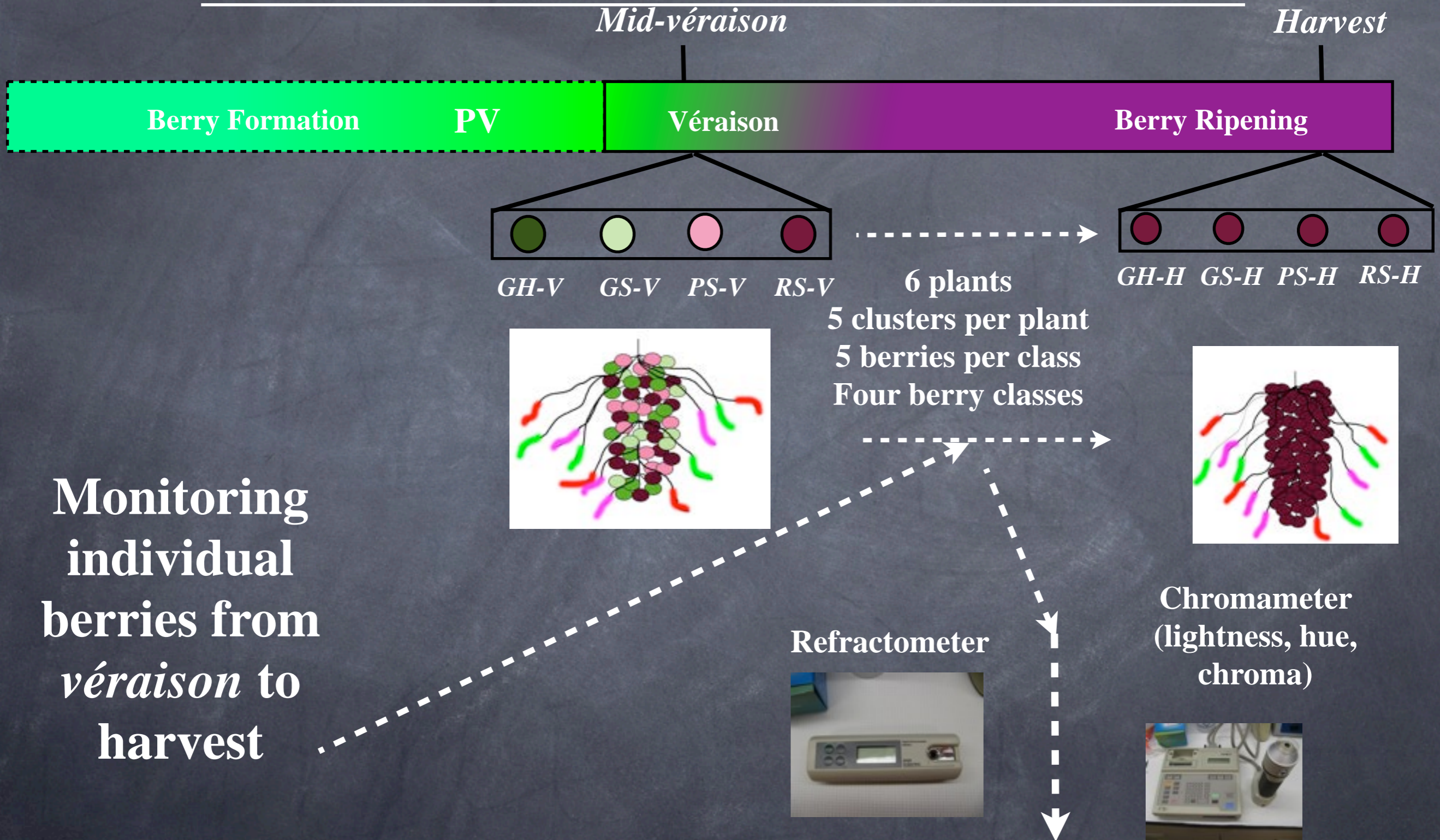
Significance test (global transcript profile versus phenotypic variables)

Phenotypic variable	Significance (p<0.05)
Plant	0.72
Position in the cluster	0.54
Diameter	0.18
Weight	0.10
Soluble Solids	0.04
Color + Firmness	0.08
Firmness	0.07
Color	0.03

Berries from same class have similar global transcript profile signatures

Global transcript signatures are significantly correlated to some phenotypic variables

# Phenological measurements of individual berries at mid-véraison and harvest

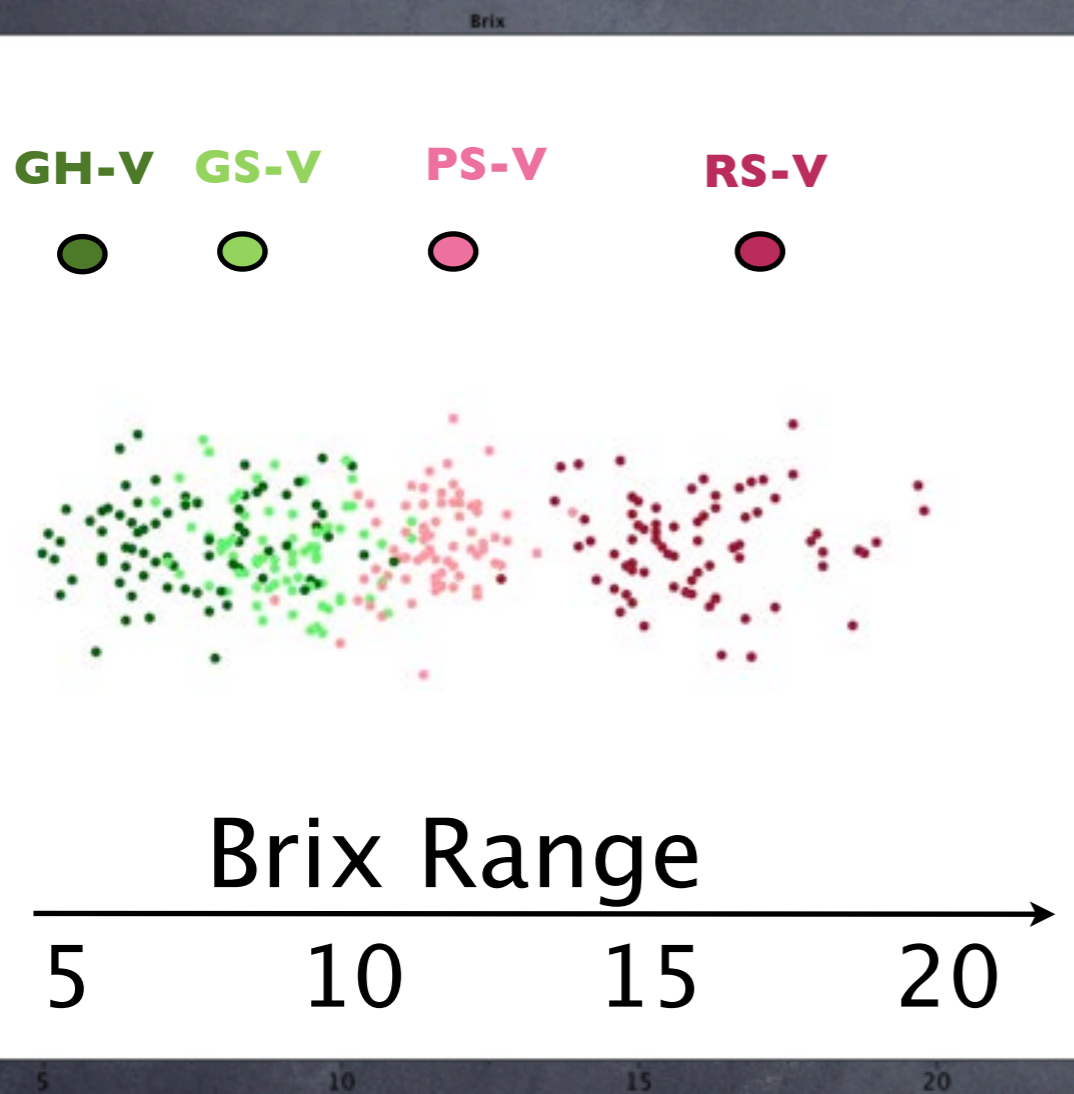


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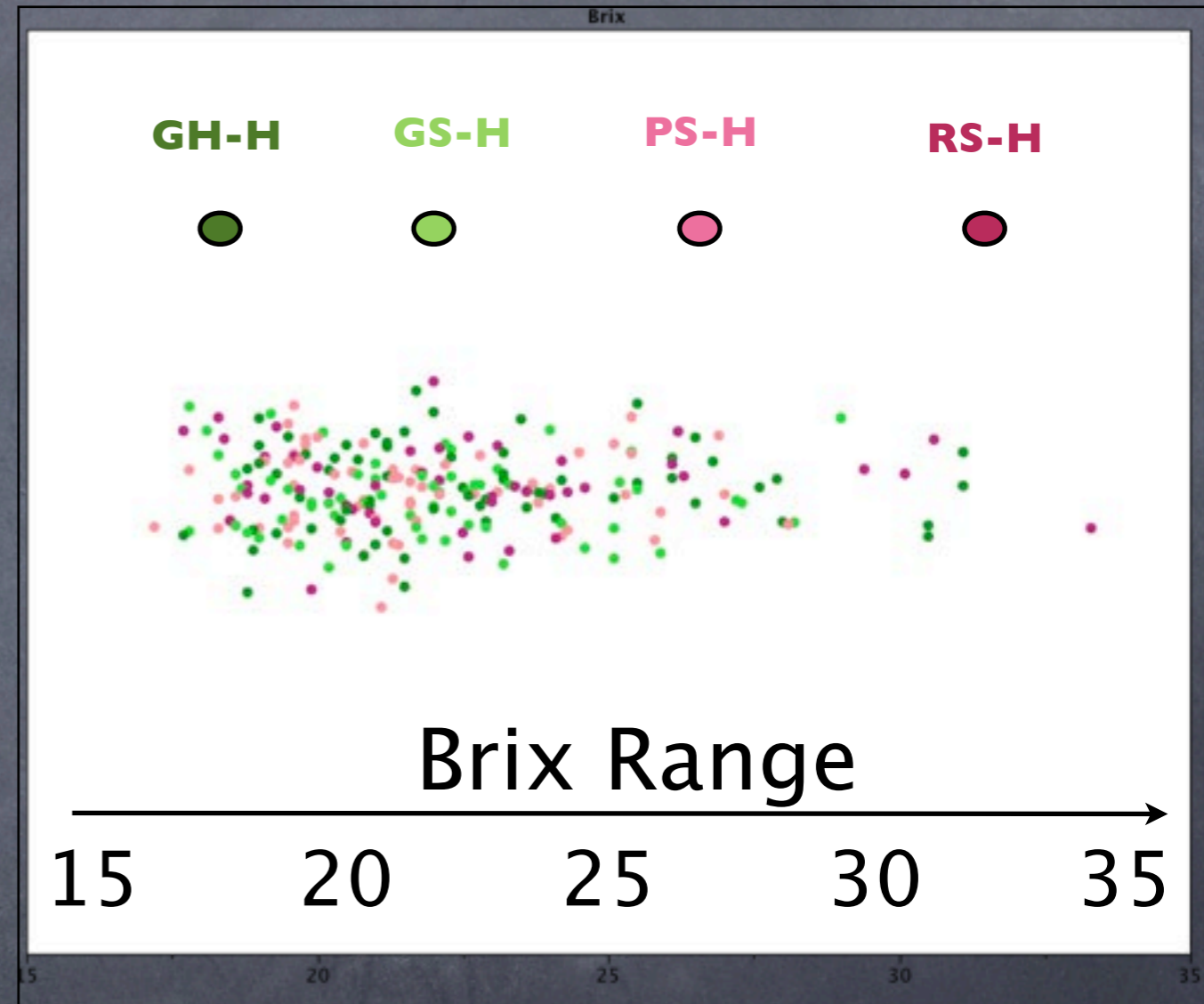


# Brix values of individual at mid-véraison and harvest

## Mid-véraison:



## Harvest:



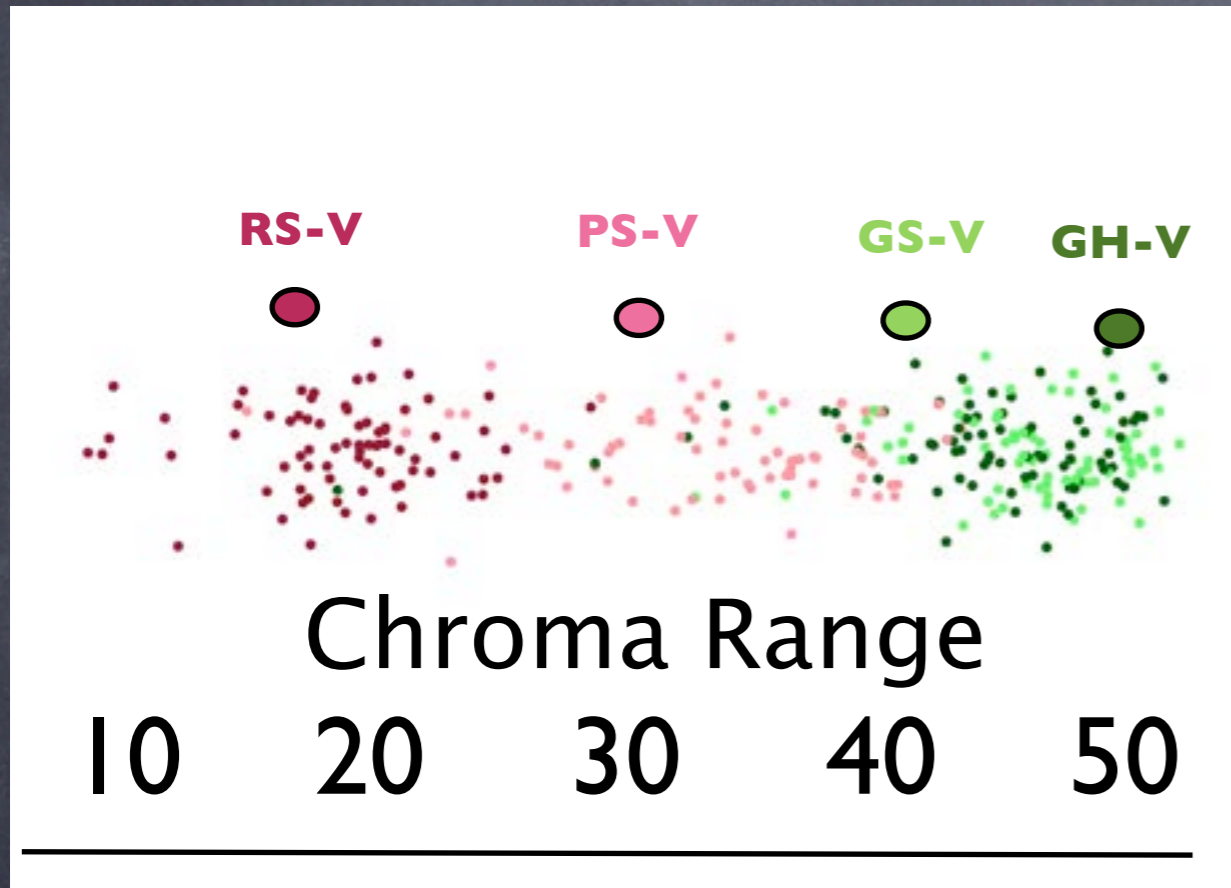
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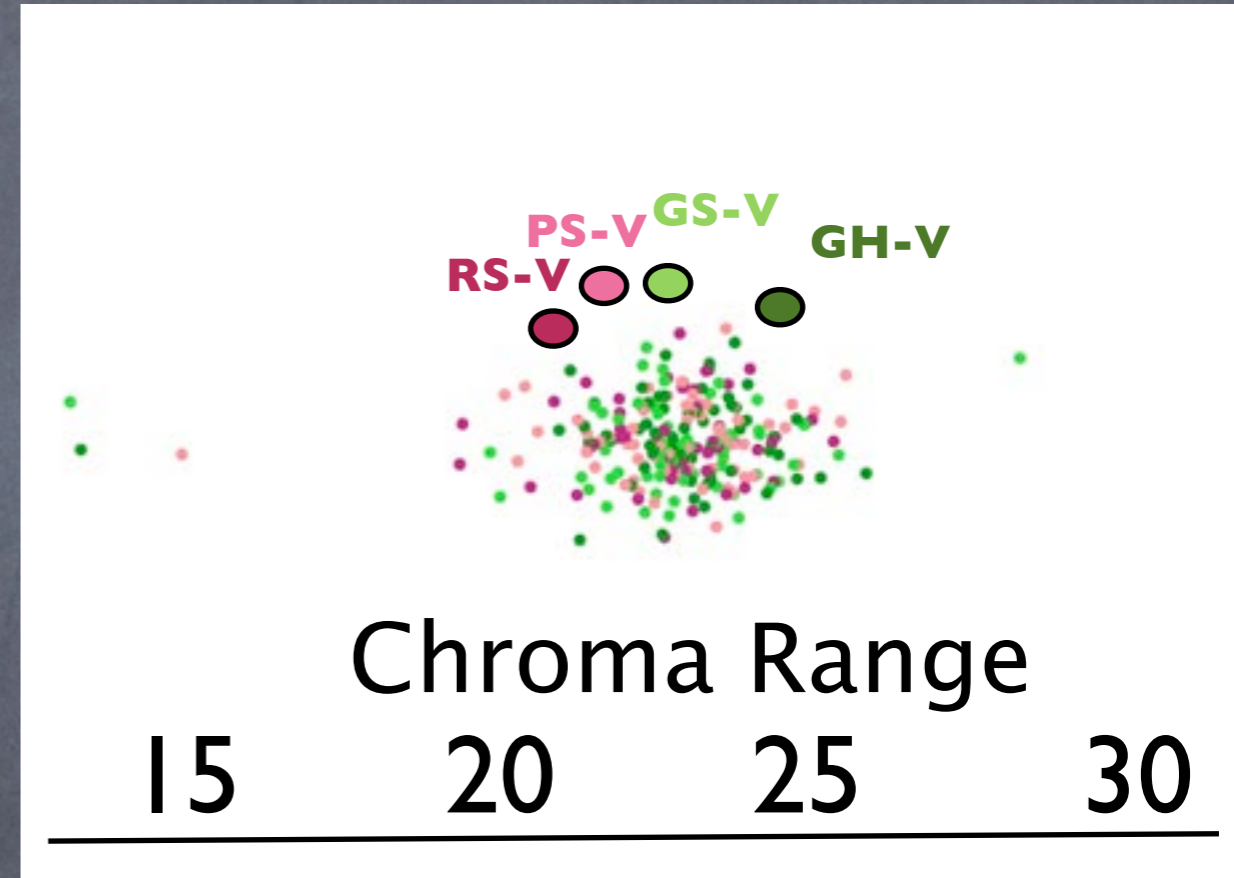


# Chroma values of individual at mid-véraison and harvest

Mid-véraison:



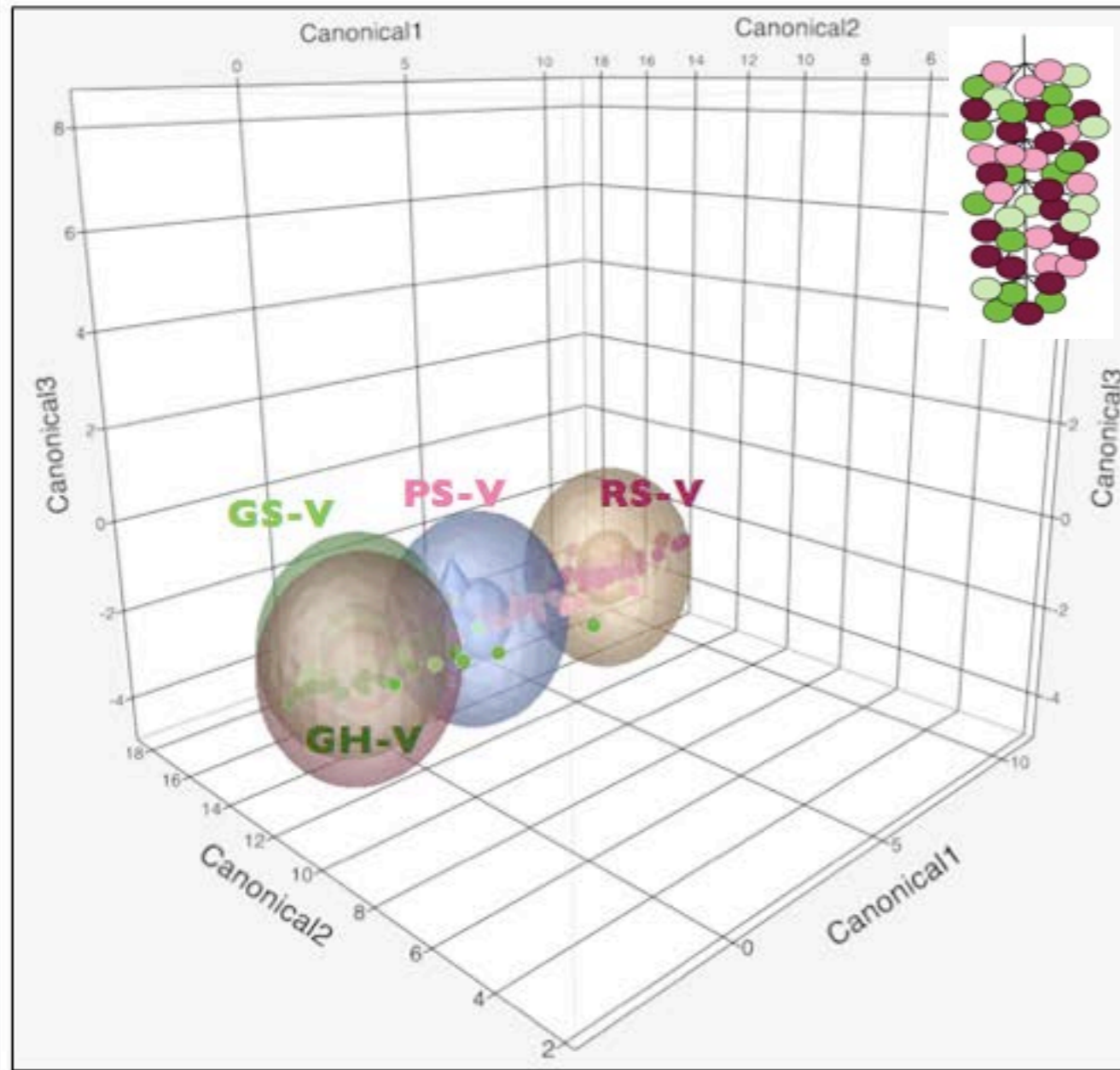
Harvest:



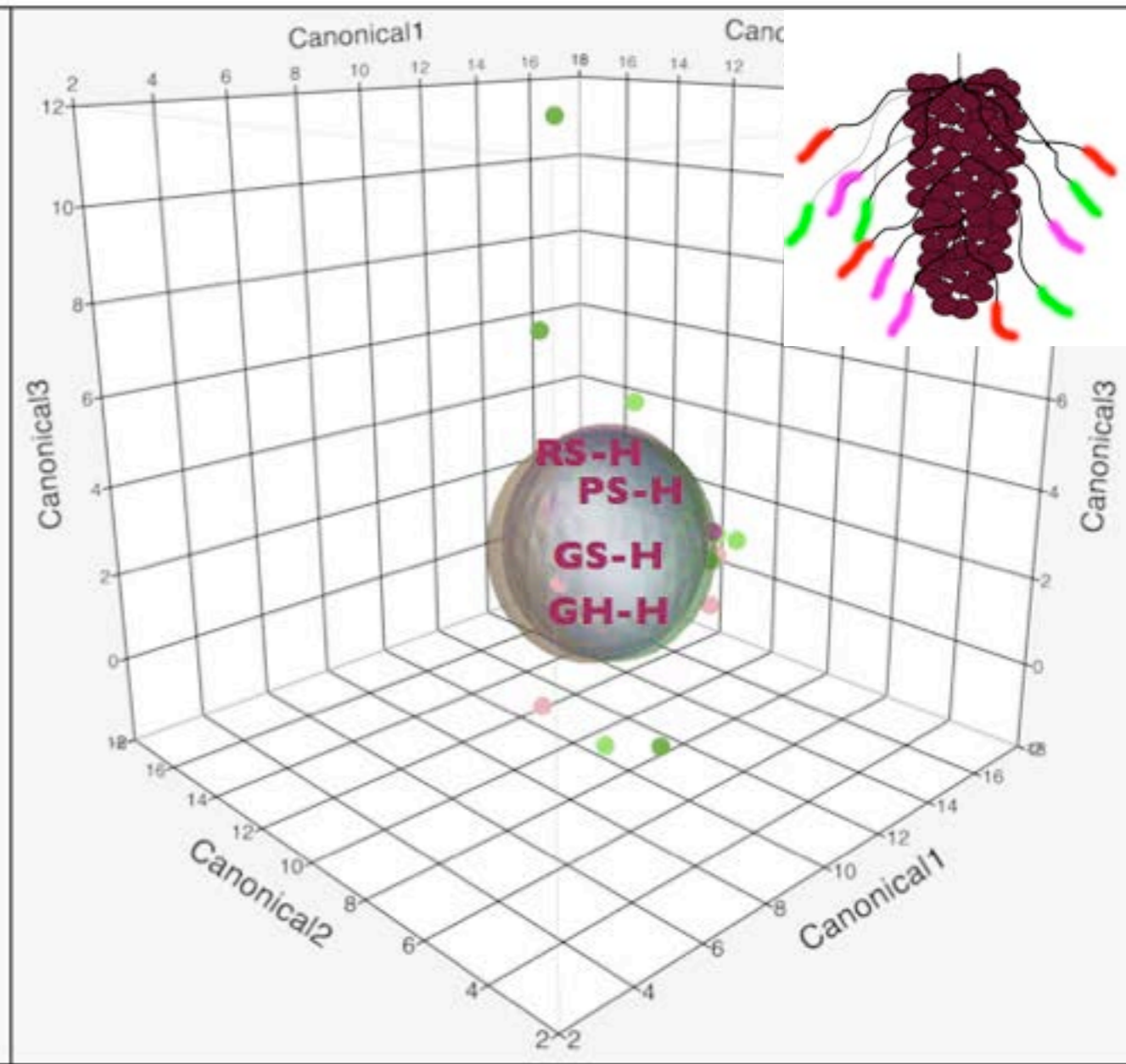


# Discriminant Analyses using Brix, lightness, hue and chroma

## Mid-véraison



## Harvest

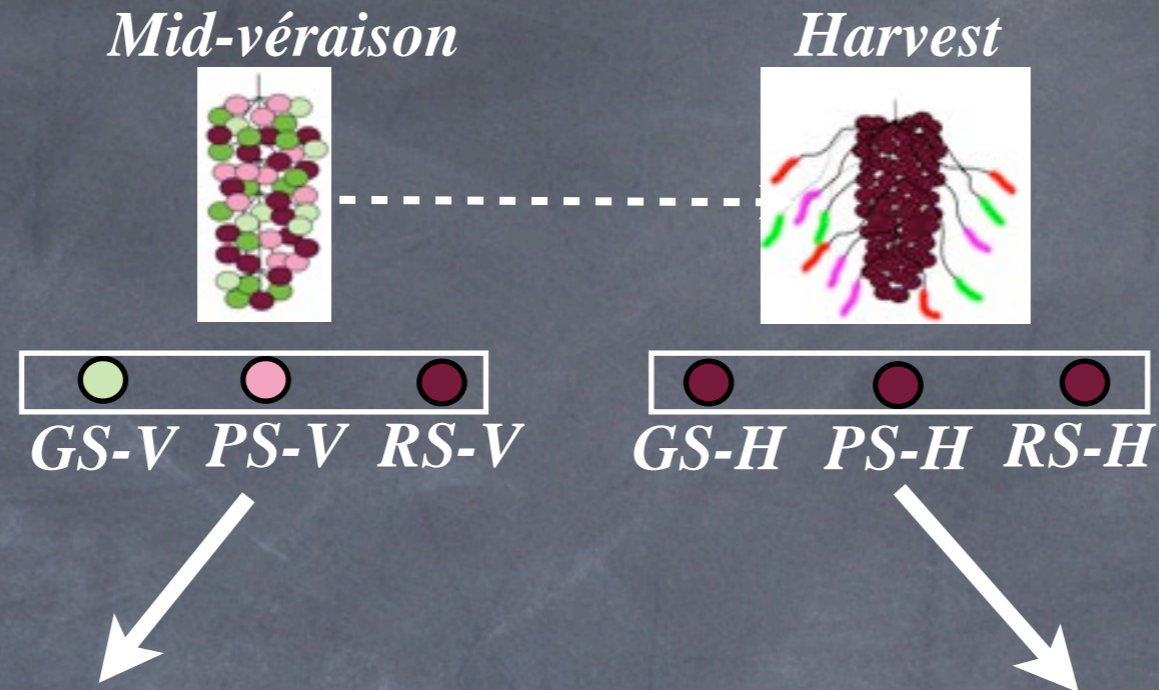


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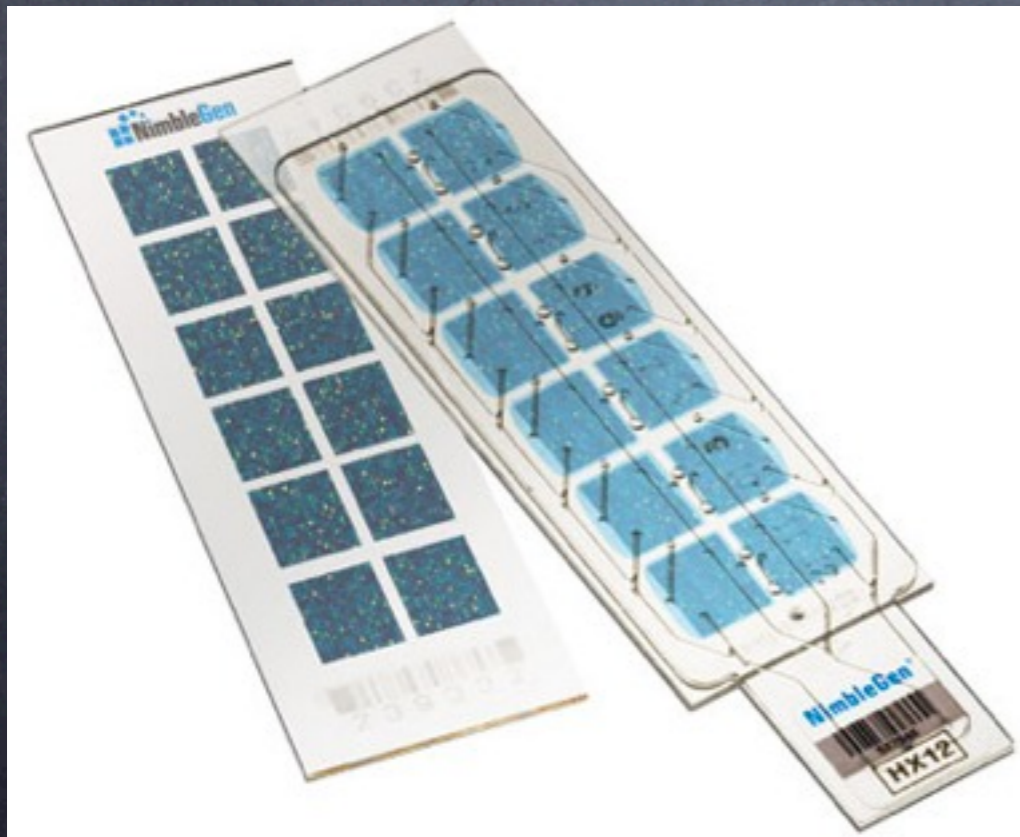


# Transcriptomic experiment (genome hybridization)



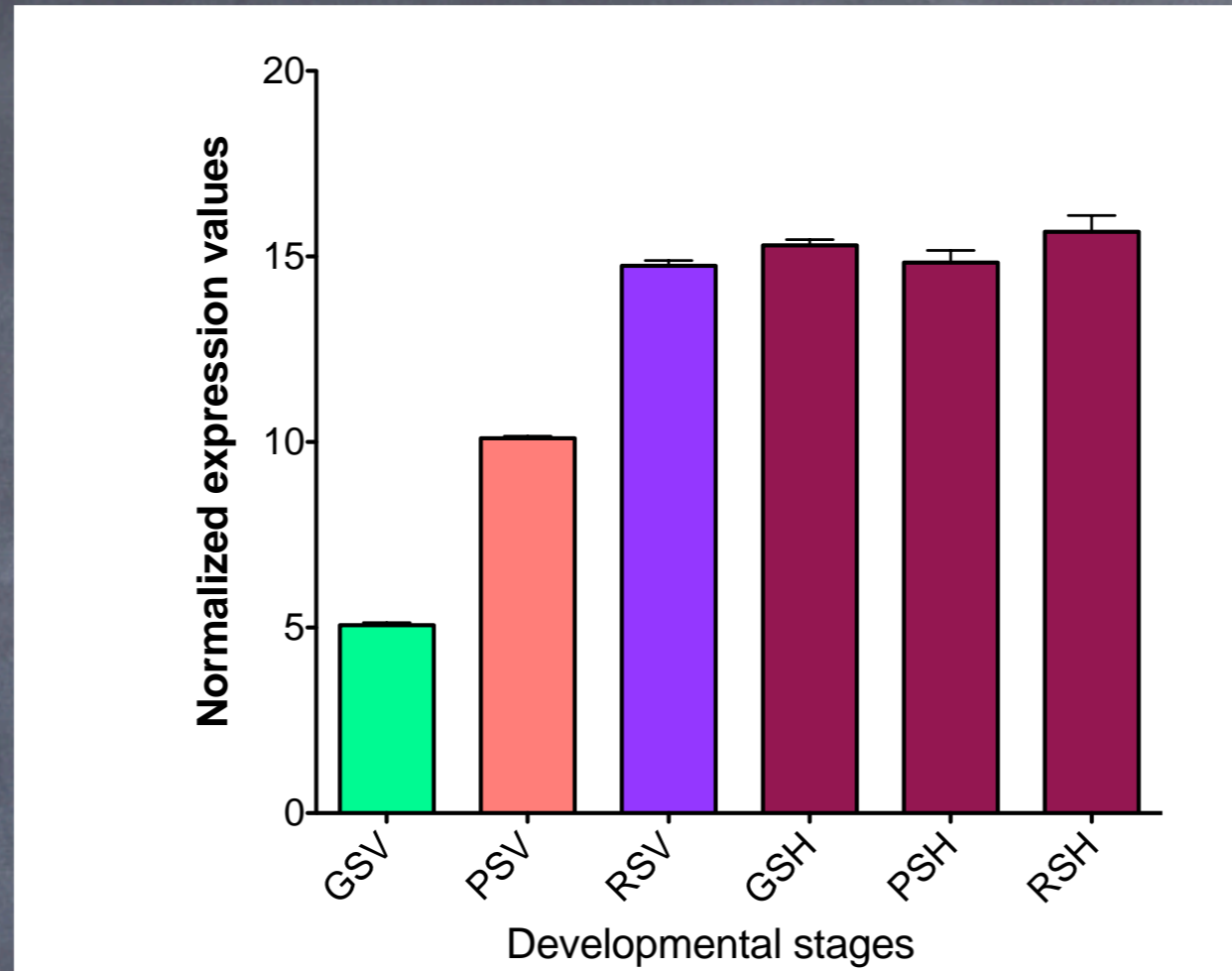
*27 arrays ( 3 tissues, 3 plants, 3 berry classes)*

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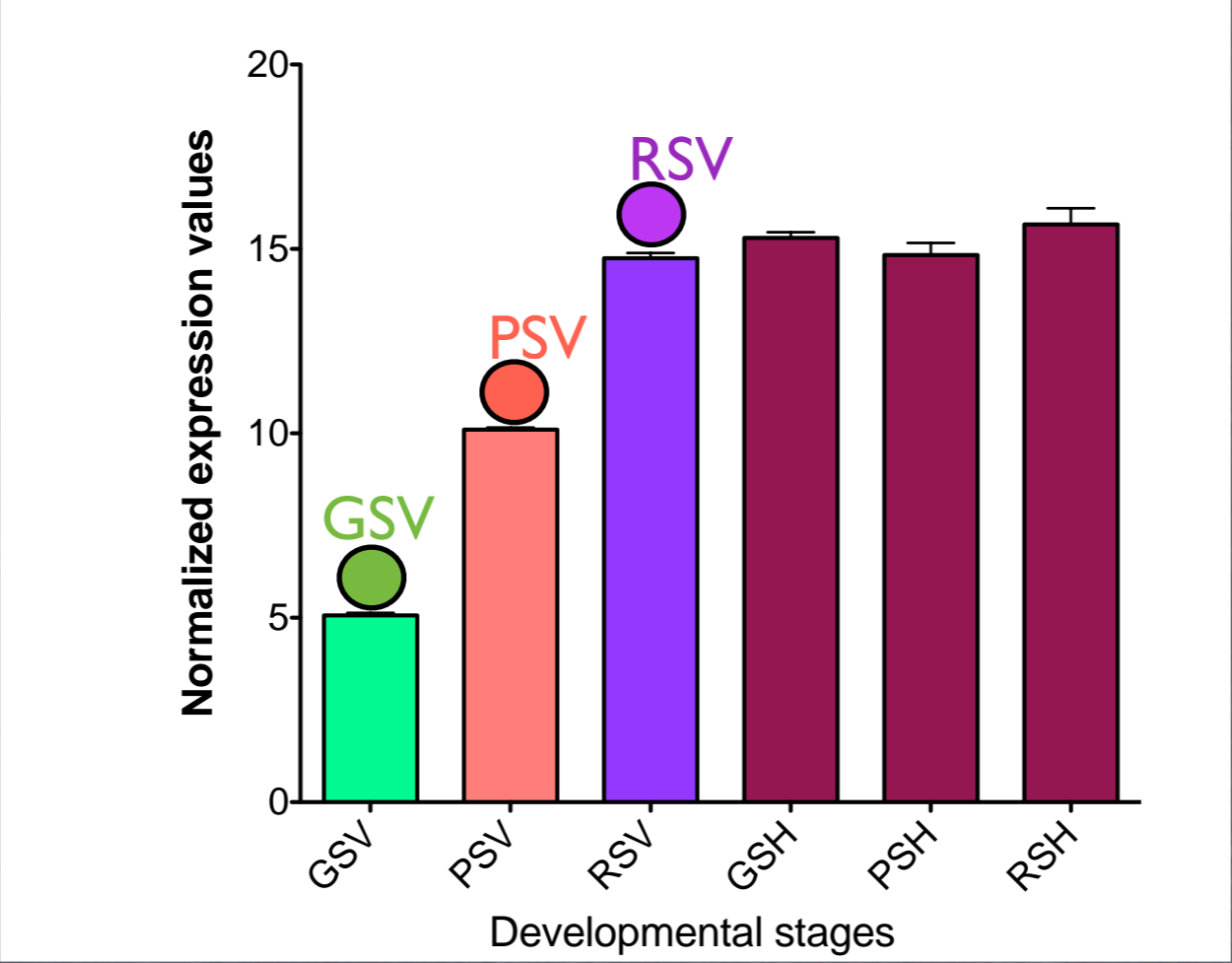
Nimblegen Grape whole transcriptome ~29,550 genes

# Transcript profile of a UFGT gene



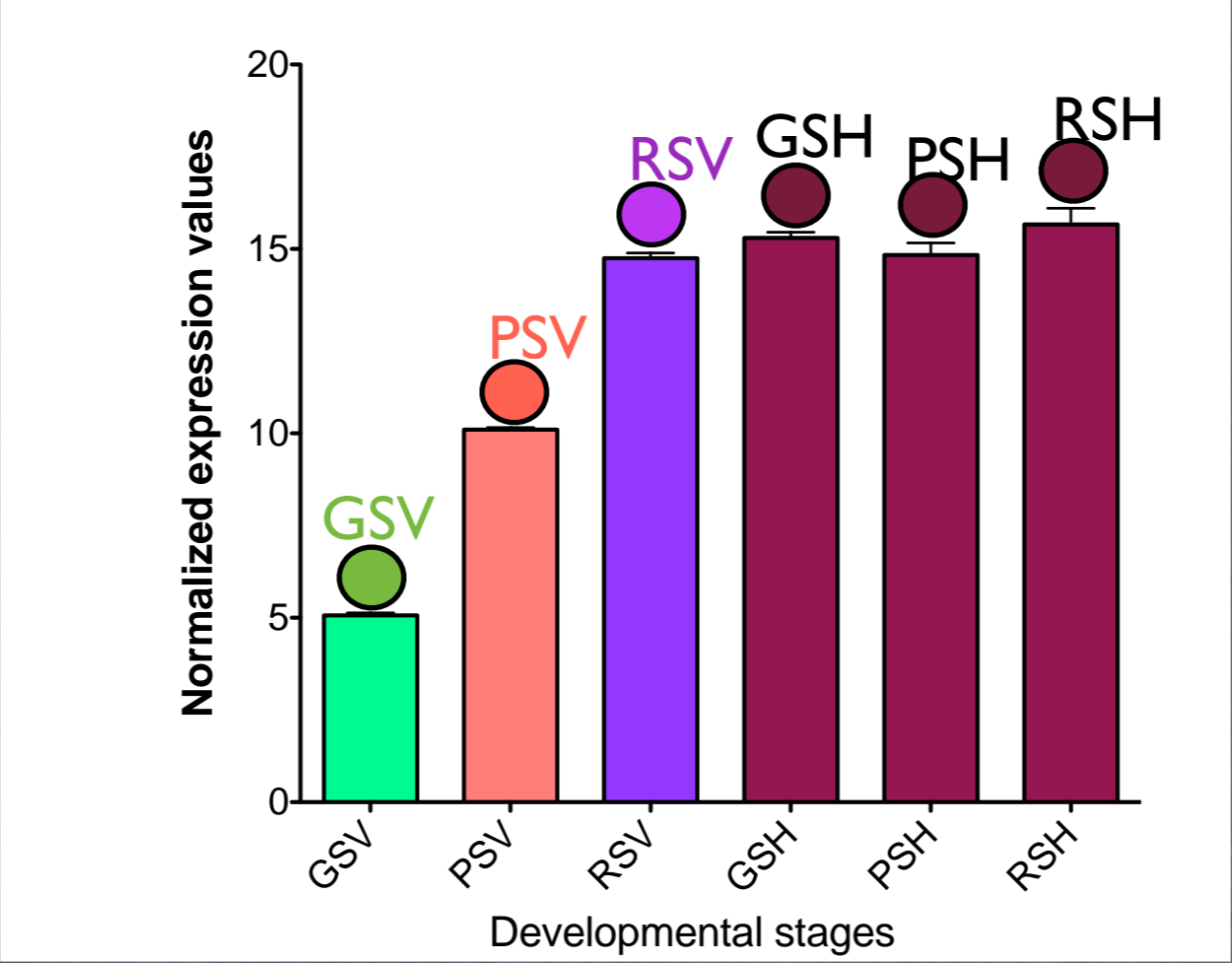
For 29,550 genes

# Transcript profile of a UFGT gene



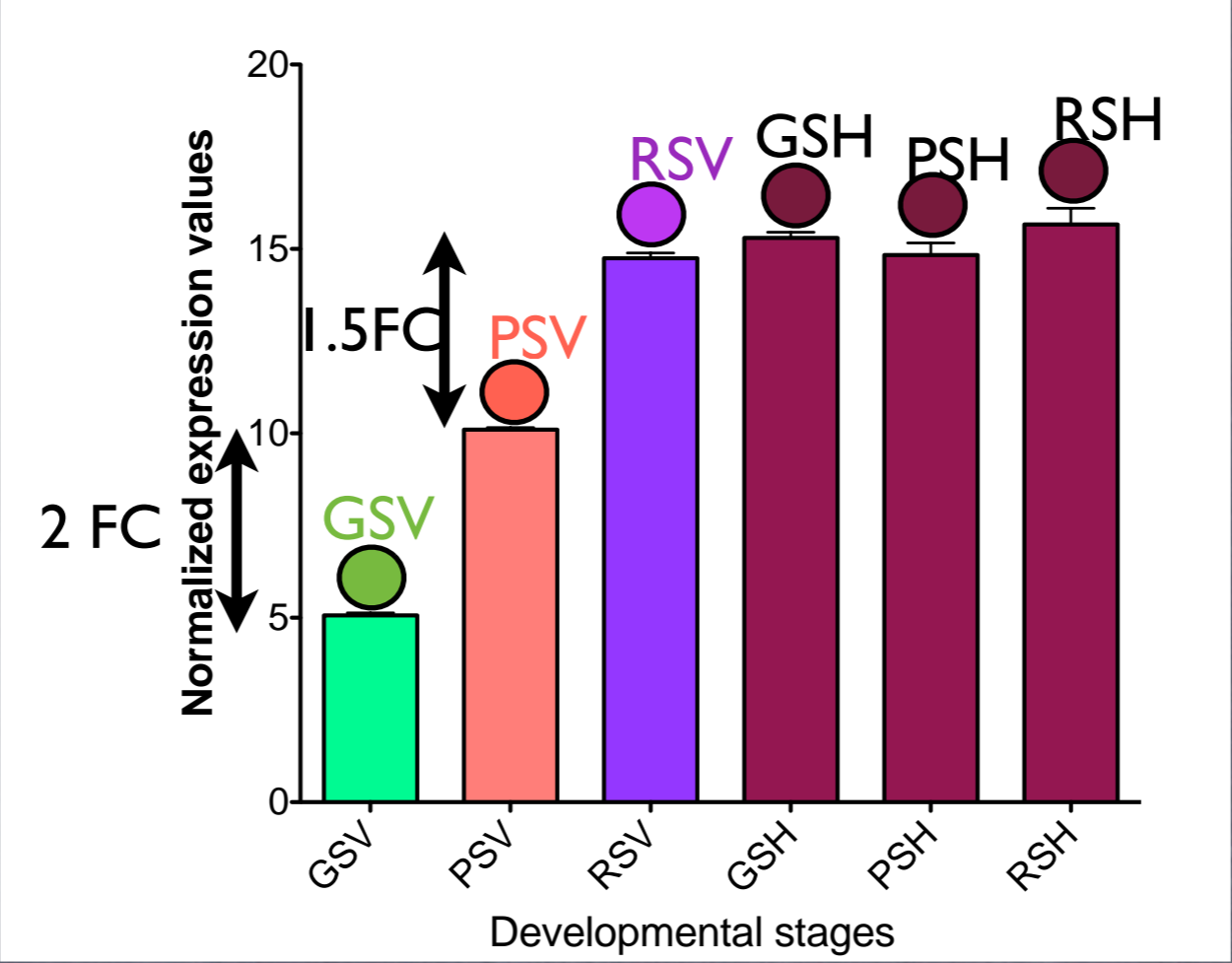
For 29,550 genes

# Transcript profile of a UFGT gene



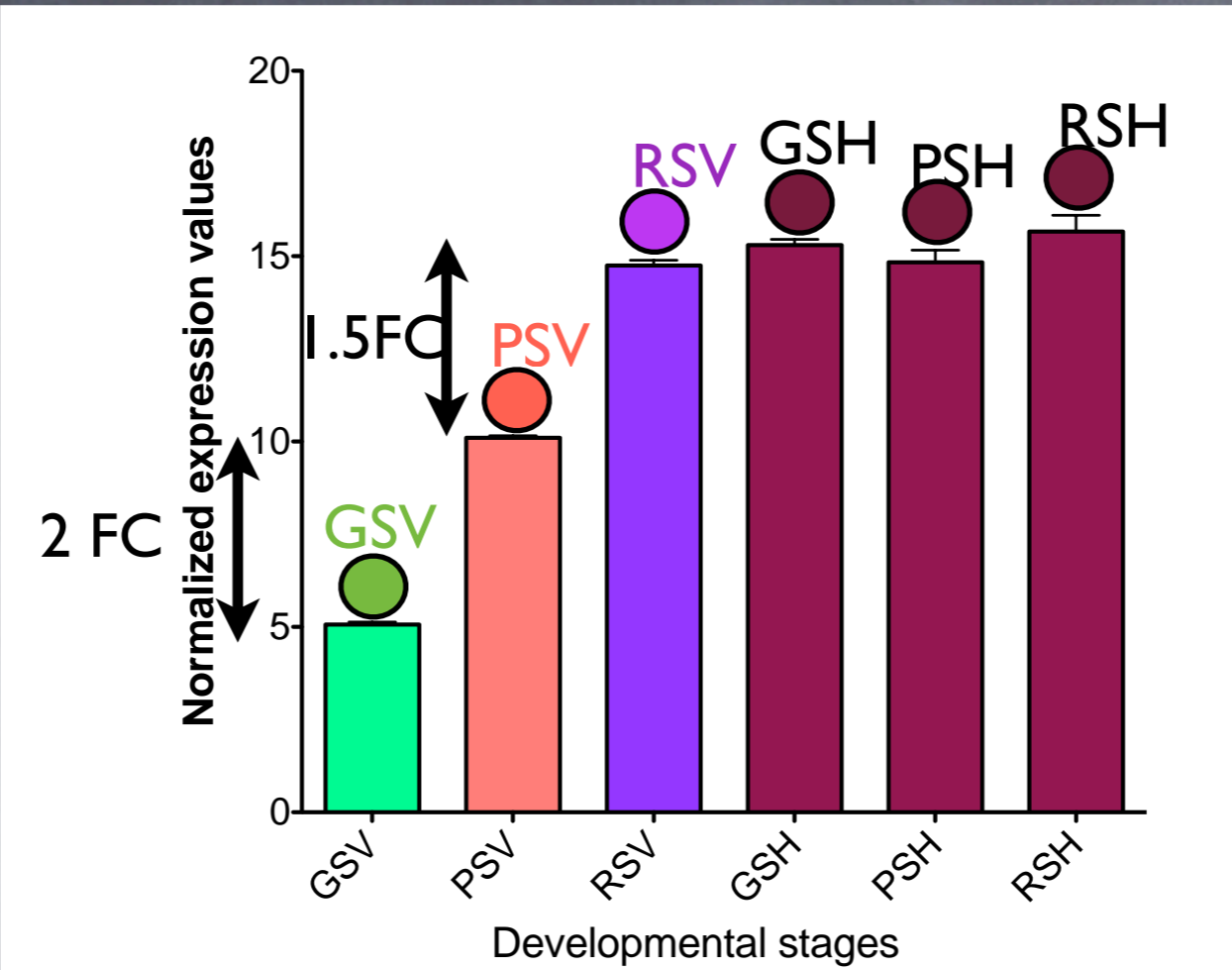
For 29,550 genes

# Transcript profile of a UFGT gene



For 29,550 genes

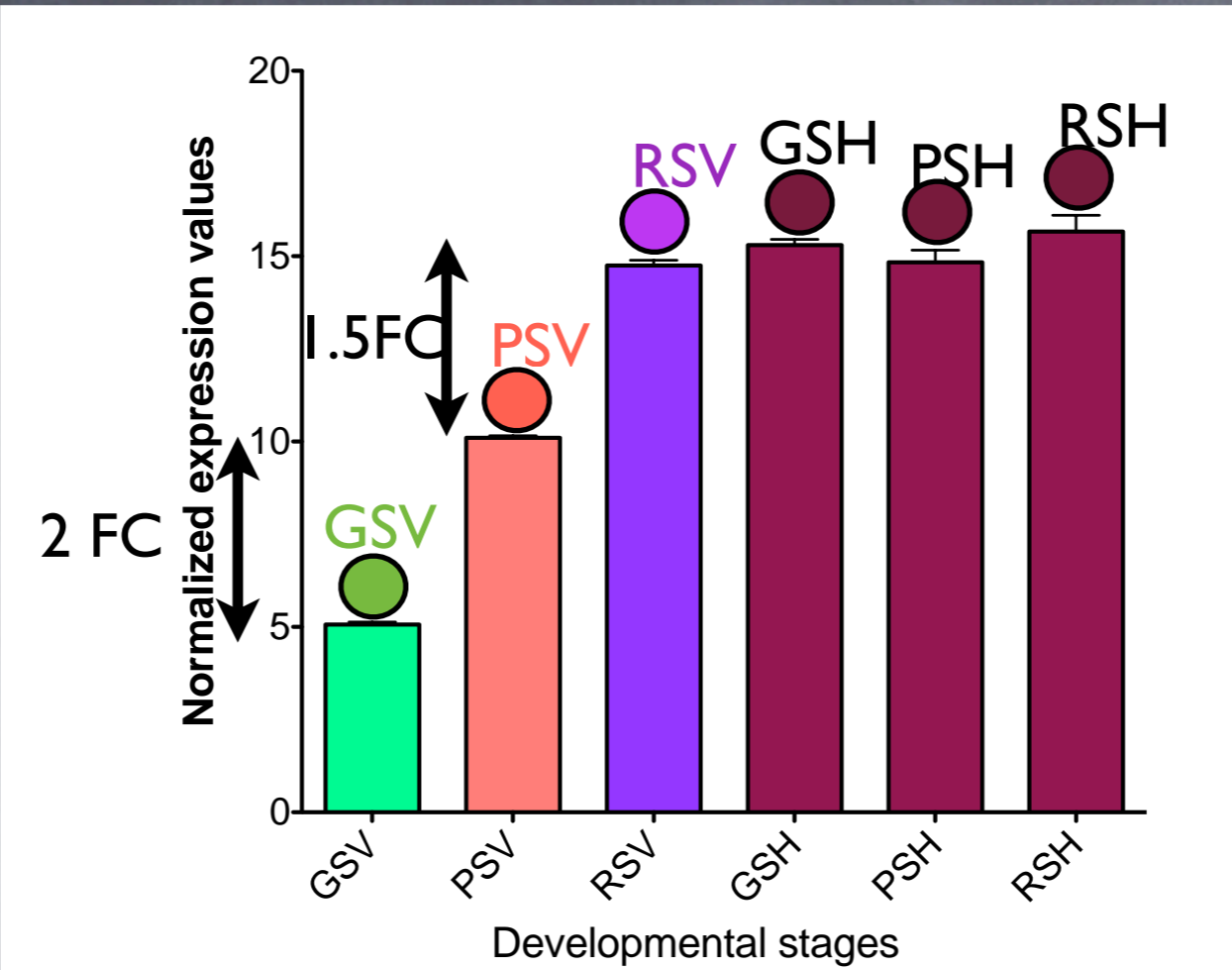
# Transcript profile of a UFGT gene



For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
Biological significance												
Genes	#	%	#	%	#	%	#	%	#	%	#	%
Pulp	4,393	15	2,180	7	1,266	4	1,060	4	200	0.6	60	0.2
Seed	3,102	10	1,416	5	781	3	1,004	4	327	1.1	116	0.4
Skin	3,986	13	1,797	6	998	3	1,748	6	535	1.8	201	0.7

Transcript profile of a UFGT gene

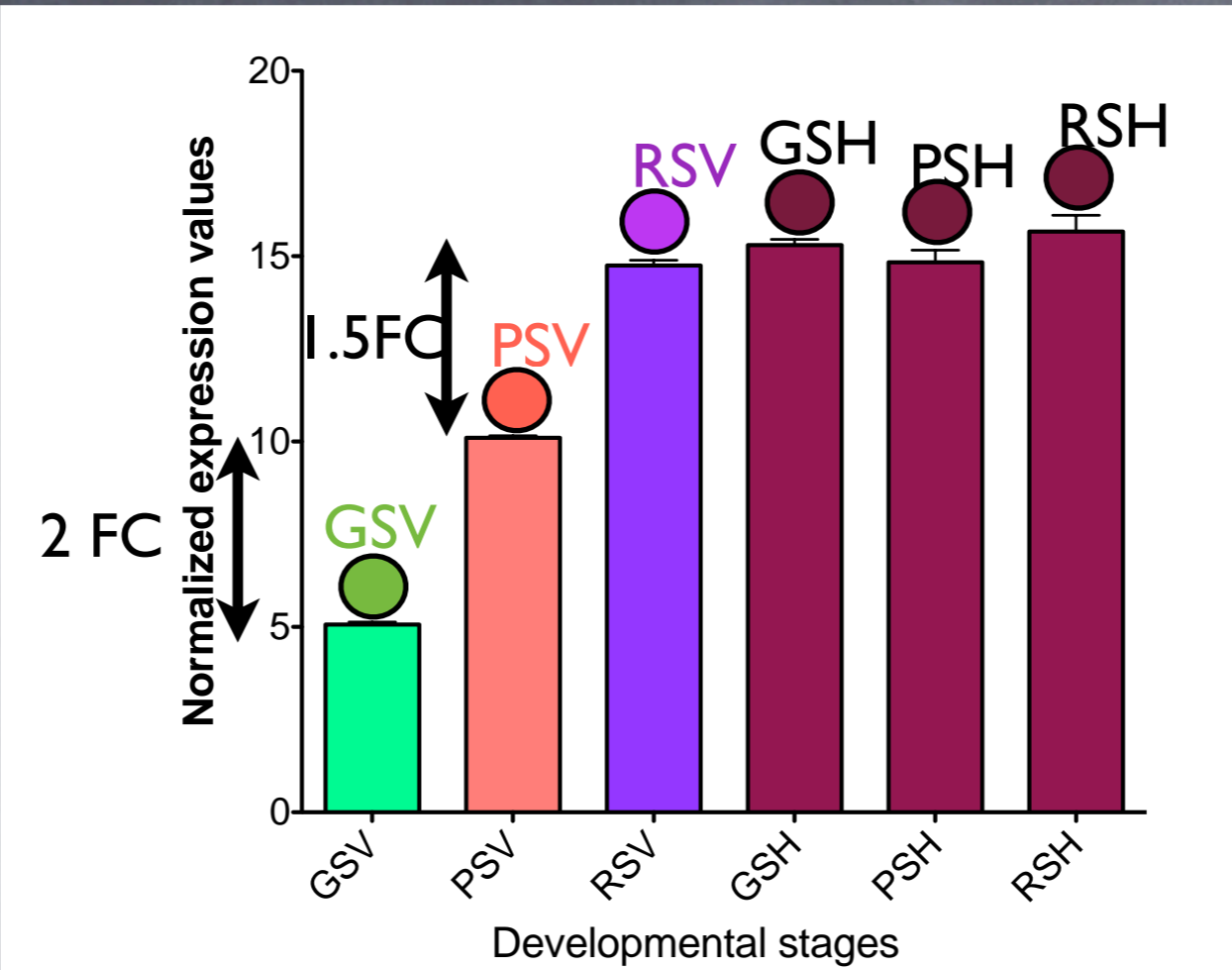


For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
Biological significance	#	%	#	%	#	%	#	%	#	%	#	%
<b>Genes</b>	#	%	#	%	#	%	#	%	#	%	#	%
<b>Pulp</b>	4,393	15	2,180	7	1,266	4	1,060	4	200	0.6	60	0.2
<b>Seed</b>	3,102	10	1,416	5	781	3	1,004	4	327	1.1	116	0.4
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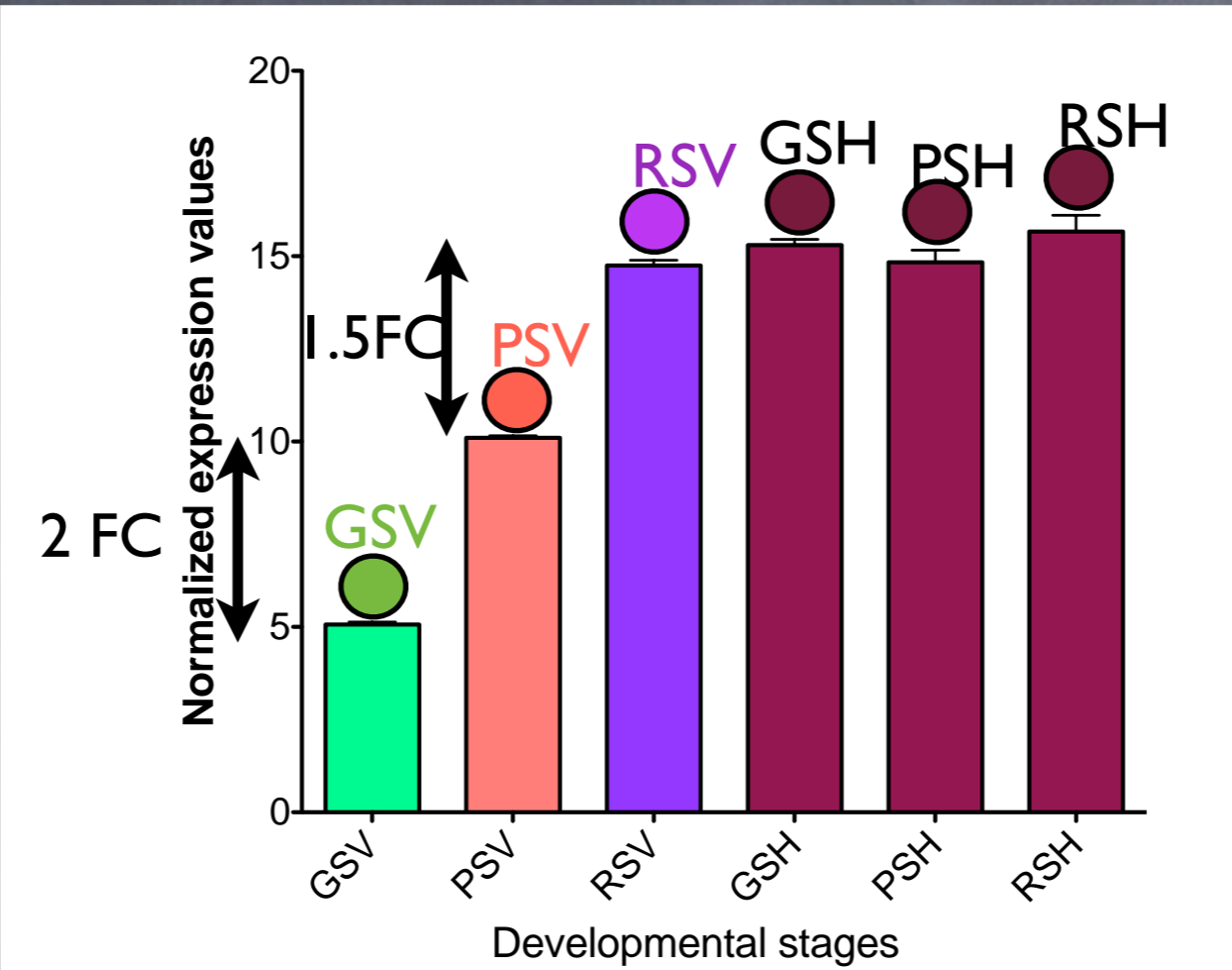
# Transcript profile of a UFGT gene



For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
Biological significance												
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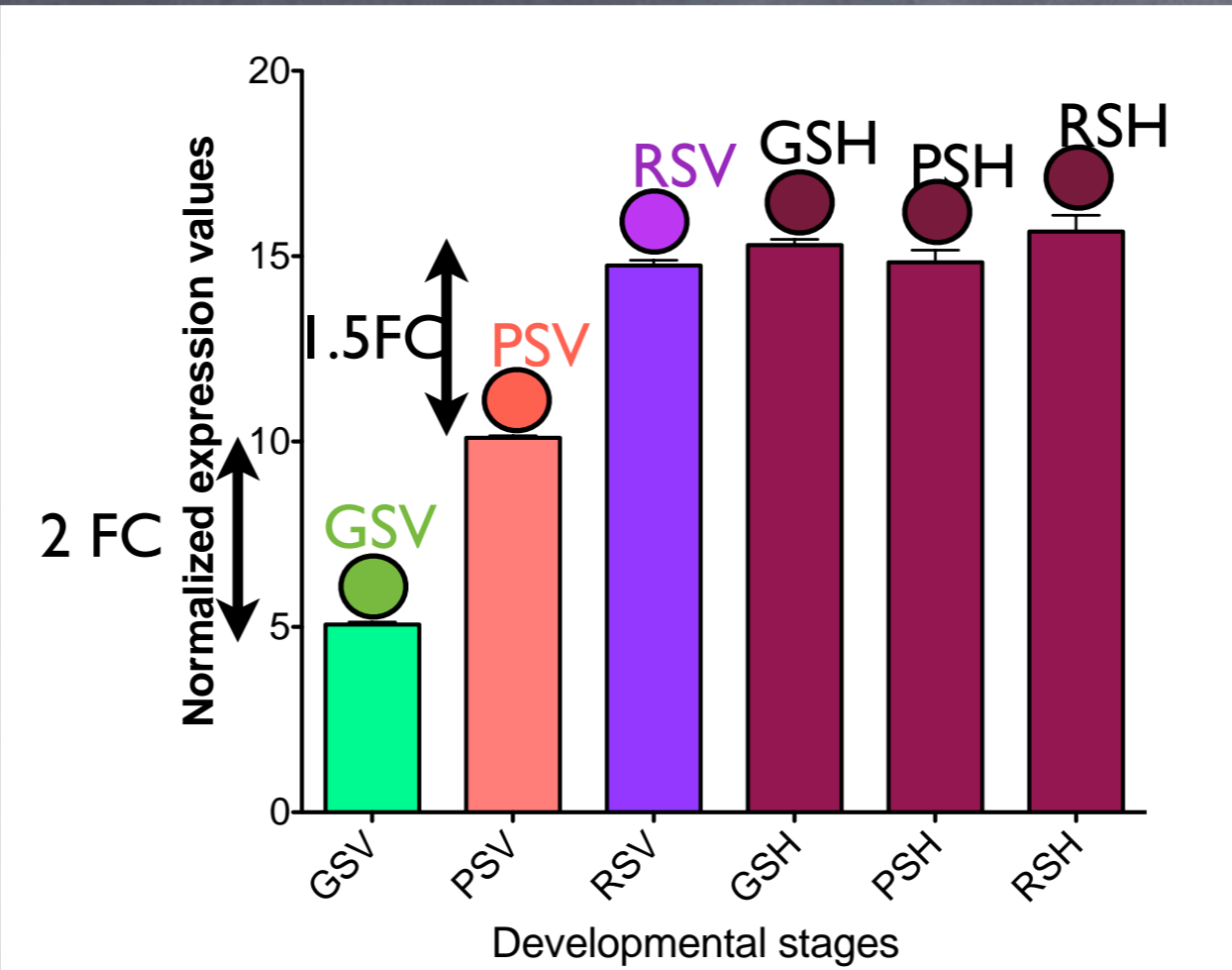
Transcript profile of a UFGT gene



For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
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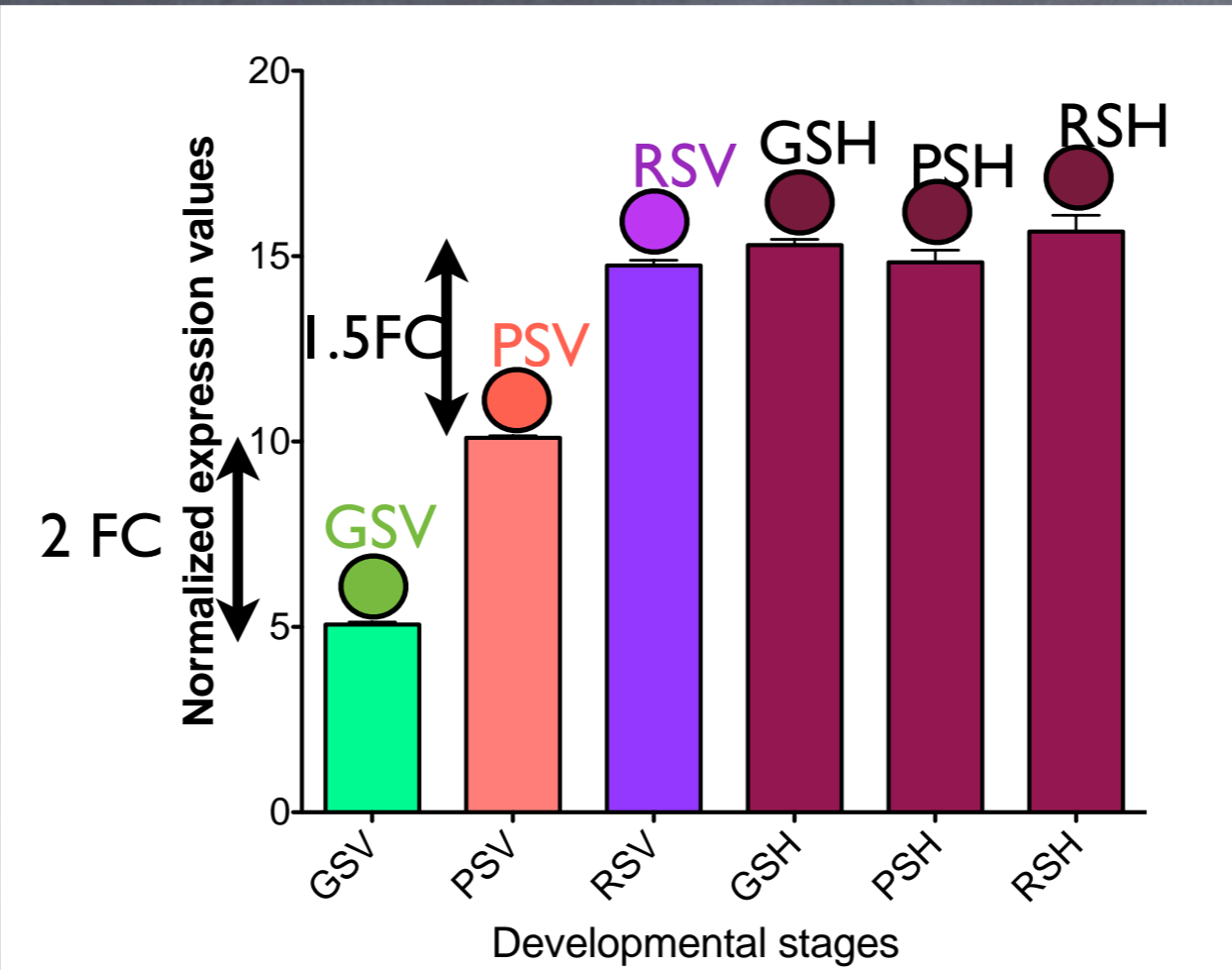
Transcript profile of a UFGT gene



For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
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Pulp	4,393	15	2,180	7	1,266	4	1,060	4	200	0.6	60	0.2
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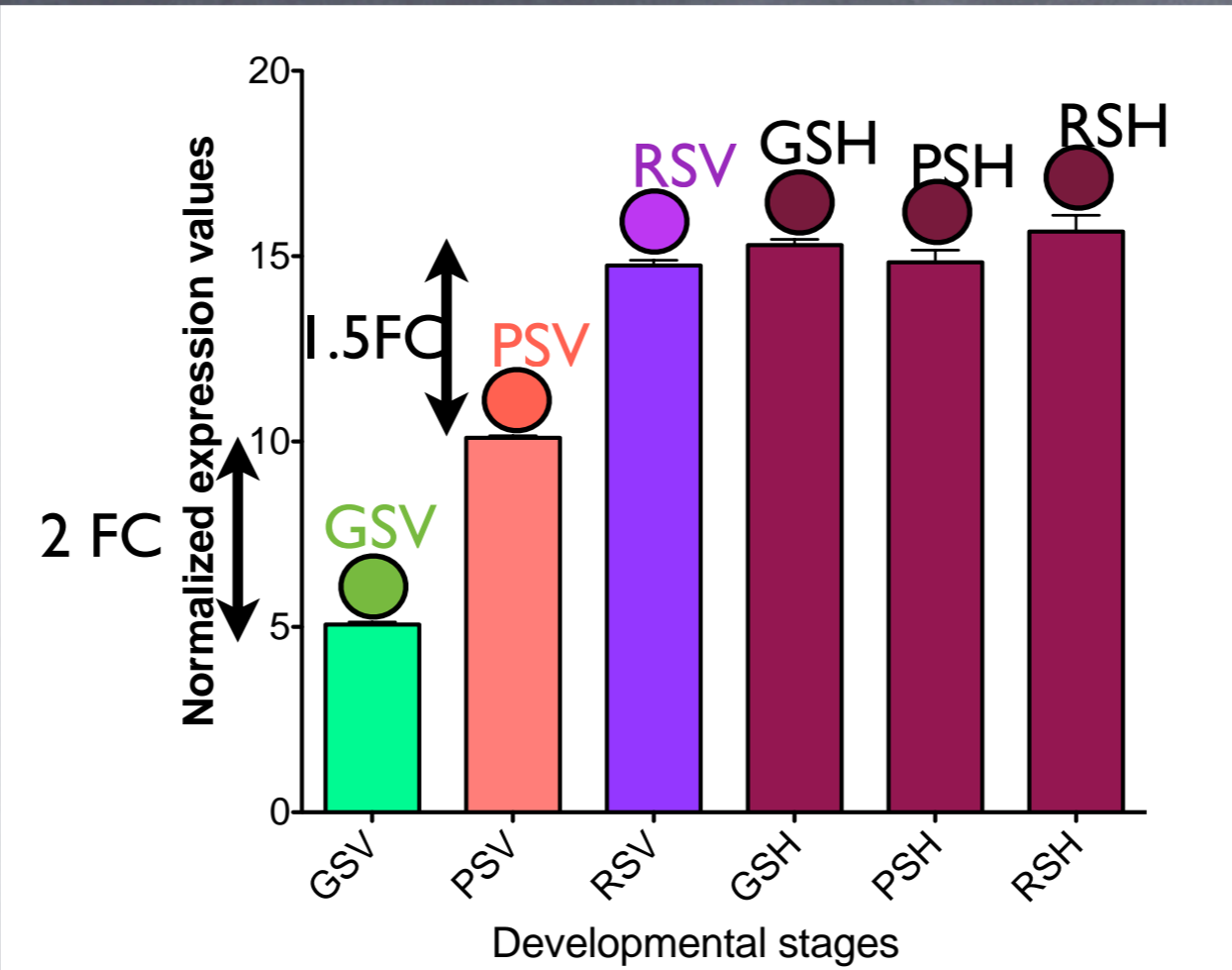
Transcript profile of a UFGT gene



For 29,550 genes

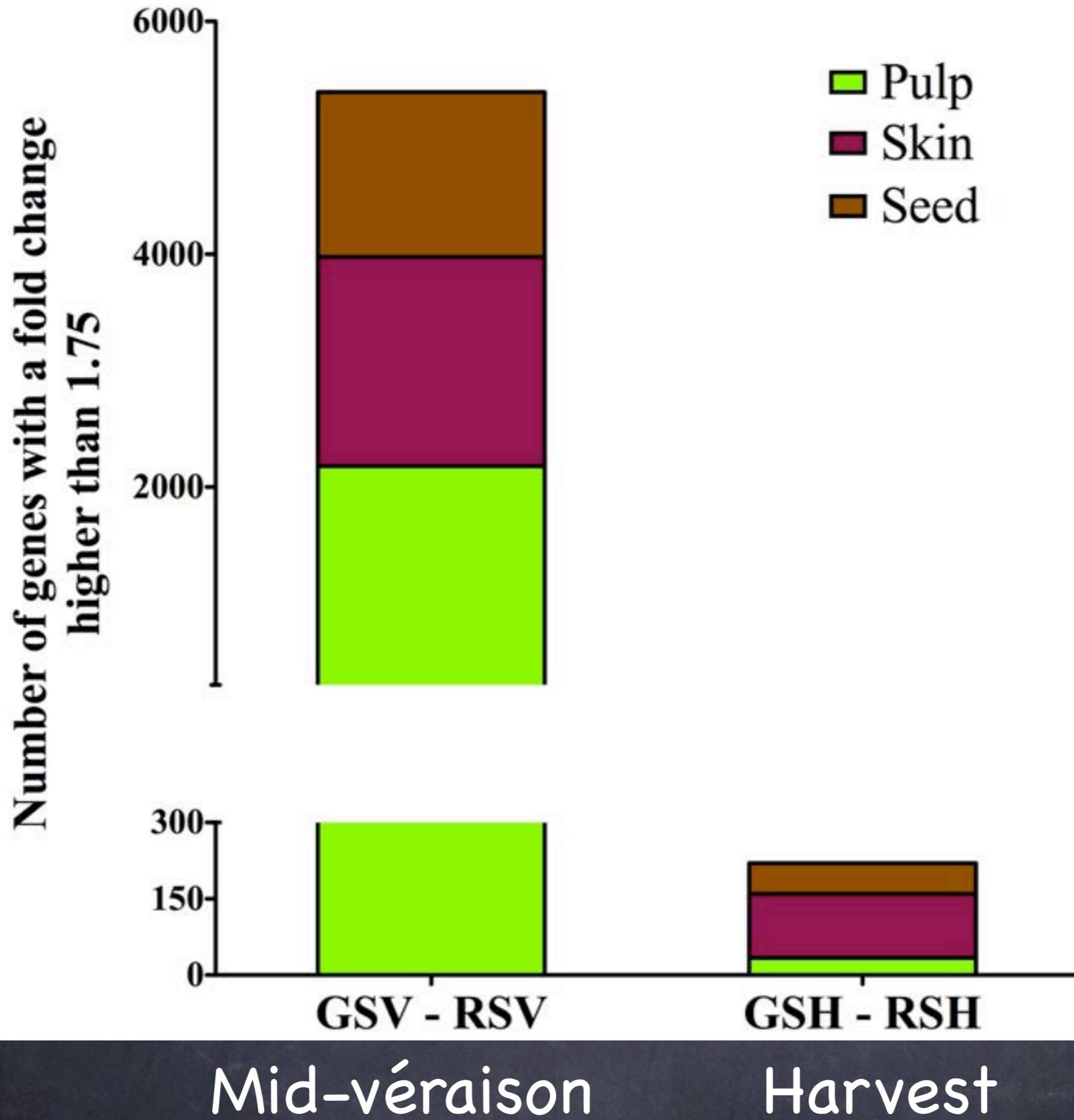
Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
Genes	#	%	#	%	#	%	#	%	#	%	#	%
Pulp	4,393	15	2,180	7	1,266	4	1,060	4	200	0.6	60	0.2
Seed	3,102	10	1,416	5	781	3	1,004	4	327	1.1	116	0.4
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# Transcript profile of a UFGT gene



For 29,550 genes

Tissue	Mid-véraison (GSV, PSV, RSV)						Near Harvest (GSH, PSH, RSH)					
	>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change		>1.5 Fold Change		>1.75 Fold Change		>2 Fold Change	
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Seed	3,102	10	1,416	5	781	3	1,004	4	327	1.1	116	0.4
Skin	3,986	13	1,797	6	998	3	1,748	6	535	1.8	201	0.7



## **Conclusion:**

**Sugar and pigmentation data indicate uniform berry populations within the cluster at harvest**

**No perfect berry classes but more «balanced» berry classes**

**Reduction of berry variability at the gene expression level at harvest is effective, few genes with high variability between not fully ripe and more advanced berries**

**This reduction affects the three tissues**

**Pulp appears to be the most synchronized tissue**

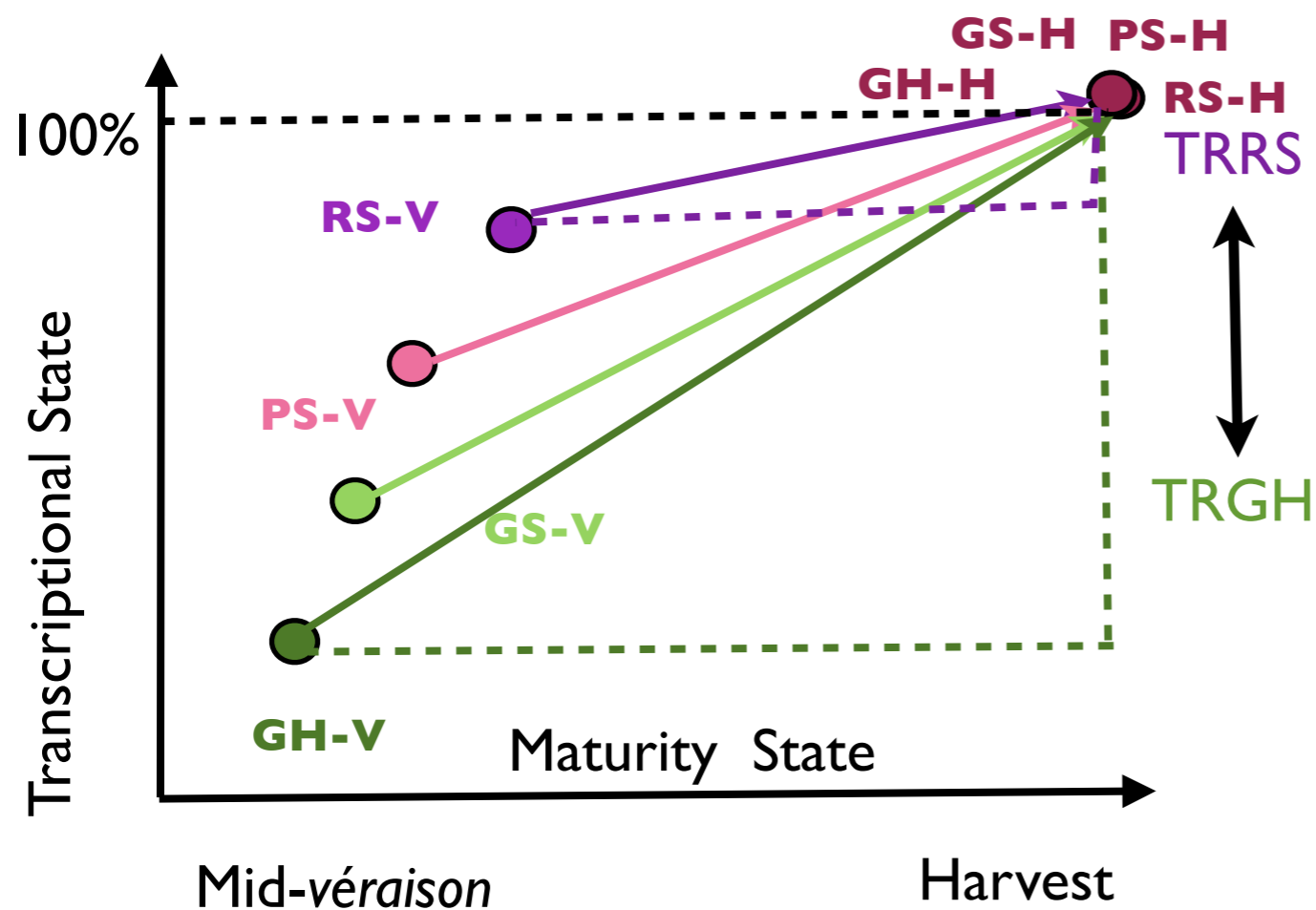
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# Difference in transcriptional rates between GH and RS berry classes

**Transcriptional Rate (TR):** Transcriptional state at harvest - Transcriptional State at mid-véraison



**TRGH:** tsGHH-tsGHV

**TRRS:** tsRSH-tsRSV

Differential of TR:

**TRGH-TRRS**





# Genes in the pulp showing an enhanced transcriptional rate in GS versus RS from Mid-V to Harvest

370 genes found with an enhanced transcriptional rate superior to 2

Gene	Functional Category	Transcriptional rate (GS/RS) from V to H
Xyloglucan:xyloglucosyl transferase	Cell wall modification	3.41
Extensin	Cell wall structural protein	6.36
CYP79A2	Glucosinolate metabolism	11.96
chalcone synthase	Flavonoid biosynthesis	6.19
Flavonoid 3',5'-hydroxylase (8)	Flavonoid biosynthesis	4.14
Flavonone- 3-hydroxylase	Flavonoid biosynthesis	3.29
Chalcone isomerase	Flavonoid biosynthesis	2.49
UDP-glucose:flavonoid 3-O-glucosyltransferase	Flavonoid-glucoside biosynthesis	7.01
Caffeoyl-CoA O-methyltransferase	Phenylpropanoid biosynthesis	7.11
Limonoid UDP-glucosyltransferase	Flavor	2.46
Sugar transporter ERD6-like 16	Sugar Transporter	5.69
MATE efflux family protein	Flavonoid Transport	4.99
Anthocyanin permease AM3	Flavonoid Transport	2.25

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# Genes in the pulp showing an decreased transcriptional rate

280 genes found with a repressed transcriptional rate inferior to 0.5

Gene	Functional Category	Transcriptional rate (GS/RS) from V to H
flavonoid 3-monooxygenase	Flavonoid biosynthesis	0.4
UDP-glucosyl transferase	Flavonoid biosynthesis	0.31
flavonoid 3-monooxygenase	Flavonoid biosynthesis	0.18
UDP-rhamnose:rhamnosyltransferase	Anthocyanin-glucoside biosynthesis	0.46
Orcinol O-methyltransferase 2	Isoflavonoid biosynthesis	0.38
(+)-neomenthol dehydrogenase	Monoterpenoid biosynthesis	0.43
CYP72A1	Monoterpenoid biosynthesis	0.33

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# What do we do next?

***What is the extent of this reduction in berry variability at the metabolic level?***

***If there is a catch up, in which direction? when the synchronization is initiated?***

***Study the post-harvest stage to estimate whether the optimum of maturity is subject to variability***

***Characterization of some genes involved in the control of ripening***

***Extending this knowledge to other cultivars to understand the genetic control of ripening and the effects of environmental factors***

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# How could these data be useful for the Industry

## Validate the usefulness of these markers:

- 1) as descriptors and predictors of the ripening,
- 2) to validate the effects or the timing of viticulture practice
- 3) to estimate the variability not at the cluster level but at the vineyard level

## Characterize the impact of berry metabolic variability in the origin of an unbalanced wine:

## Development of a breeding program:

- to screen best clones that will propose uniform clusters

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## Deluc Lab:

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Alex Moeller

## Center for Genome Research and Biocomputing:

Caprice Rosato



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