



# Success and bottlenecks of large scale plant protein turnover measurement

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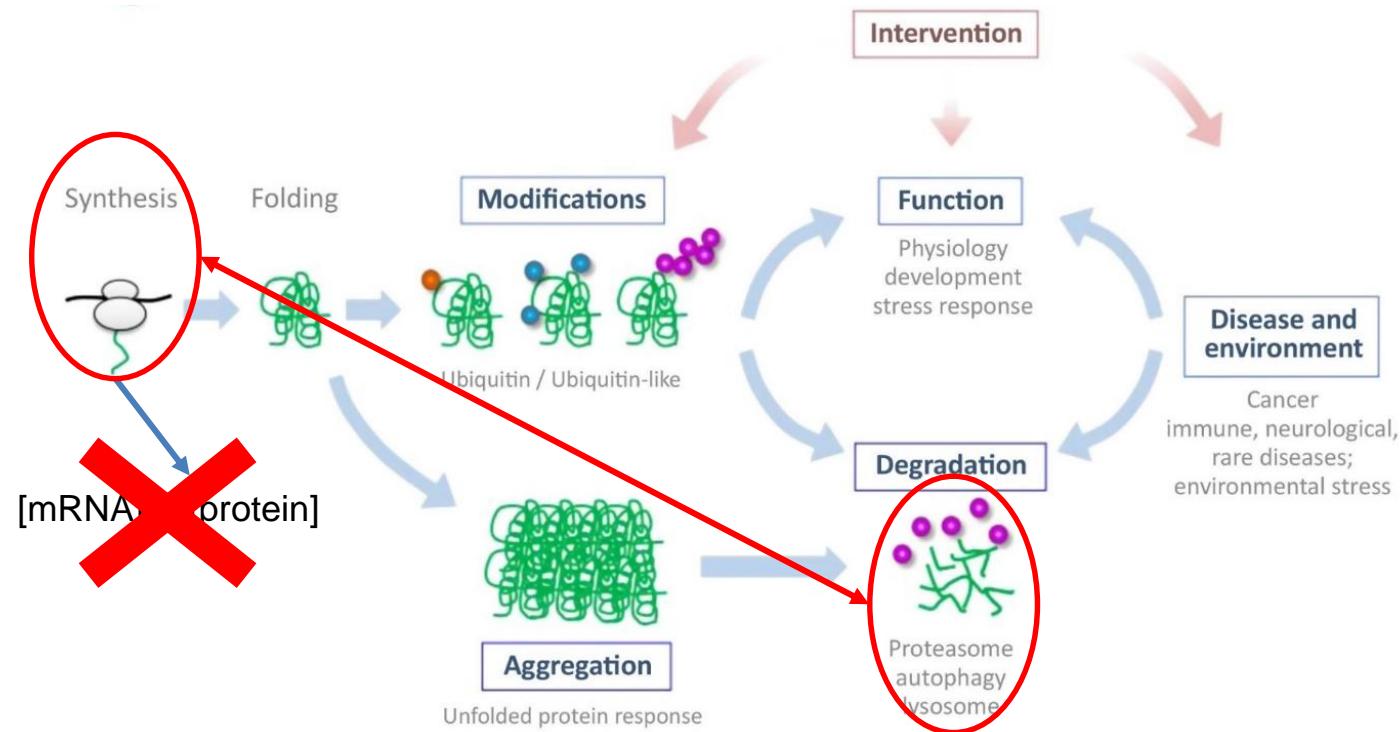
Génétique  
Quantitative  
et Évolution  
Le Moulon



# Success and bottlenecks of large scale plant protein turnover measurement: a tomato fruit maturation case study

Willy Bienvenut, IDEEV, GQE, Gif/Yvette, France

# Proteostasis and large scale protein turnover rate determination



Dissmeyer *et al.* Trends Biochem. Sci. 2019

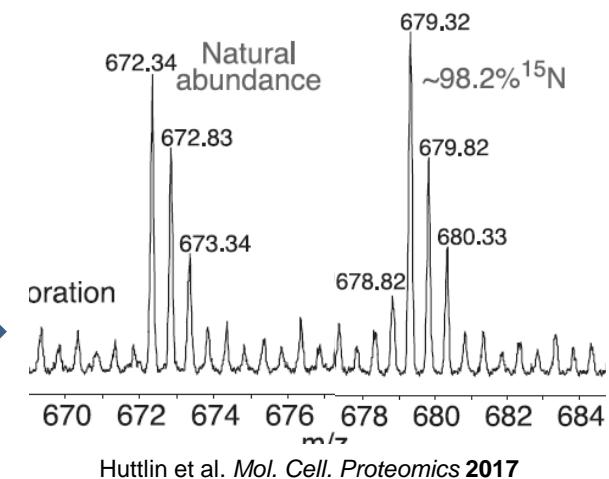
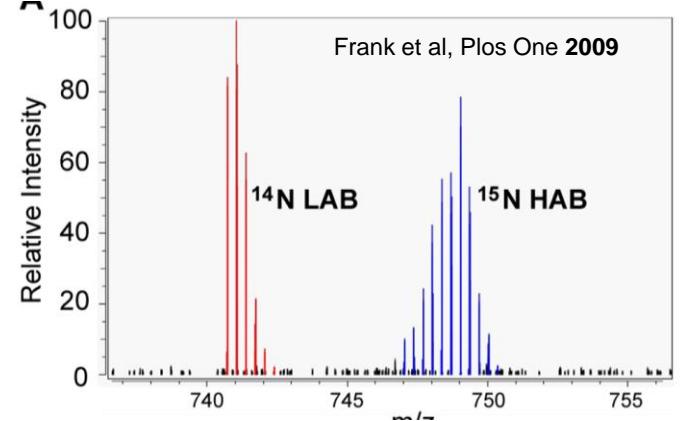
Influence of protein Half-life and Turnover rate  
in protein abundance variability

# Protein turnover in phototrophic organisms...



*S. Lycopersicum*  
Var. Moneymaker  
hydroponic culture  
In mineral substrate

Water, CO<sub>2</sub>,  
fertilizers,  
trace  
elements...



# Usual $K_d/K_s$ equations

Very few technical approaches proposed for plants...

$$K_D = - \frac{\ln(FCP \cdot (1 - LPF))}{t}$$

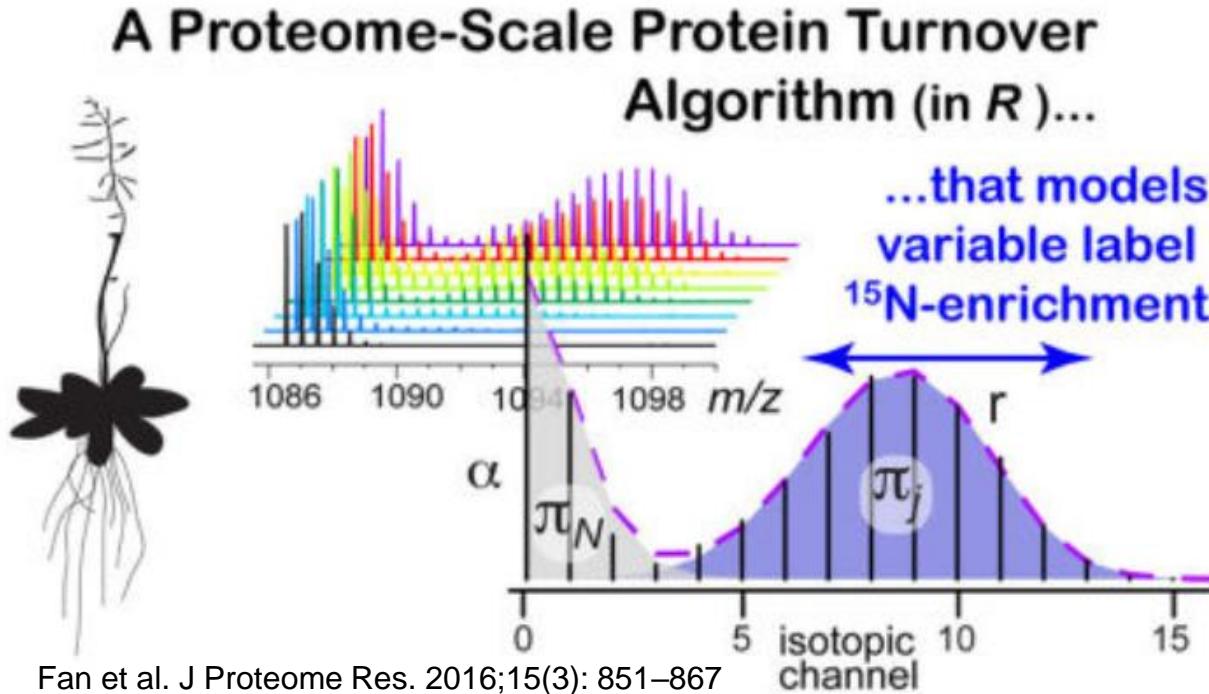
$$\frac{K_S}{A} = \frac{FCP - e^{-K_D \cdot t}}{1 - e^{-K_D \cdot t}} \cdot K_D$$

## Determination of the:

**FCP:** Fold Change in Protein

**LPF:** Labelled Protein Fraction

# Is that simple...?



- High data processing
- Some scripts not available
- $\text{N}$  labelled peptides are not compatible with MS characterisation:
  - Most of the acquired data are useless

# An alternative approach... ?

- Adapted to the processing tools available locally
  - i2MassChroQ (X!Tandem identification tool and inference filtering interface)
  - MassChroq (peptide quantitation)
  - MCQR (Protein quantitation & statistics R package)
- Providing a dedicated pipeline to perform large scale protein turnover measurement
- Reducing data complexity (MS) and cost ( $^{15}\text{N}$ ) ...

Determination of the FCP coefficient

(Fold Change in Protein)



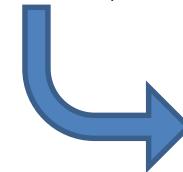
Label-free quantitation

MCQR tool

Adapted version compatible with  $^{15}\text{N}$  labelling

Determination of the LPF

(Labelled Protein Fraction)



Based on the isotopic distribution

MassChroQ &

dedicated R scripts

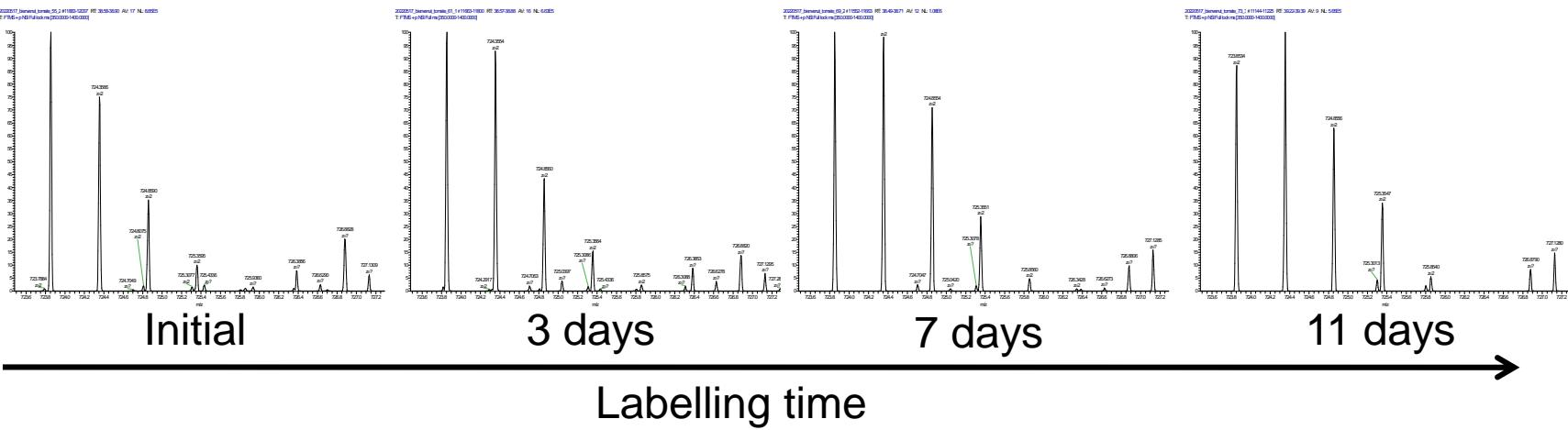
# Isotopic distribution: a new way to explore...

*S. Lycopersicum*  
Var. *money Maker*  
*hydroponic culture*  
*In coconuts fiber*



5-10%  $^{15}\text{N}$   
enriched media  
→  
Metabolic  
labelling pulse

Standard proteomics  
sample processing



# How to estimate the LPF ?

$$LPF = 1 - \frac{R_1^{\text{Exp}} - R_1^{15N}}{R_1^{\text{Nat}} - R_1^{15N}}$$

$$R_1 = \frac{I_1}{I_0}$$

Exp: experimental measurement

$^{15}\text{N}$ : Max  $^{15}\text{N}$  incorporation at  $t^\infty$  (Theo/Calc)

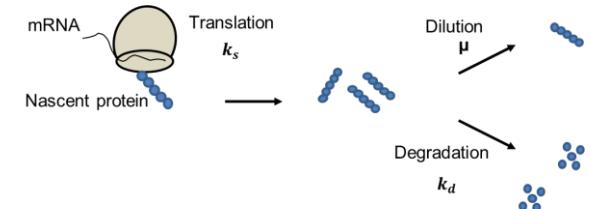
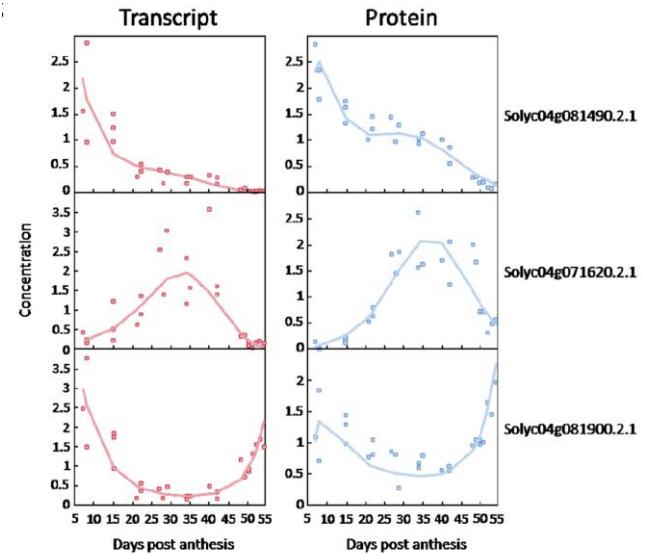
Nat: Natural isotopic distribution (Theo/Calc)

Sénécaut N. *et al.* J Proteome Res. 2021; 20(3):1476-1487

# The PROOFER project:

exPeRimental validatiOn OF protEin tuRnover

- Protein turnover model developed on tomato fruit
  - Belouah *et al.* 2019, J. Proteomics
- Protein turnover model using:
  - Transcriptomics data:
    - Absolute quantitation of 22877 transcripts with spikes in the experimental RNAseq
  - Proteomics data:
    - XIC absolute quantification of 2490 proteins using peptide intensity modelling



$$\frac{dp(t)}{dt} = k_s r(t) - (k_d + \mu(t))p(t)$$

# Samples preparation and proteins ID



- \* All samples collected at 12 days post anthesis

- \* Labelling time was variable (0- 12 days)

Samples collected at multiple  $^{15}\text{N}$  labelling time points



Standard proteomics sample preparation



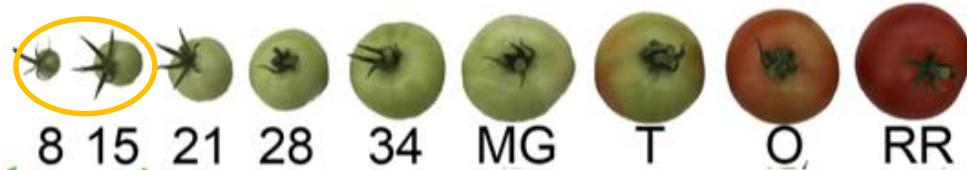
i2MassChroQ

X! Tandem

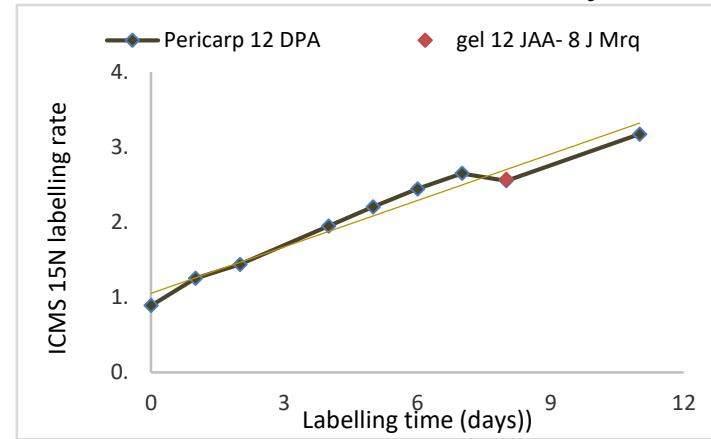


REFERENCE-DB.protein.fasta

Tomato developmental stages



Evolution of the  $^{15}\text{N}$  rate by ICMS



List of identified



peptides and  
proteins

MassChroQ

Label free quantitation  
& Isotopic distribution  
exports

Isotoptic



profiles

# MCQR: R package for quantitation

## How to determine the Fold Change in Protein (FCP)

- Drawbacks:

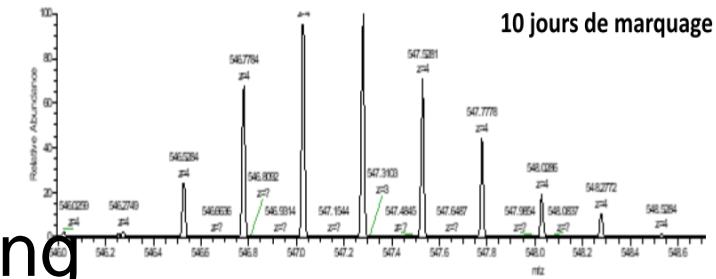
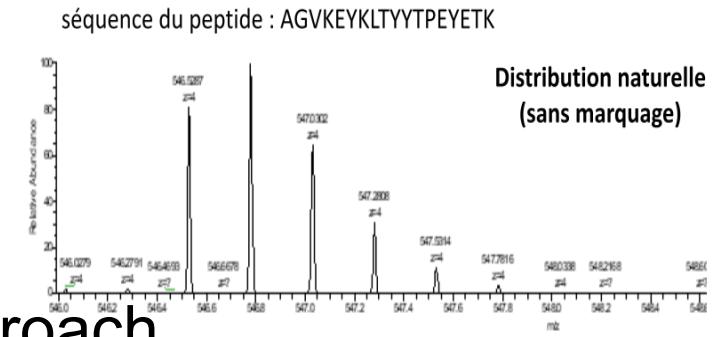
- Based on 1<sup>st</sup> or most intense isotope
- <sup>15</sup>N metabolic labelling modify the isotopic distribution

- Modified processes:

- Intensity normalisation
- Alternative “missing value” approach
- 

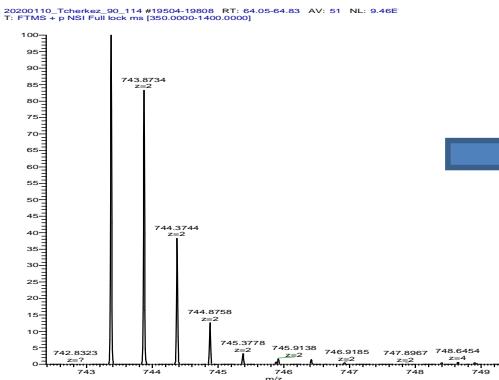
- New **IsoProt\_MCQR** script

- Compatible with <sup>15</sup>N labelling



# Display of the isotopic ratios

Peptide experimental isotopic distribution



$$\rightarrow R_i = \frac{I_i}{I_0}$$

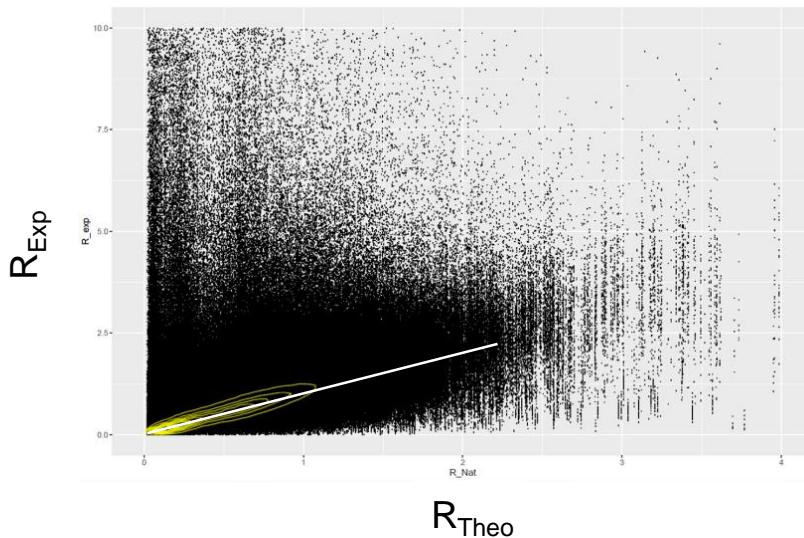
## Experimental Intensity

Extracted by MassChroQ  
Converted in  $R_i$  by R script

## Theoretical $R_i$ calculation

IsoSpec python library by M. Lacki;  
adapted by P. Rusconi

Example for the Ctrl sample at t=0



5.190.167  $R_i$   
106.219 peptides profiles  
29,840 nr pepIDs

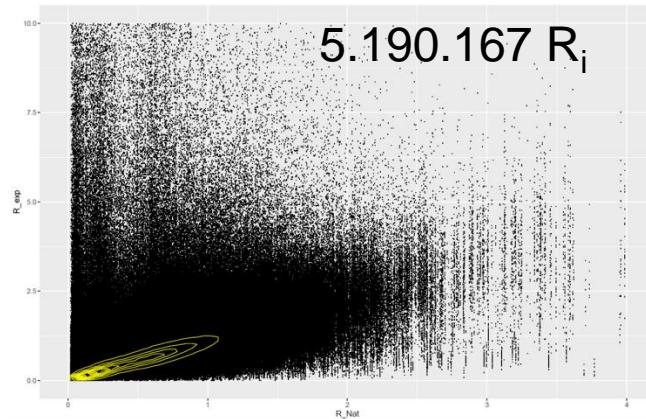
$$R_{\text{exp}} = 0.76 * R_{\text{theo}} + 0.39, R^2 = 0.02, \text{IFS} = 0.01$$

# Filtering the raw data

- Minimum number of isotopes:
  - Usually 4 isotops min ( $I_0-I_3$ )
- Maximum  $^{15}\text{N}$  incorporation rate:
  - based on theo. Max.  $^{15}\text{N}$  rate isotopic distribution
- $R_i$  dispersion:
  - Av  $\pm 1$  SD range retained (Theo. 70% retained)
- **Minimum data redundancy:**
  - Min 2/3 Anal. Rep.,
  - >4 isotopic profiles / Rep,
  - >3 Isotopes / isotopic profil
  - ...

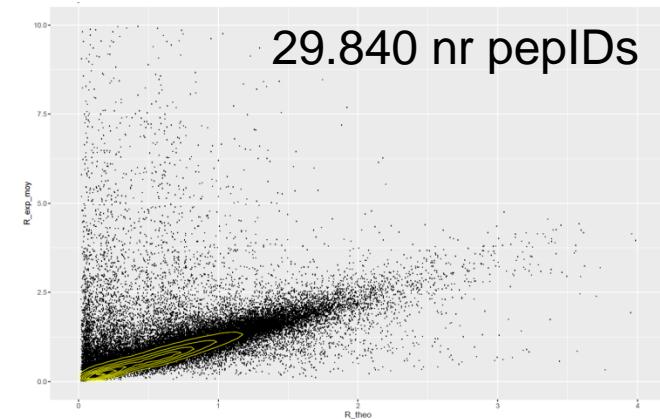
# Filtering the raw data (initial vs. final)

$$R_{\text{exp}} = 0.76 * R_{\text{theo}} + 0.39, R^2 = 0.02, \text{IFS} = 0.01$$

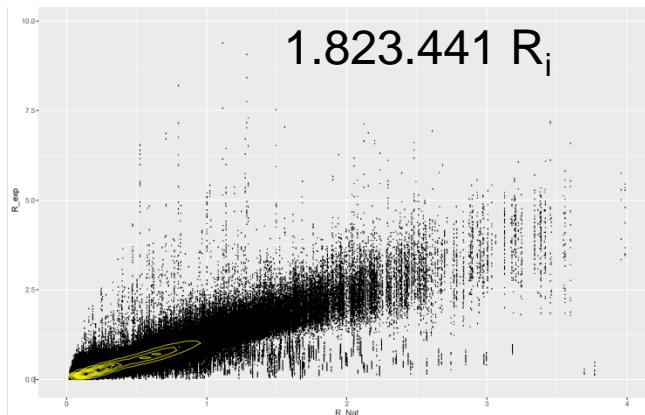


$$R_{\text{exp}} = 0.75 * R_{\text{theo}} + 0.43, R^2 = 0.07, \text{IFS} = 0.03$$

Initial data

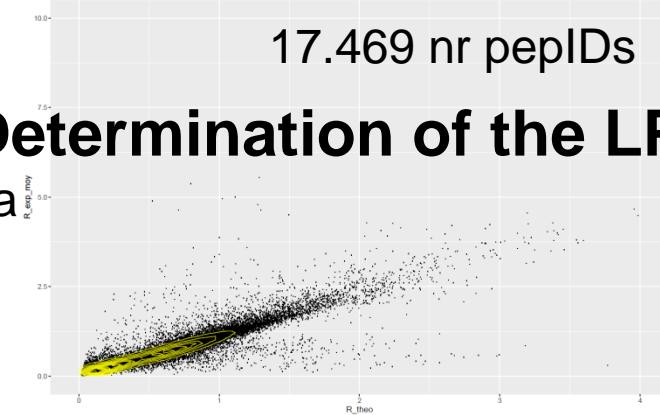


$$R_{\text{exp}} = 1.01 * R_{\text{theo}} + 0.05, R^2 = 0.95, \text{IFS} = 1.27$$



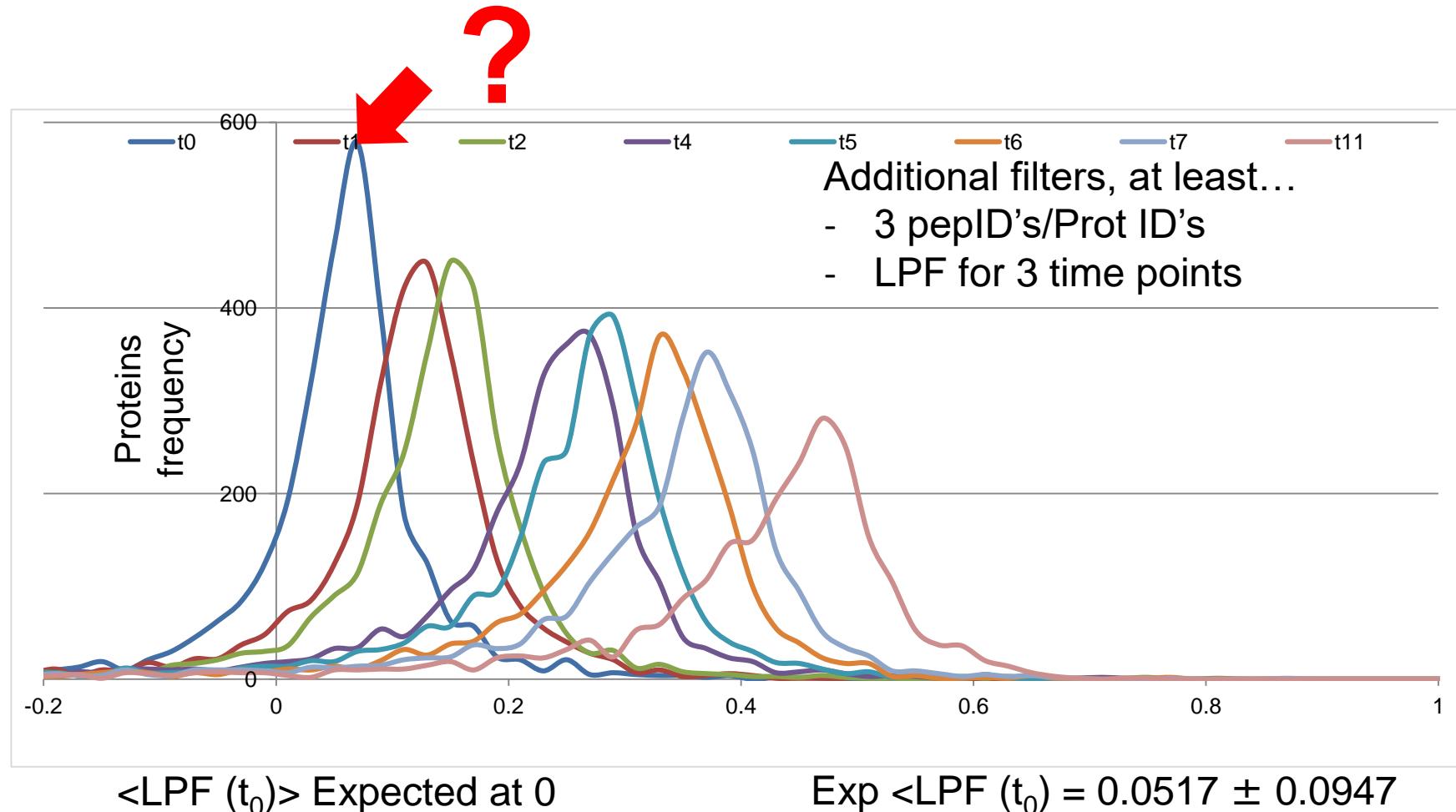
Filtered data

$$R_{\text{exp}} = 1.00 * R_{\text{theo}} + 0.06, R^2 = 0.94, \text{IFS} = 1.21$$



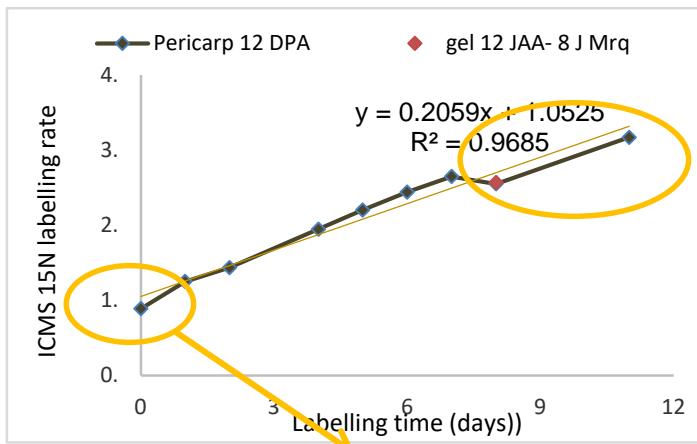
## Determination of the LPF

# Protein LPFs distribution



# Experimental drawbacks...

## Evolution of the $^{15}\text{N}$ rate by ICMS



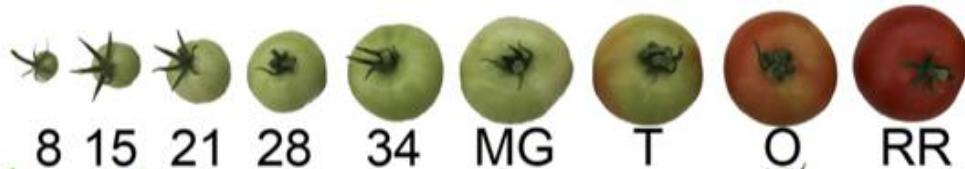
$$\text{Natural } ^{15}\text{N rate} = 0.364\%$$



Technical problem related to automatic plant watering system during the WE

Labelling time + 1 days

## Tomato developmental stages



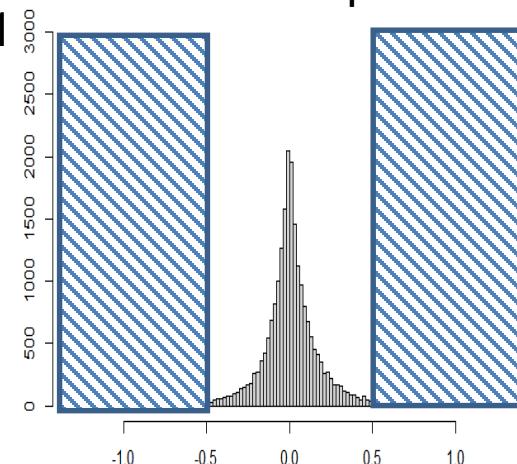
All samples were collected at 12 days post anthesis

Labelling time was variable (0- 12 days)

By definition:  $2511/2973$  proteins  
 $\text{FCP} = 1$



Selection of proteins with no significant variation in quantity



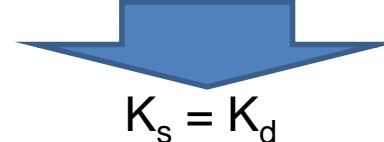
# Determination of the $K_d/K_s$

$$K_D = - \frac{\ln(FCP \cdot (1 - LPF))}{t}$$

Since FCP = 1

$$K_d(t) = \ln(1 - LPF)/t$$

$$\frac{K_s}{A} = \frac{FCP - e^{-K_D \cdot t}}{1 - e^{-K_D \cdot t}} \cdot K_D$$



$$K_s = K_d$$

**Av. individual values**

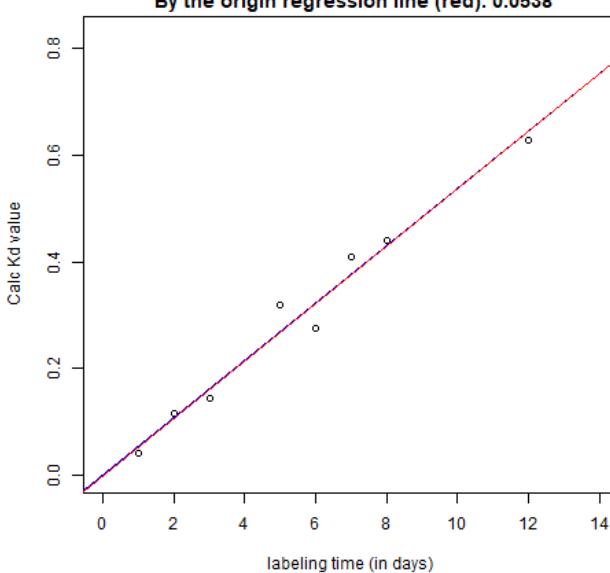
$$\langle K_d \rangle = 0.0528 \pm 0.0077 \text{ d}^{-1}$$

**Classic linear reg.**

$$K_d = 0.0536 \text{ d}^{-1}, R^2 = 0.987$$

**Reg. line through the origin**

$$K_d = 0.0538 \text{ d}^{-1}$$



**Av. individual values**

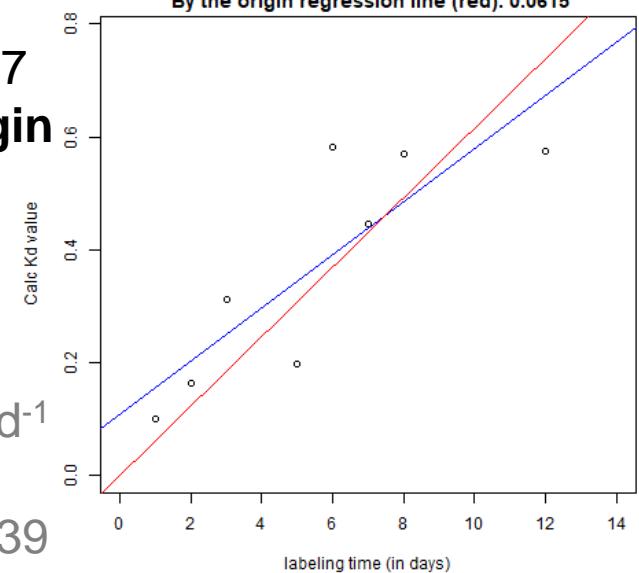
$$\langle K_d \rangle = 0.0756 \pm 0.0242 \text{ d}^{-1}$$

**Classic linear reg.**

$$K_d = 0.0469 \text{ d}^{-1}, R^2 = 0.839$$

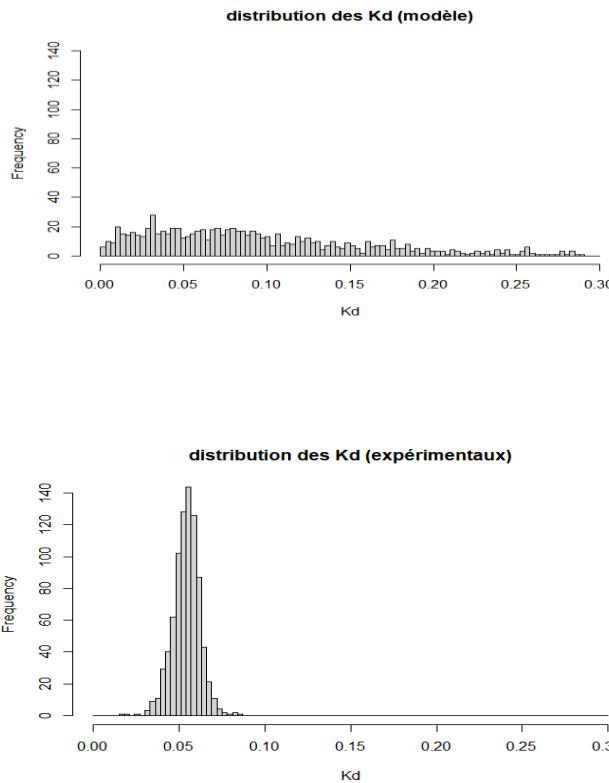
**Reg. line through the origin**

$$K_d = 0.0615 \text{ d}^{-1}$$



# Results:

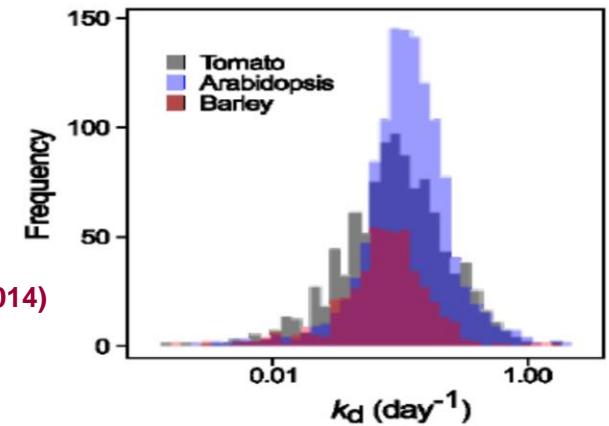
## Predicted $K_d$ (Belouah *et al.* model) vs. Exp $K_d$



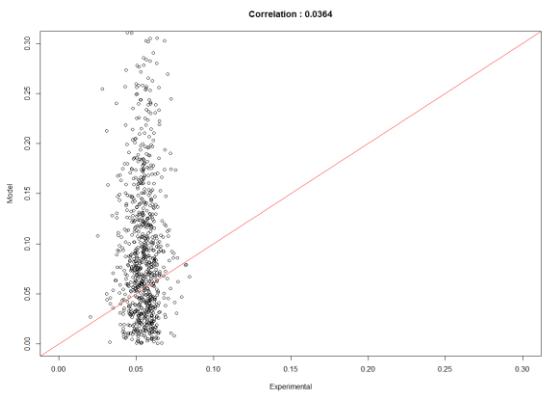
Distribution of 1103 predicted  $k_d$  of the tomato fruit : 0,113 day

Arabidopsis leaf median  $0.113 \text{ day}^{-1}$ ,  
~ 9 days (1228  $k_d$  from Li *et al.*, 2017)

Barley leaf median  $0.076 \text{ day}^{-1}$ ,  
~13 days (505  $k_d$  from Nelson *et al.*, 2014)



Averaged individual values  
 $\langle K_d \rangle = 0.0601 \pm 0.0121 \text{ d}^{-1}$



Similar  $\langle K_d \rangle$  value but with a much more restricted range  
Strong discrepancies between the model and Exp. values

# Conclusions

- Partial N metabolic labelling:
  - compliant with LPF estimation
  - compatible with  $^{15}\text{N}$  labelling
- Calculation of  $K_d(K_s)$  is possible
- Restricted Kd range may be related to:
  - The sampling plan (too short, 1-10 days)
  - Probable influence of the metabolic lag time (28h)

# Questioning...

- Some additional investigations are required to understand the results discrepancies...
  - What about a model using mRNA & Protein quantitative values to determine protein turnover...
  - Some major bias in Millar's processing pipelines (related to the FCP)...
  - Too few reliable studies in plant for an accurate comparison of the results

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(BFP, Bordeaux)



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