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Cross-species analysis of age-related transcriptome data

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Multi-species approach

	Scientific name	Common name	Lifetime
	<i>Caenorhabditis elegans</i>	Worm	2-3 weeks
	<i>Nothobranchius furzeri</i>	Killifish	3 months
	<i>Danio rerio</i>	Zebrafish	30 months
	<i>Mus musculus</i>	Mouse	36-48 months

- 4 - 5 age levels
- 2 - 5 replicates at each level

Data flow

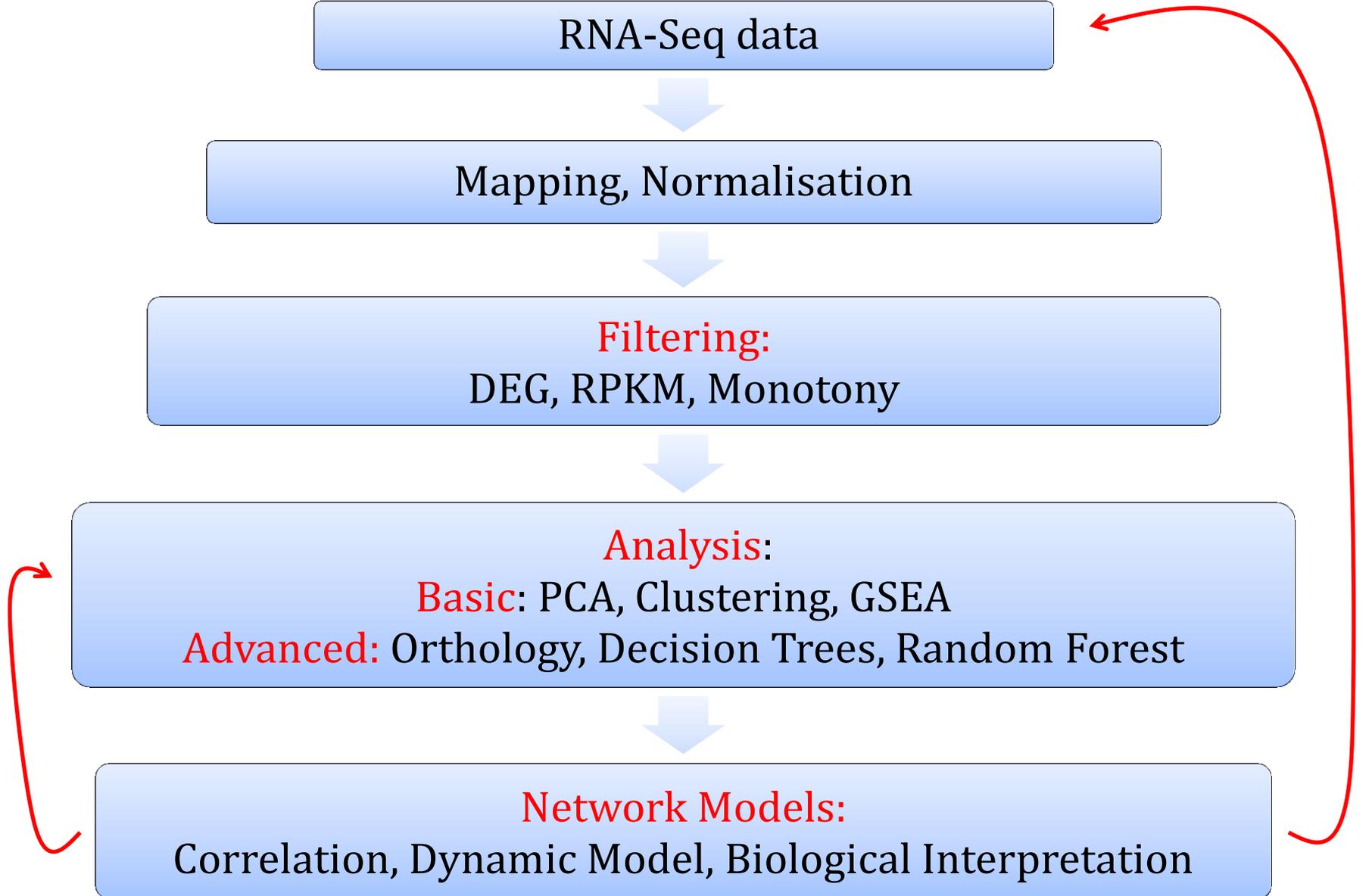
RNA-Seq data

Mapping, Normalisation

Filtering:
DEG, RPKM, Monotony

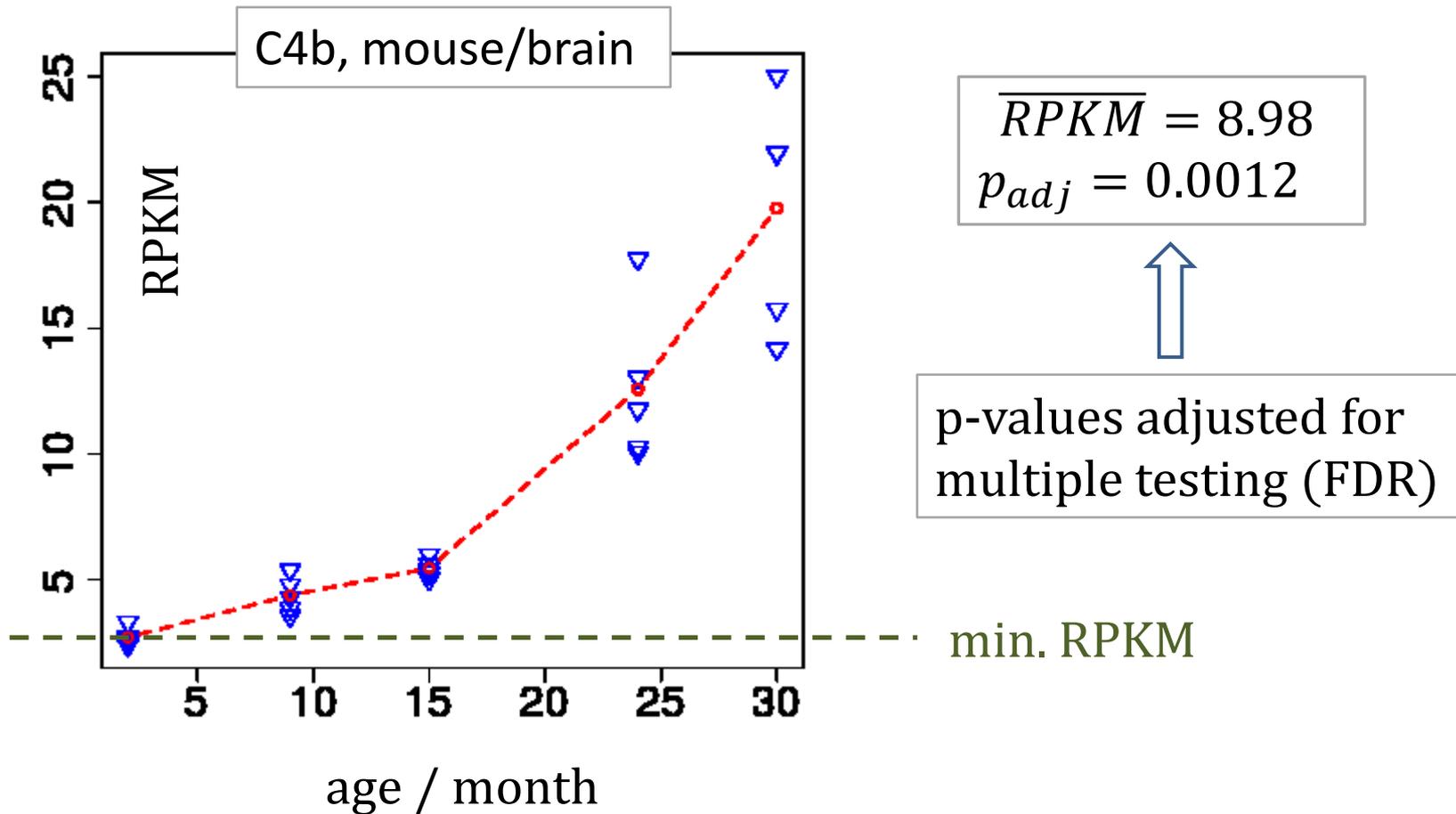
Analysis:
Basic: PCA, Clustering, GSEA
Advanced: Orthology, Decision Trees, Random Forest

Network Models:
Correlation, Dynamic Model, Biological Interpretation



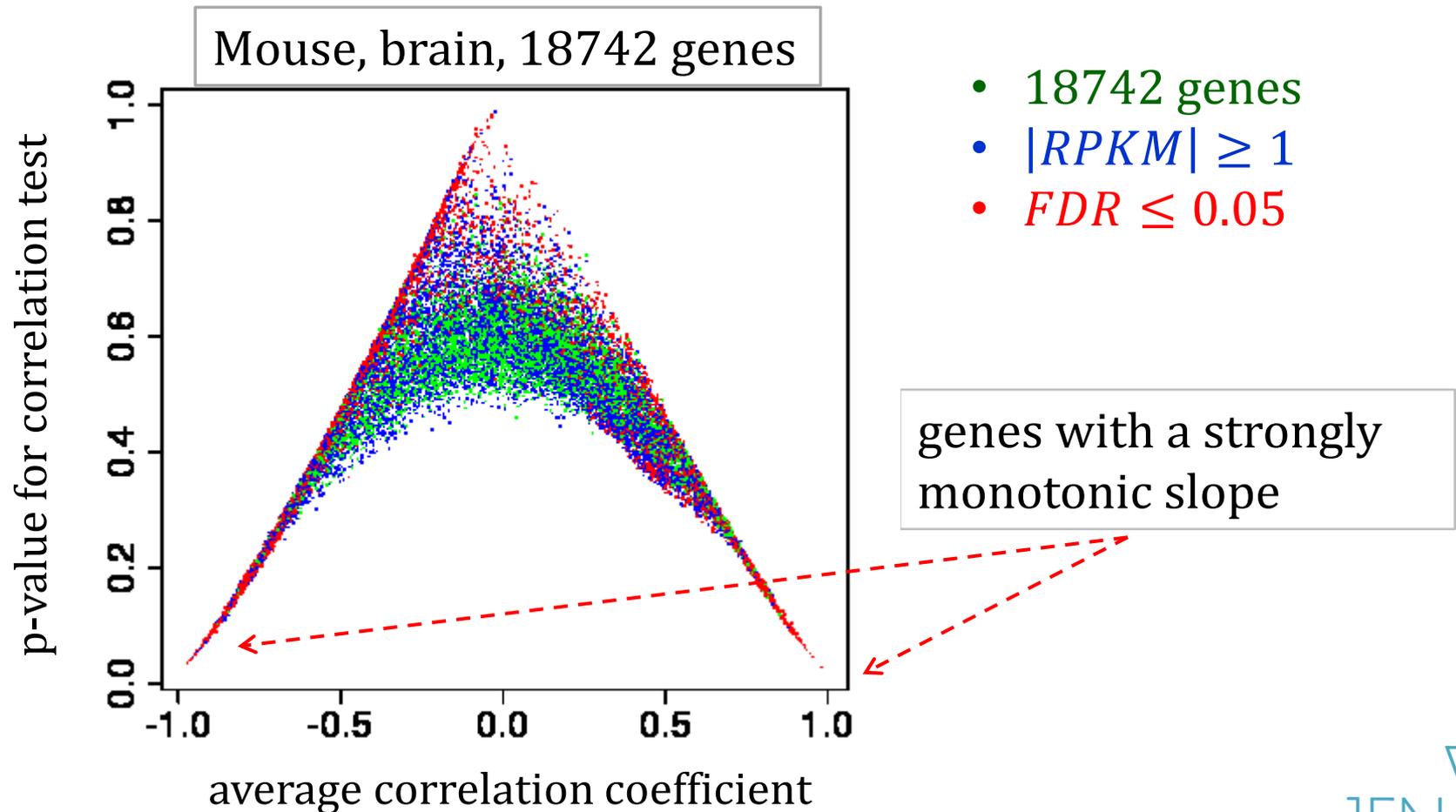
Filtering: 1) DEG, 2) RPKM

1. **DEG**: Differentially Expressed Genes (over age)
 - edgeR, DESeq, baySeq (NegBin, overdispersion)
2. **RPKM**: Reads per Kb of exon model per Million mapped reads



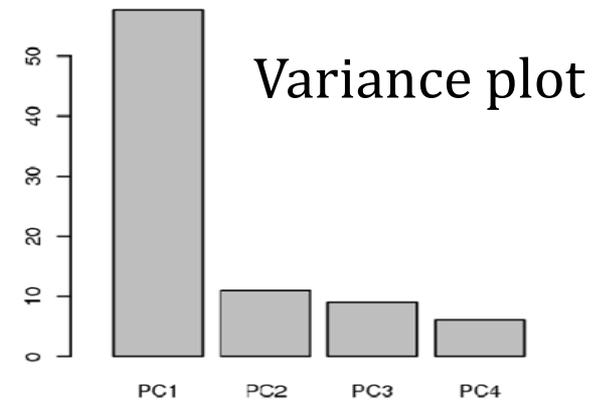
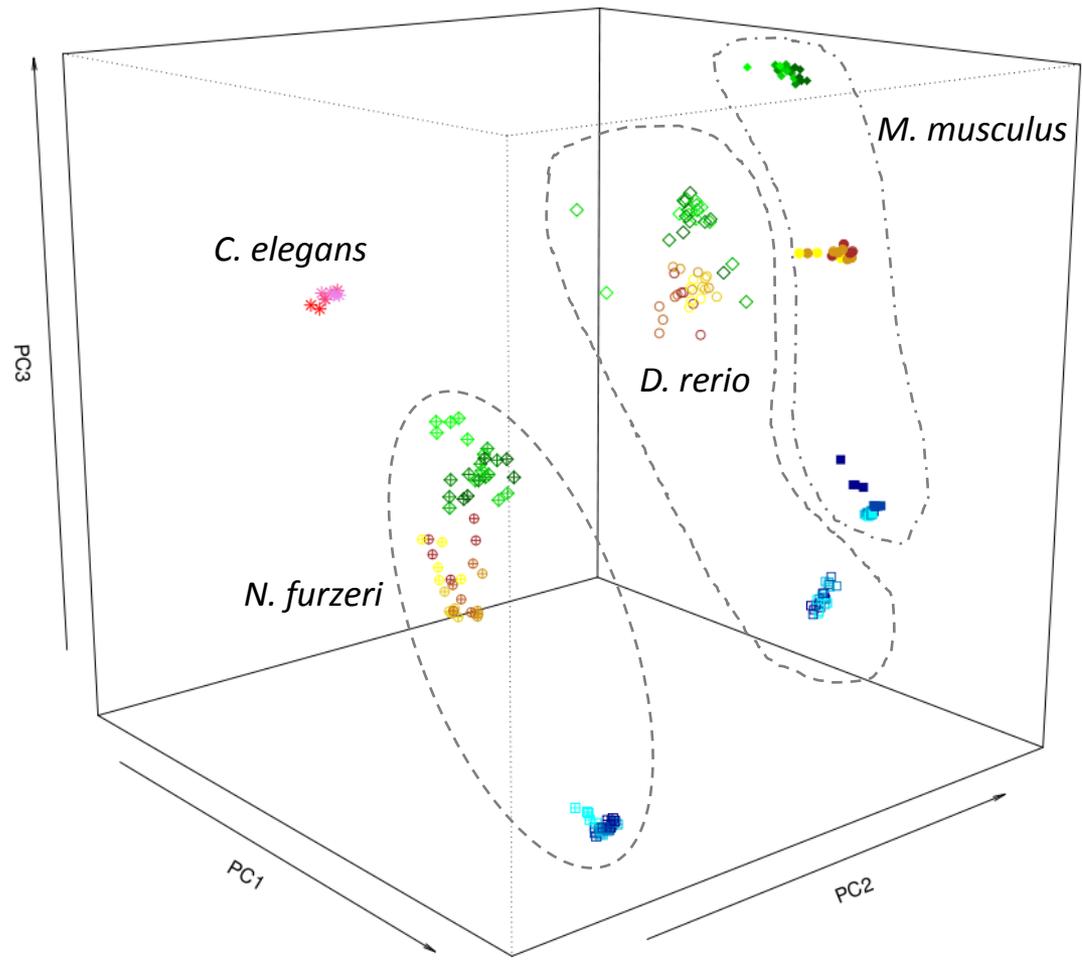
Filtering: 3) Monotony with age

- Genes changing monotonically with age are interesting
- Spearman rank correlation to prototype (Spearman's ρ)
- permutation test to calculate p-values

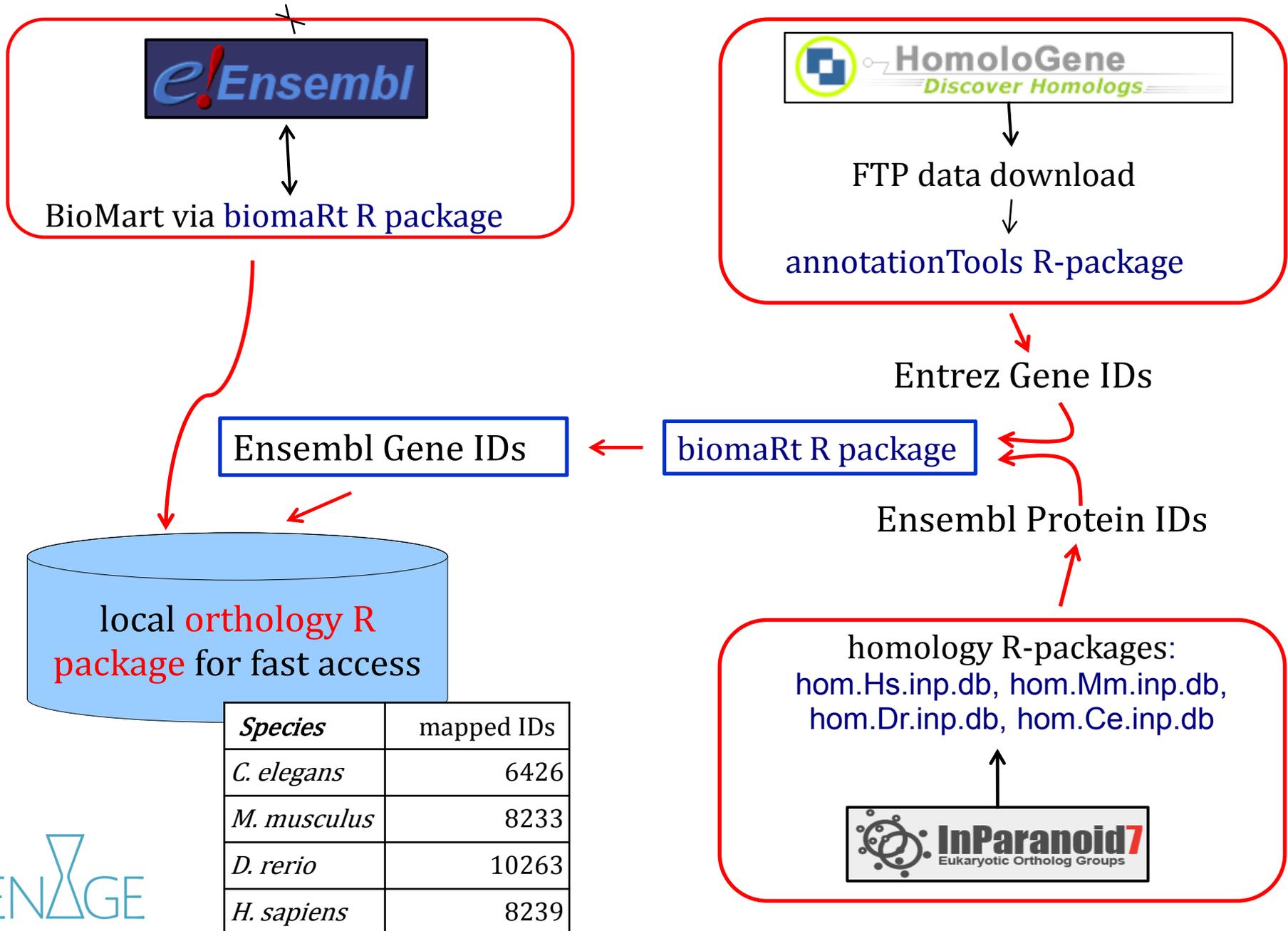


Principal Component Analysis (PCA)

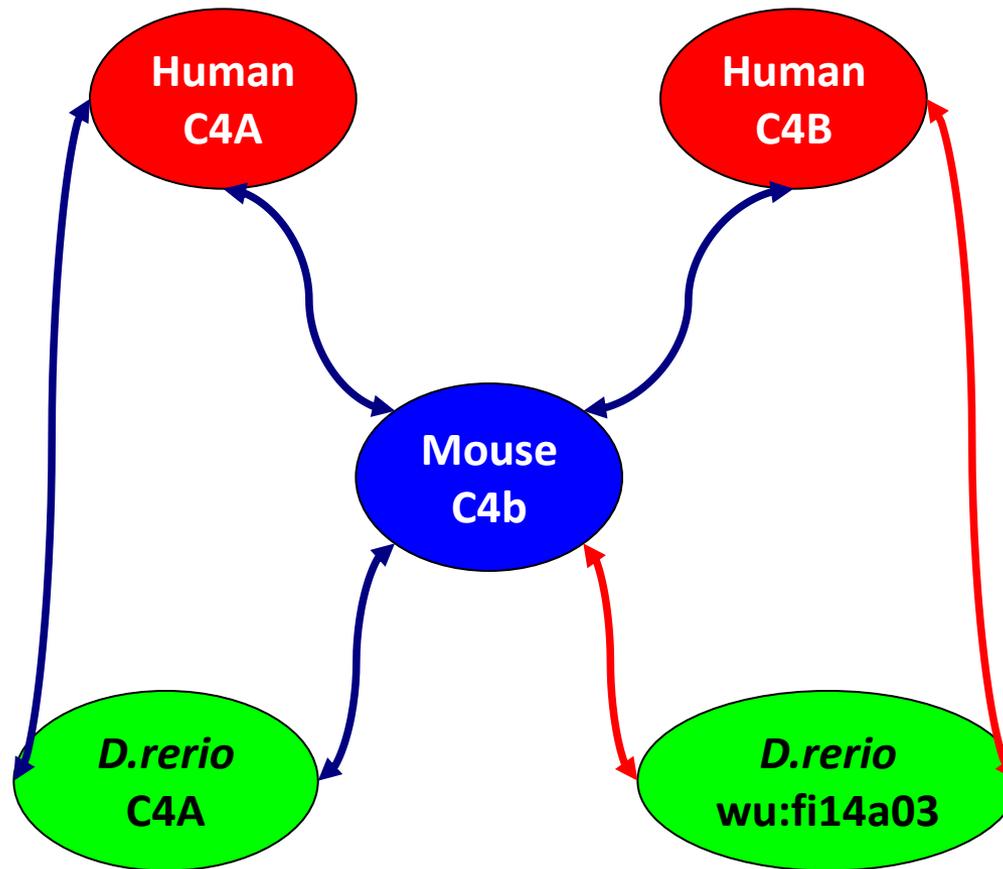
Principal components 1, 2 & 3



Orthology: New, integrated R-package



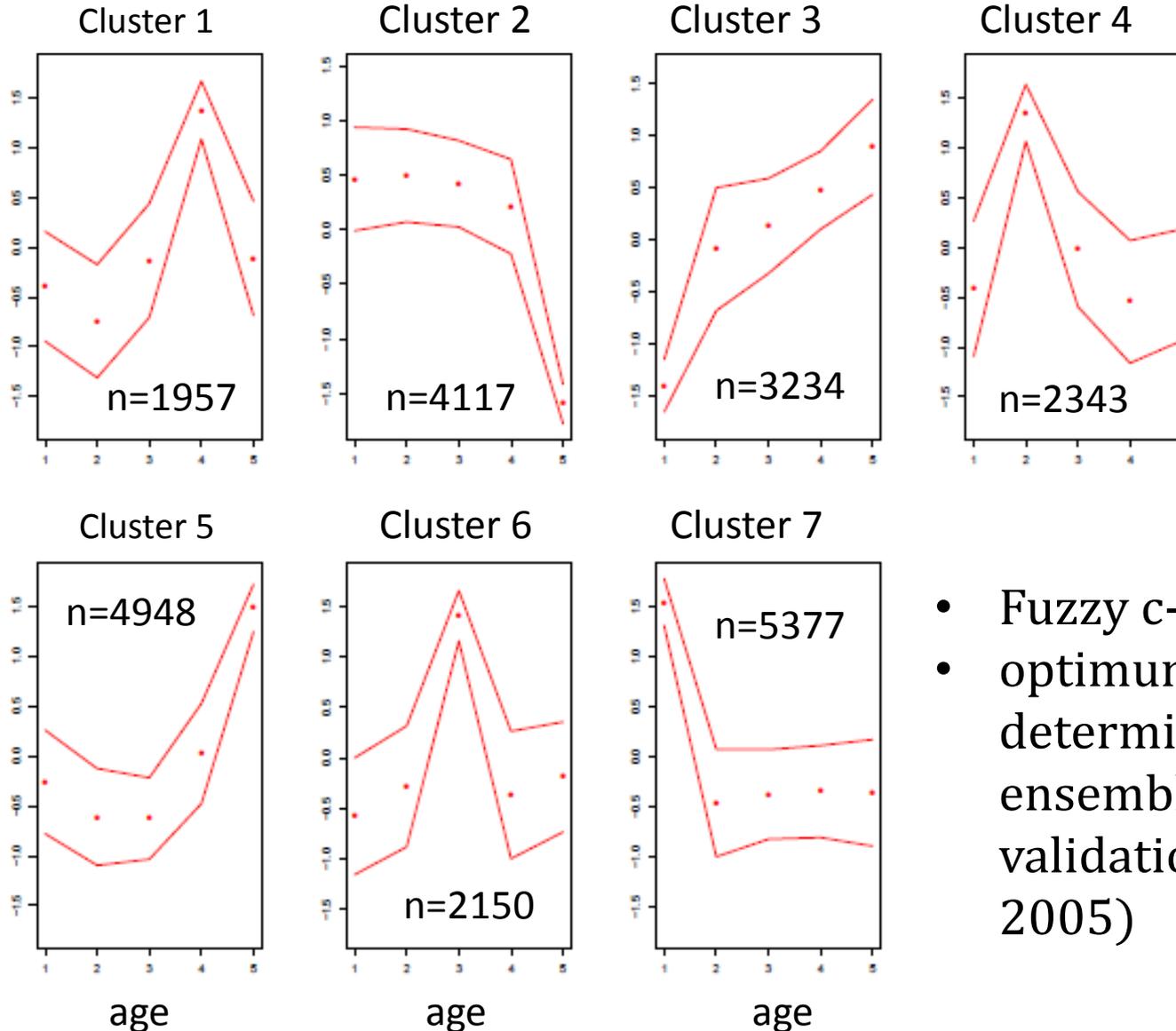
Orthology



color of the arrows indicates database of origin

Clustering (across species)

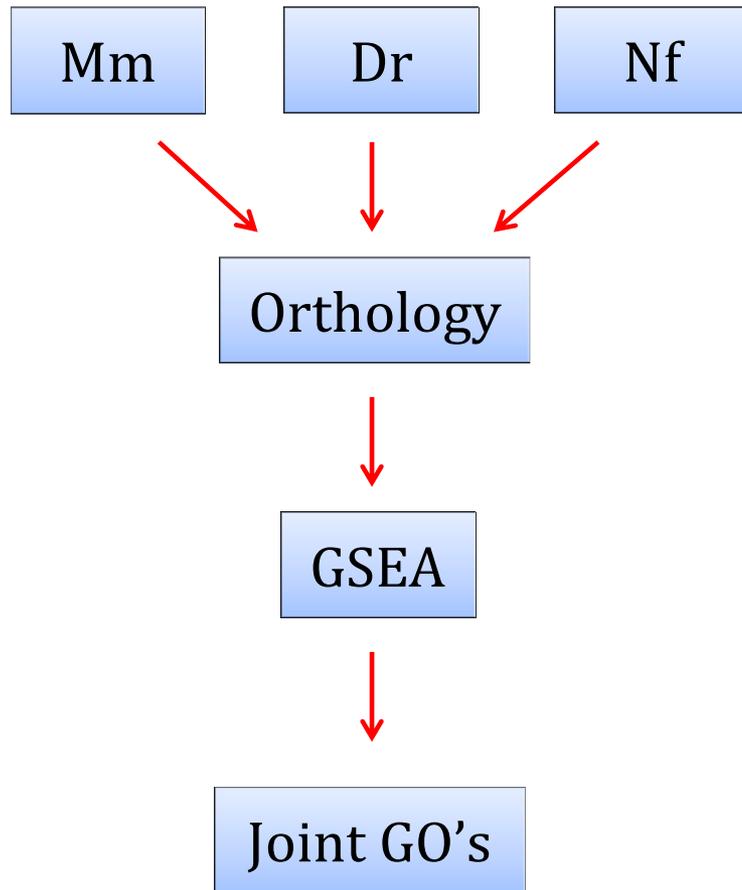
Union of Species: *M. musculus*, *D. rerio*, *N. furzeri*, *C. elegans*



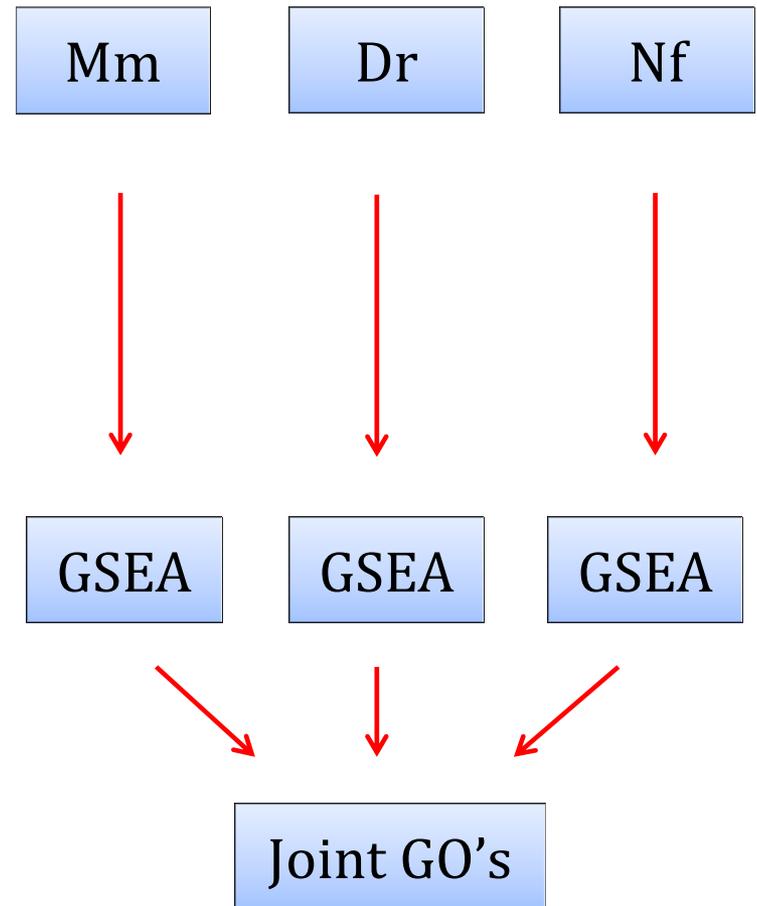
- Fuzzy c-means clustering
- optimum no. of clusters determined using an ensemble of cluster validation indices (Guthke 2005)

Gene Set Enrichment Analysis (GSEA)

Approach I



Approach II



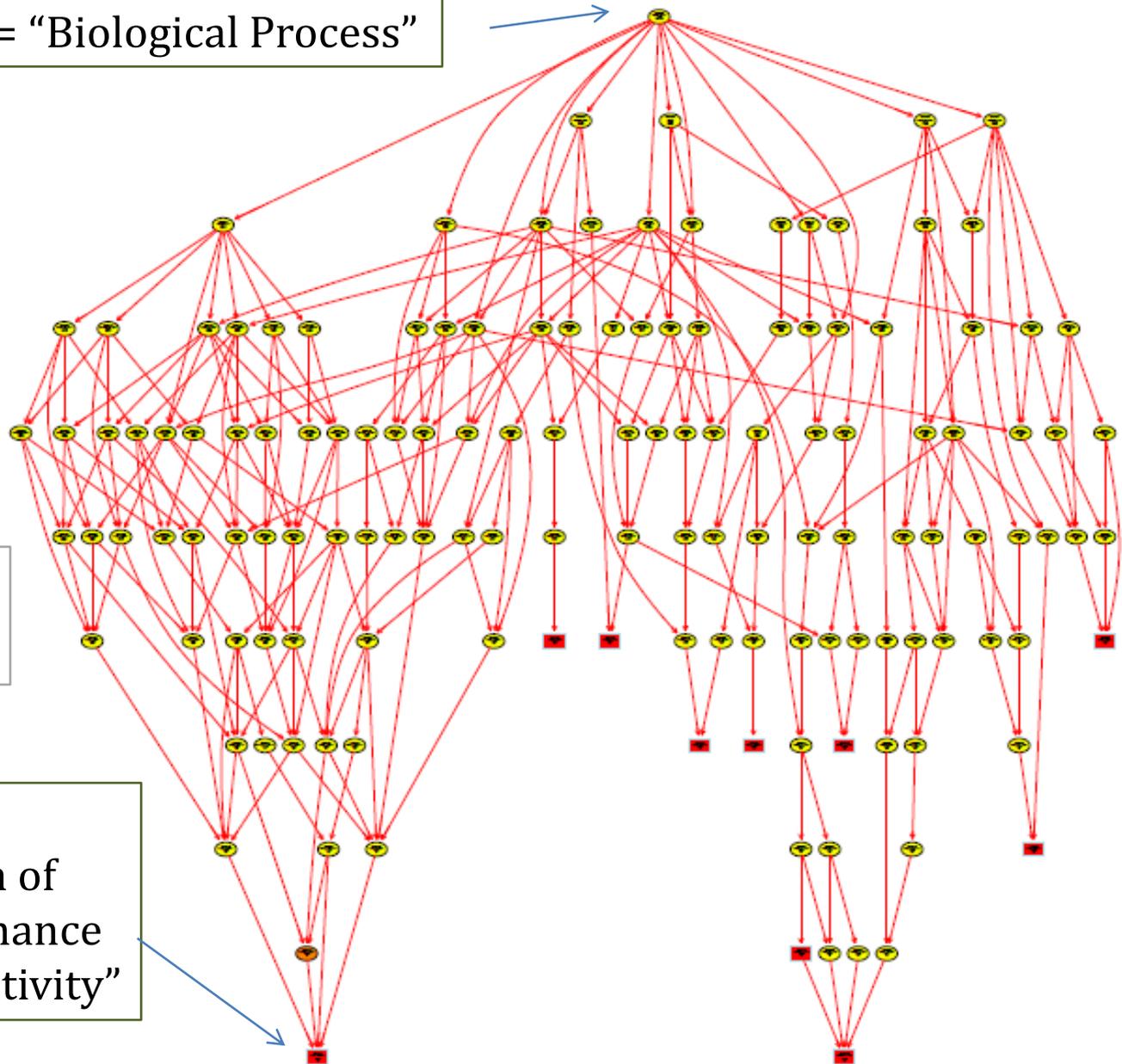
GSEA: Graphical Representation

GO:0008150 = "Biological Process"

Mouse, brain:
176 up-genes
 $FDR \leq 0.05$
 $\overline{RPKM} \geq 3$
 $\bar{\rho} \geq 0.75$

**KS-Test with
elimination**

GO:0032211 =
"down-regulation of
telomere maintenance
via telomerase activity"



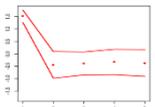
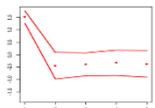
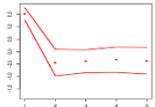
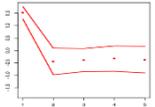
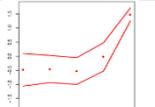
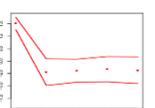
GSEA: Table of Enriched GO's & Pathways

Clustering

GSEA

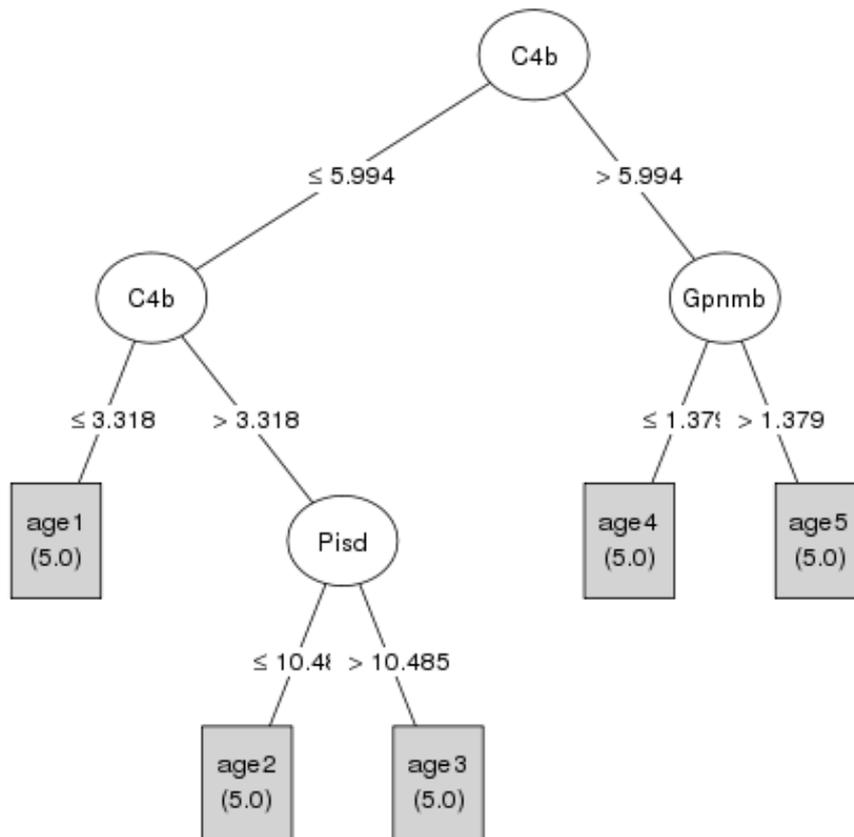
Intersection

3 species: mouse + 2 fishes, > 200 enriched GO's

Tissue	GO-ID	Description	Cluster
brain	GO:0007399	nervous system development	
brain	GO:0007017	microtubule-based process	
brain	GO:0007169	transmembrane receptor protein tyrosine ...	
brain	KEGG:04512	ECM-receptor interaction	
skin	GO:0042113	B cell activation	
skin	GO:0031012	Extracellular matrix	
liver	GO:0000278	Mitotic cell cycle	

Supervised Machine Learning: Decision Trees

- Classifier, can be used when the number of variables (genes) is higher than the number of observations (transcriptome data sets)
- pinpoints genes that are informative with regard to some attribute, e.g. age, tissue, or species.



Mouse, brain

$$|\bar{\rho}| \geq 0.85$$

no FDR or RPKM-filter

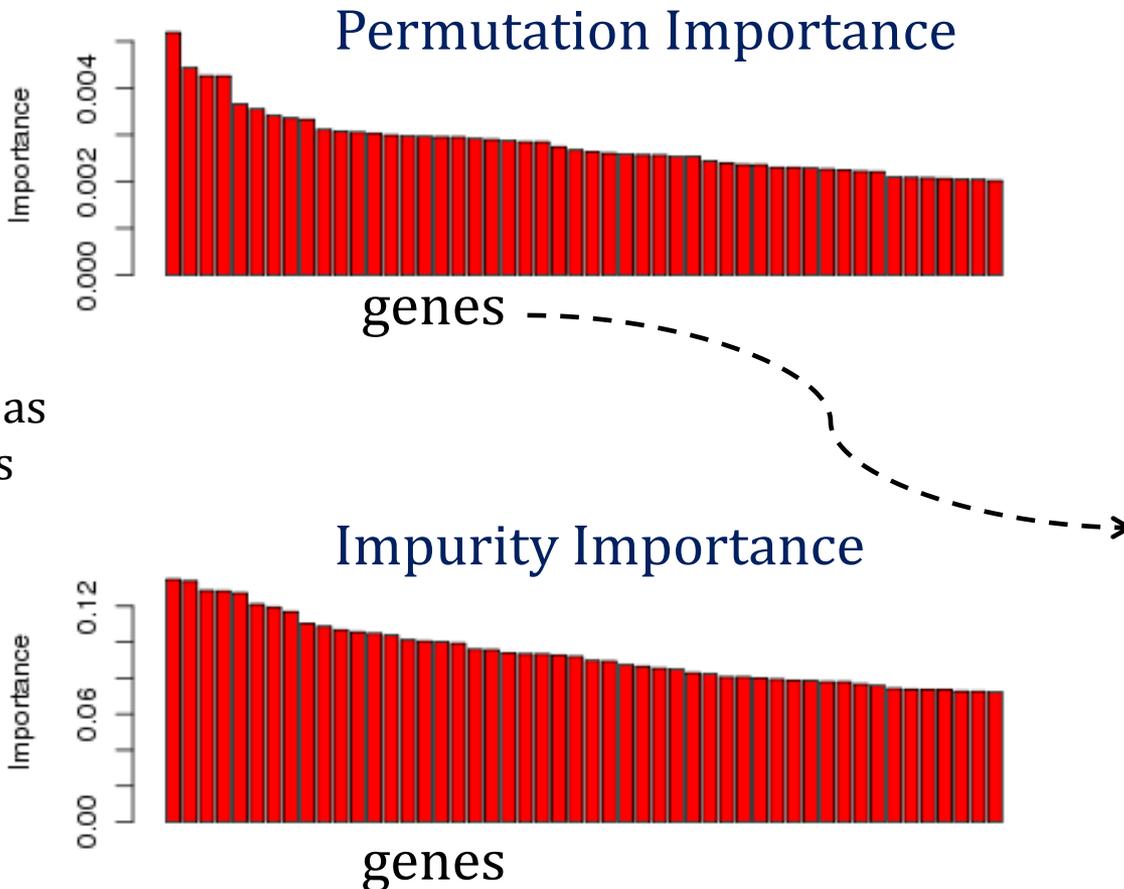
C4b	Complement component 4B
Gpnmb	Glycoproteine (transmembrane)
Pisd	Phosphatidylserine decarboxylase

- C4.5 algorithm (Quinlan 1993).
- Overfitting reduced by pruning

Random Forest: Variable Importance

- Ensemble classifier, builds many Decision Trees (Breiman, Cutler)
- random exclusion of a part of variables (and samples) in each tree
- **Variable importance:** measures explanatory power of a variable (gene)

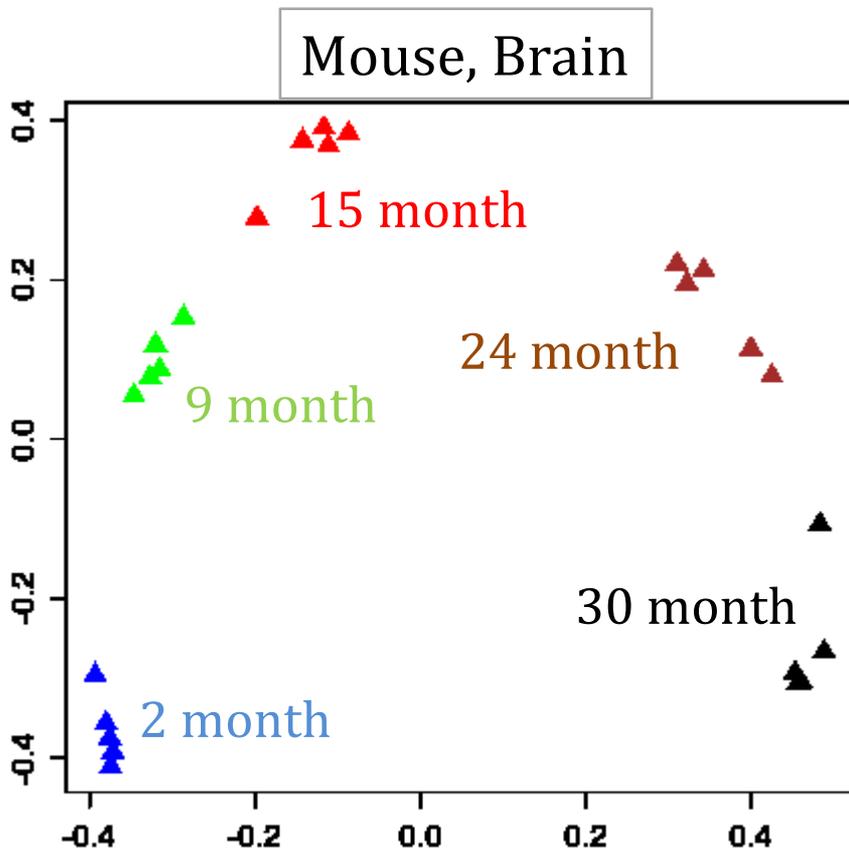
Putative
biomarkers:



Apc
C4b
Pisd-ps1
ing4
depdc1a
Pcdh20
Pisd
Plekhb1
Epha3
Tmem167
Pcdhb9
Gpnmb
Prkci

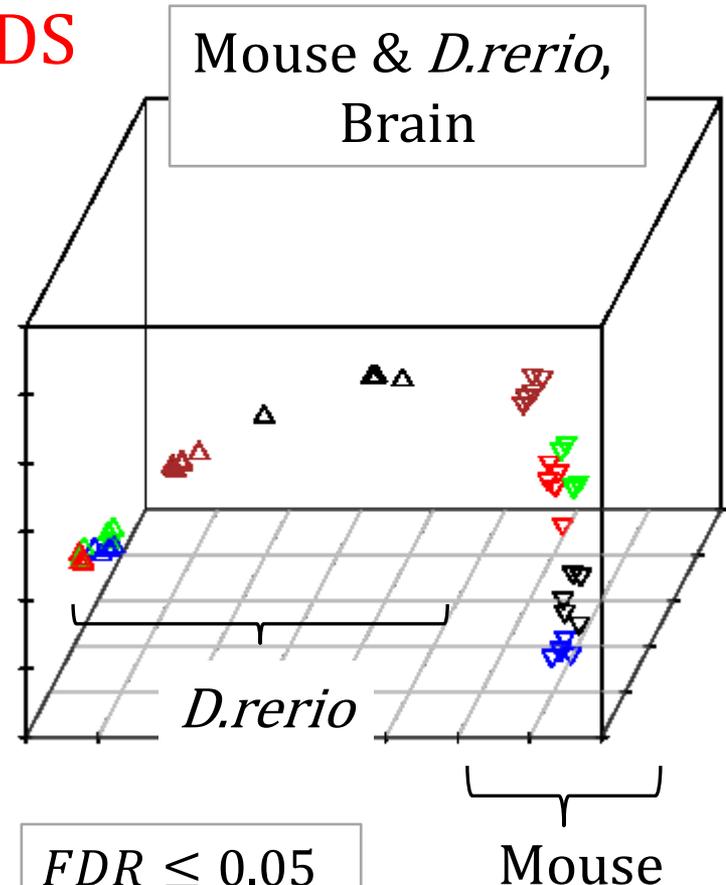
Random Forest: Sample Proximity

- A RF-classifier estimates the pairwise sample proximity (SP)
- samples are “close” if they end up in the same leaf frequently



$|\bar{\rho}| \geq 0.85$
no FDR or RPKM-filter

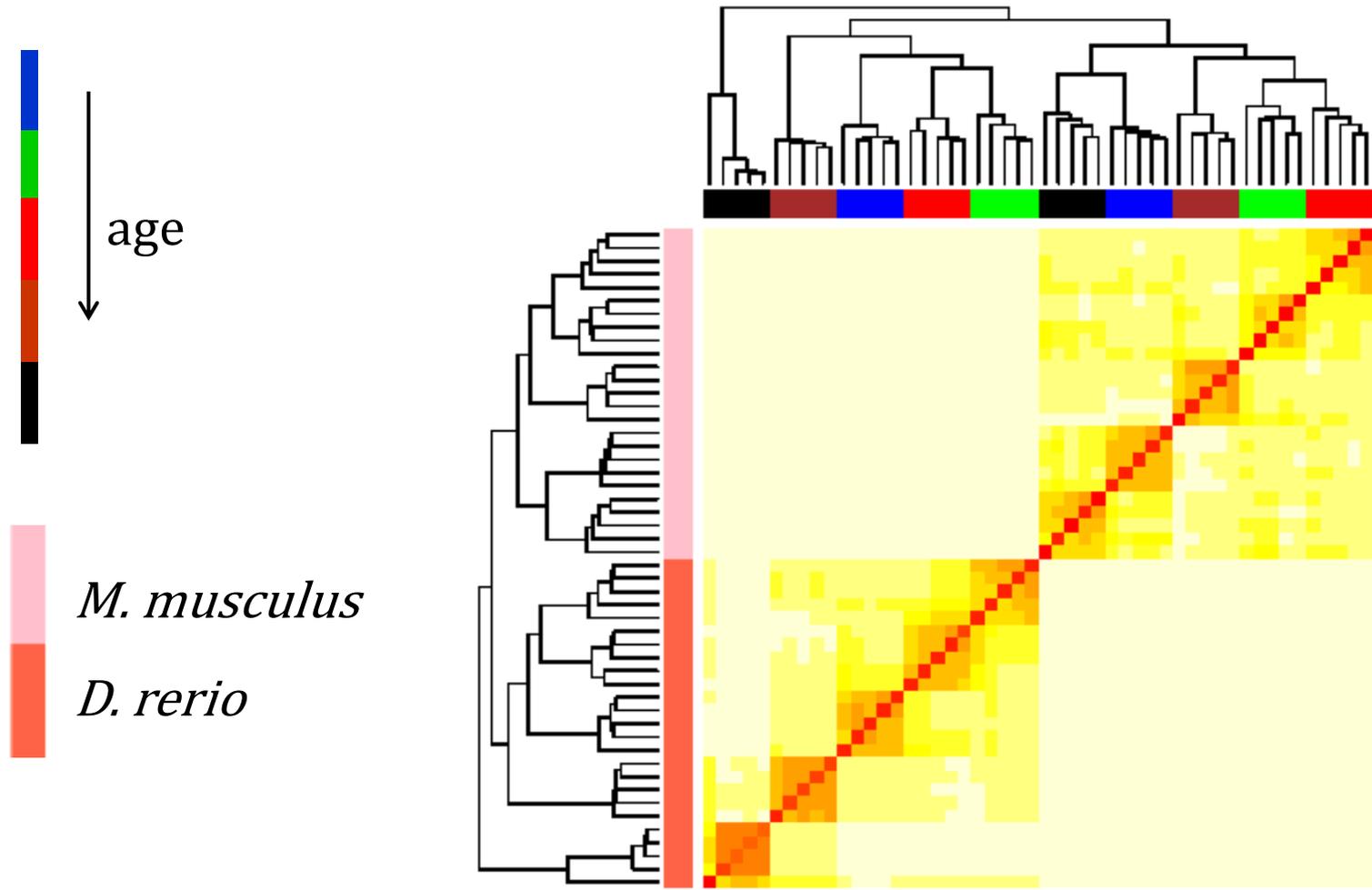
MDS



$FDR \leq 0.05$
 $|RPKM| \geq 1$
 $|\bar{\rho}| \geq 0.3$

Random Forest: Hierarchical clustering

- Hierarchical clustering for *M. musculus* and *D. rerio* based on proximity from Random Forest



Pathways with significantly changed co-expression

In the difference network:

- count fraction of connected genes in a pathway/GO-term
- count fraction of connected genes in the whole network
- pathways with disproportionately many connected genes

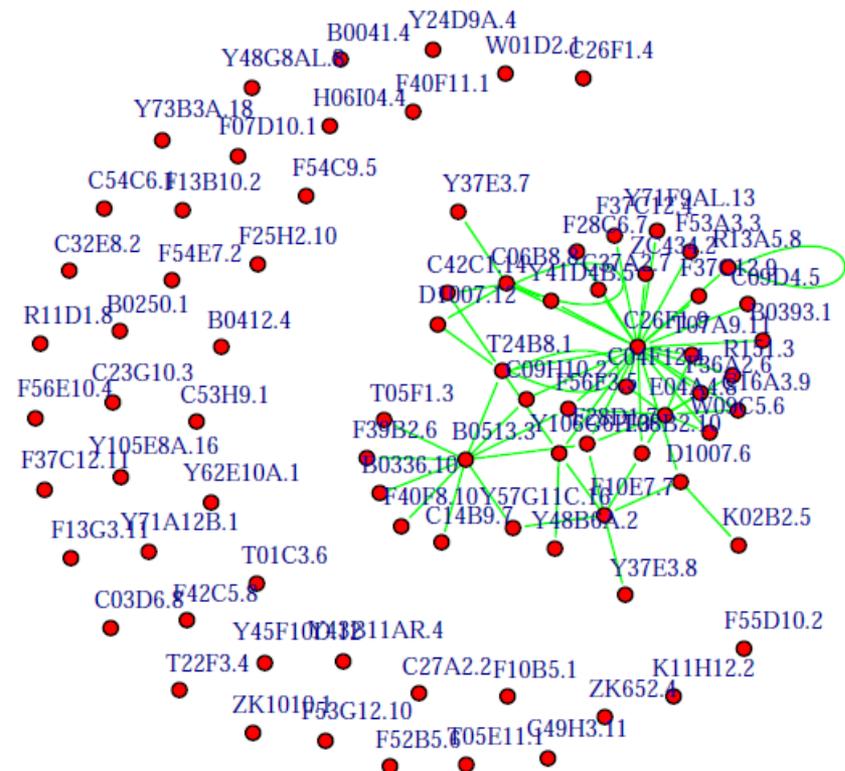
Fisher's Exact Test
fisher.test



Do this for:

- 2881 GO terms and
- 125 KEGG-pathways

DN: DOG-DMSO



Pathways with significantly changed co-expression

- KEGG-pathways whose co-expression is affected most by the perturbation (DOG, glucose restriction):

KEGG-ID	Term	p.adj	Genes	... connected
3010	Ribosome	0.0000	83	41
1100	Metabolic pathways	0.0000	596	214
190	Oxidative phosphorylation	0.0000	100	39
3018	RNA degradation	0.0159	37	9
350	Tyrosine metabolism	0.0396	19	3

- Inclusion of human cell lines into the analysis (orthology)
- Perturbed ageing for more species (currently Ce, Mm, Hs only)
- Dynamic models of hormesis connected to ageing
 - mTOR
 - preliminary results achieved
- Identification of relevant biomarkers for ageing, and of targets to support healthy ageing
- Wet-lab validation

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- Reinhard Guthke



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- Matthias Platzer
- Jürgen Sühnel



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- Michael Ristow
- Stefan Schuster



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