

Ako rozumieť genetickej rôznorodosti?

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GJH

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Mat, fyz, inf v biológii?

- 80 000 000 000 neurónov, ako sú poprepájané?
 - Elektrické napätie v čase zo stoviek z nich naraz
- Genóm *ATGC ... CTGA*, dlhý 3 000 000 000 báz
 - Skladanie sekvencie z miliónov chybových utržkov
 - Porovnávanie DNA 100 000 jedincov
- 20 000 rôznych génov, ~10 000 000 *bežných* mutácií
 - Meranie koncentrácie 20 000 druhov RNA naraz v 10 000 bunkách naraz

Veľké otázky genetiky

Genotyp → fenotyp

Do akej miery sú vlastnosti dedičné? (Nature vs. Nurture)

Kde v genóme sú zakódované?

Akým spôsobom sa efekty mutácií skladajú?

Zopár silných vs. veľa slabých efektov?

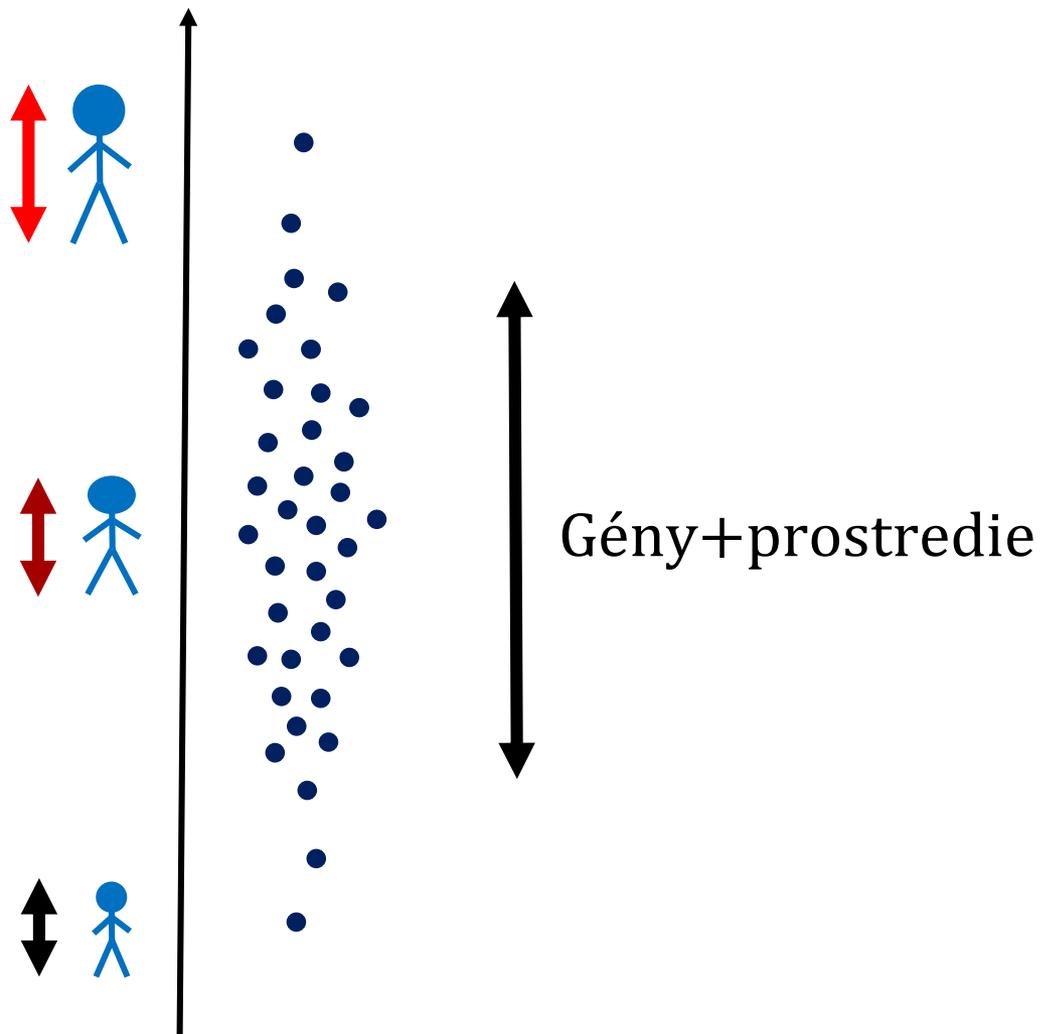
Bežné vs. vzácne mutácie?

Porozumenie evolúcii, individualizovaná diagnostika a liečba

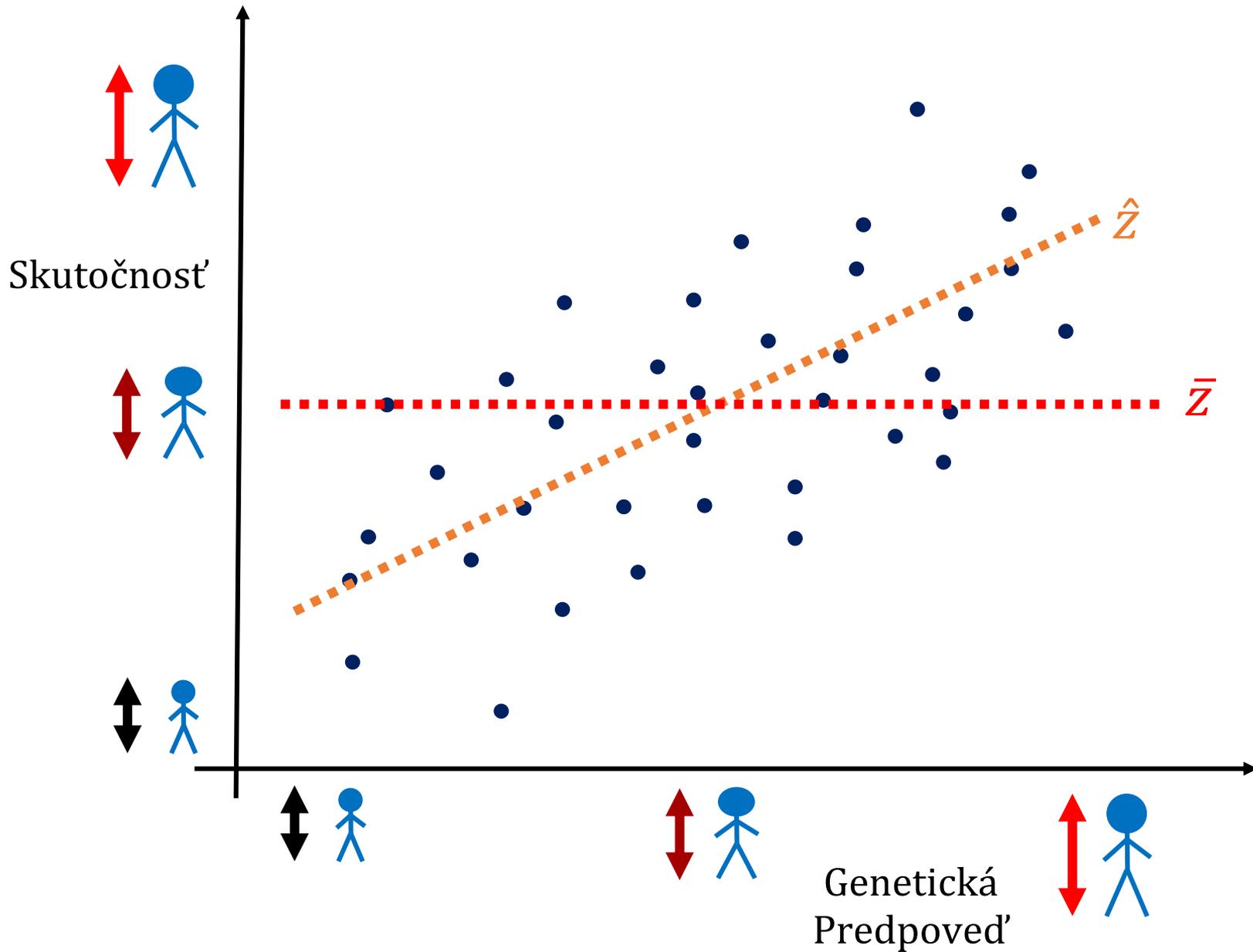
<https://www.genomicsengland.co.uk/the-100000-genomes-project/understanding-genomics/>

Klasická genetika

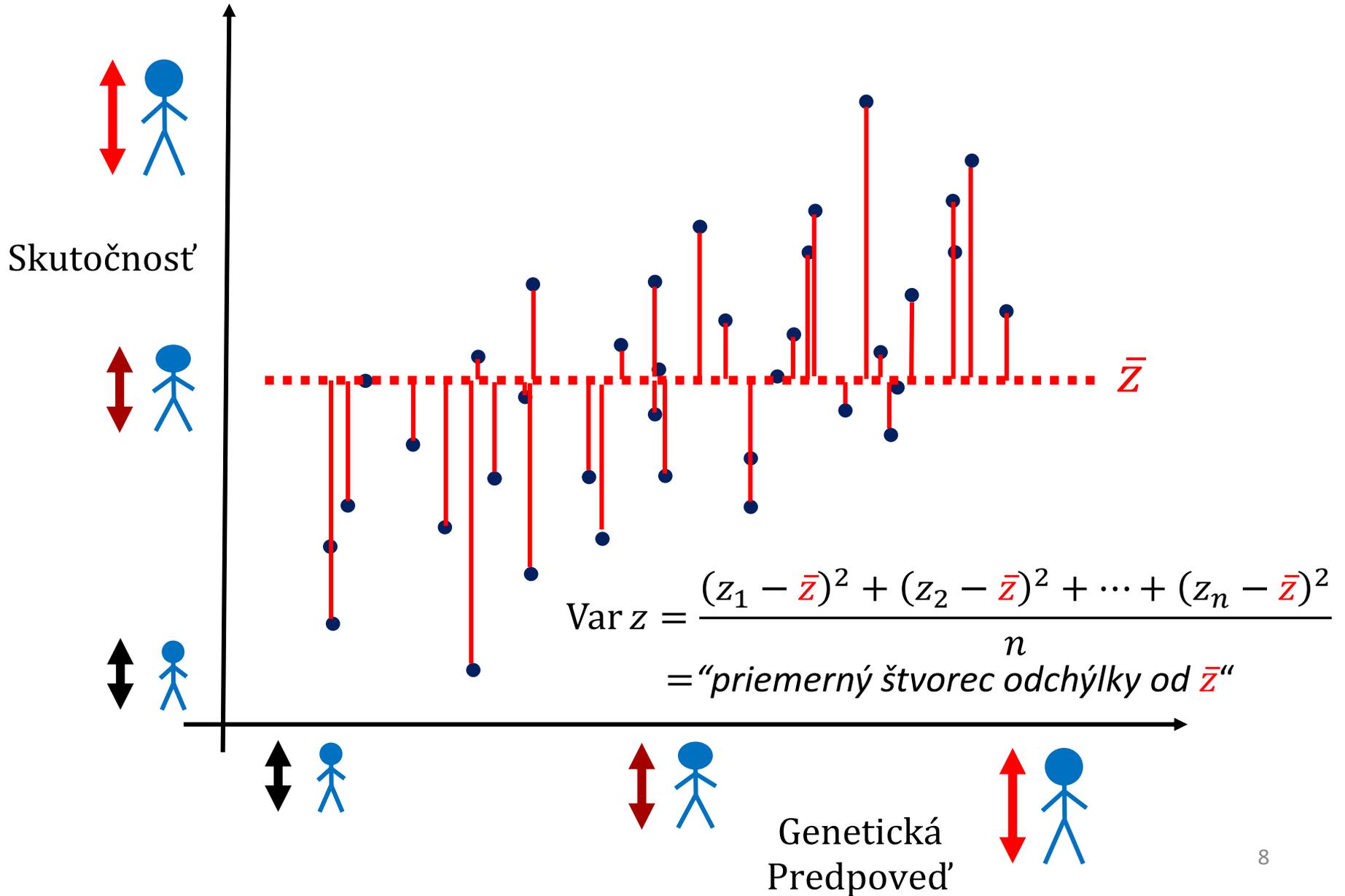
Výška z



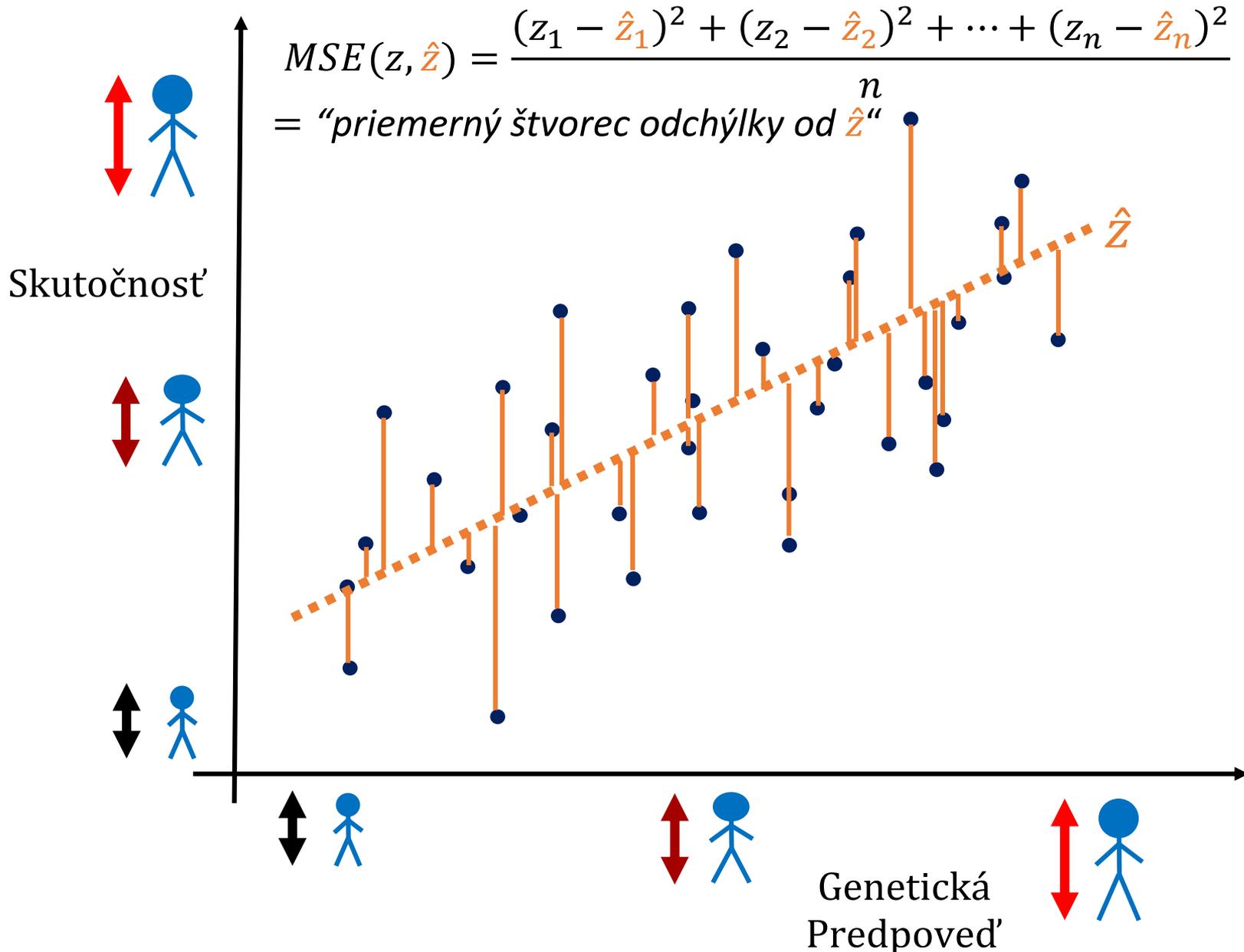
Výška z



Výška z



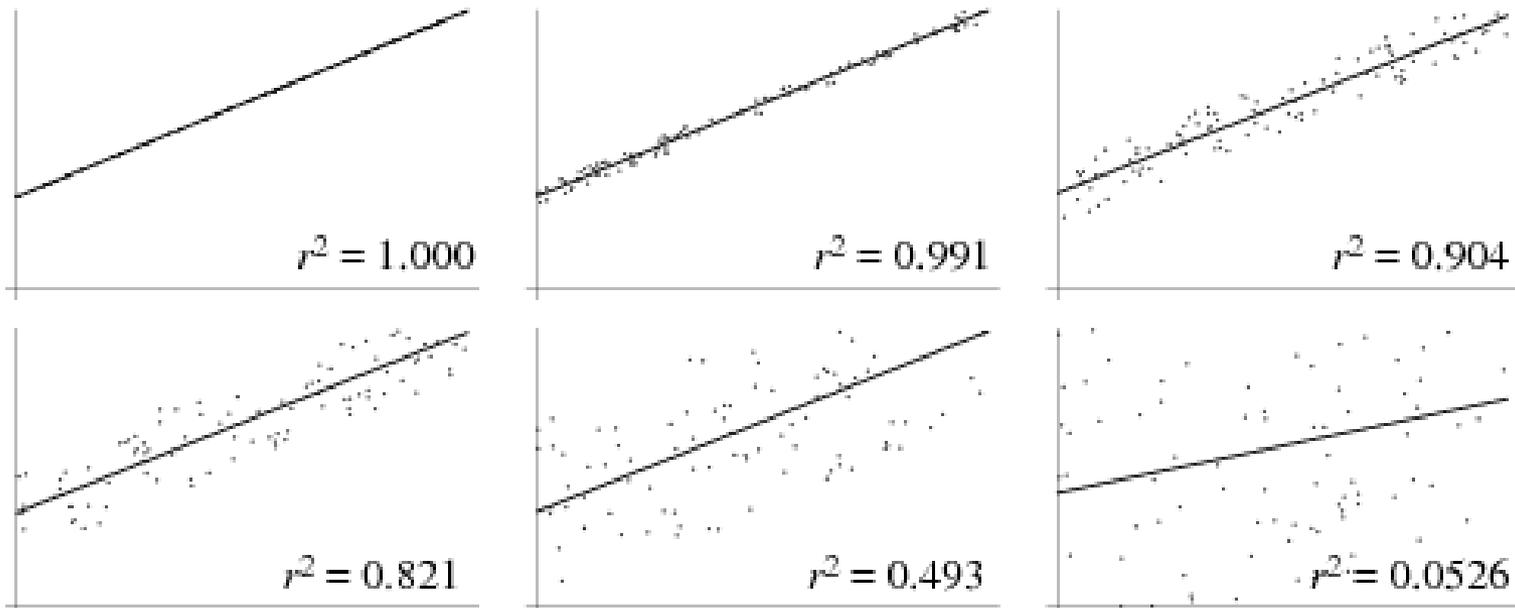
Výška z



Dedičnosť: $H^2 = 1 - \frac{MSE(z, \hat{z})}{\text{Var } z}$
 = „podiel variancie vysvetlený genetikou“

Aditívny model: $z = z_{mut1} + z_{mut2} + \dots + z_{Prostredie}$

Dedičnosť $h^2 \approx 2(r_{jednovaječné}^2 - r_{dvojvaječné\ dvojčky}^2)$



Klasická genetika

Základné pravidlo biológie:

Vždy je to zložitejšie, než si myslíš.

- Neaditívne efekty génov?
 - Dáta: aspoň 50% dedičnosti je aditívna*
- Efekty génov závisia na prostredí?
- Dedičnosť sa v čase mení

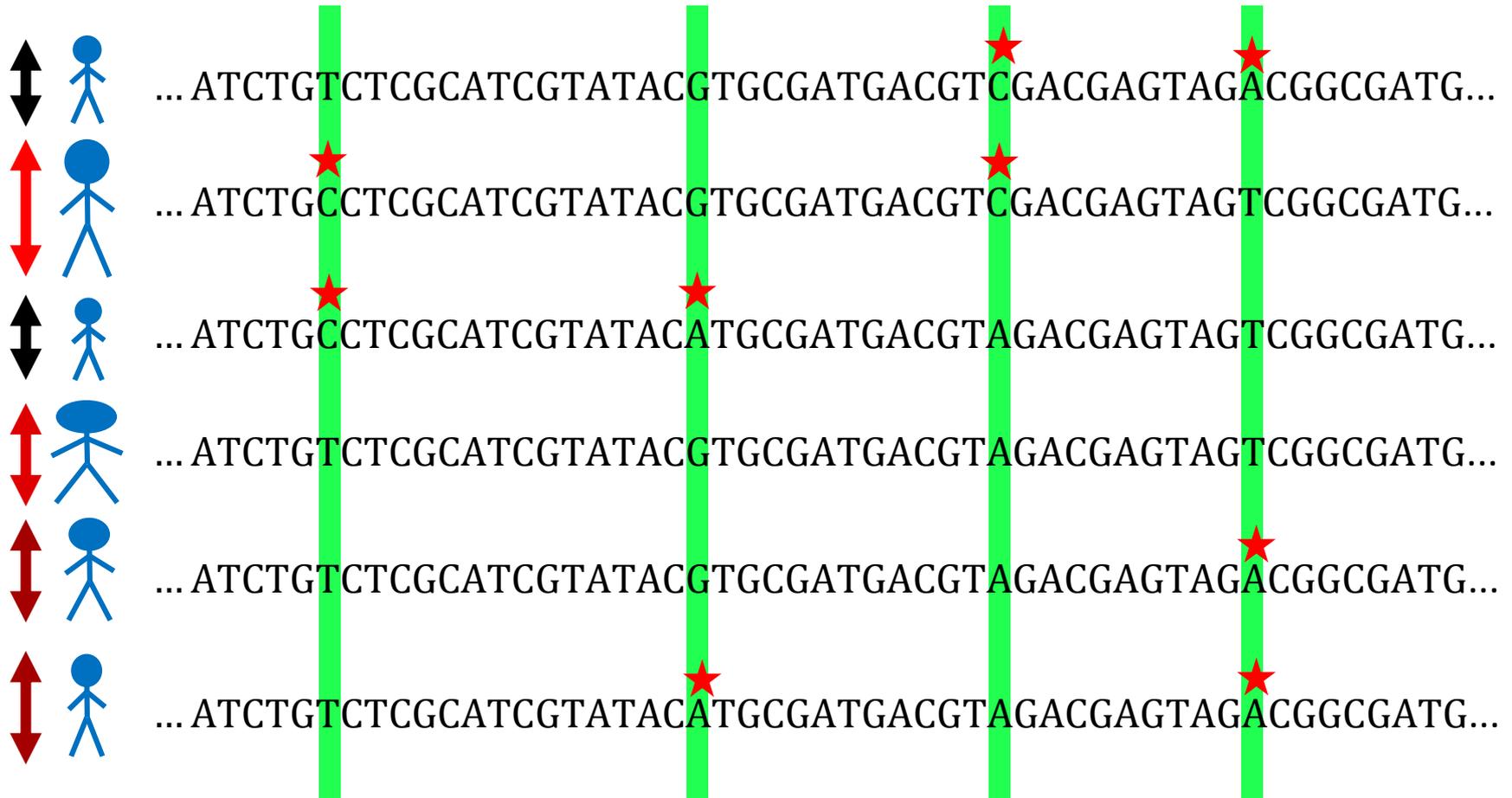
*William G. Hill, Michael E. Goddard, Peter M. Visscher, *Data and Theory Point to Mainly Additive Genetic Variance for Complex Traits*. PLoS Genet 4(2): e1000008 (2008). doi:10.1371/journal.pgen.1000008

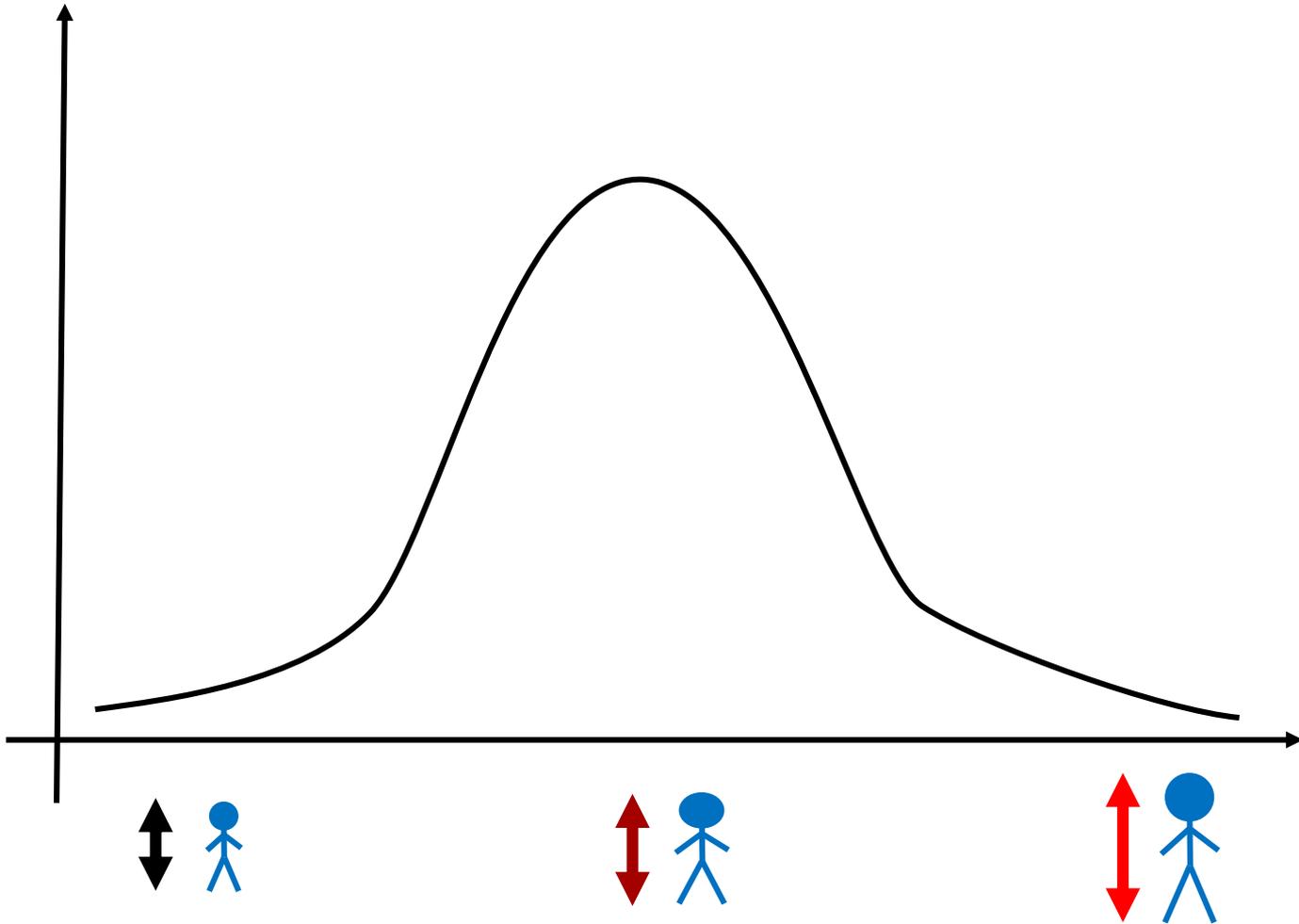
Trait or Disease	h^2 Pedigree Studies
Type 1 diabetes	0.9 ⁹⁸
Type 2 diabetes	0.3–0.6 ¹⁰⁰
Obesity (BMI)	0.4–0.6 ^{101,102}
Crohn's disease	0.6–0.8 ¹⁰³
Ulcerative colitis	0.5 ¹⁰³
Multiple sclerosis	0.3–0.8 ¹⁰⁴
Ankylosing spondylitis	>0.90 ¹⁰⁵
Rheumatoid arthritis	0.6 ¹⁰⁷
Schizophrenia	0.7–0.8 ¹⁰⁸
Bipolar disorder	0.6–0.7 ¹⁰⁸
Breast cancer	0.3 ¹¹⁰
Von Willebrand factor	0.66–0.75 ^{112,113}
Height	0.8 ^{115,116}
Bone mineral density	0.6–0.8 ¹¹⁷
QT interval	0.37–0.60 ^{119,120}
HDL cholesterol	0.5 ¹²²
Platelet count	0.8 ¹²³

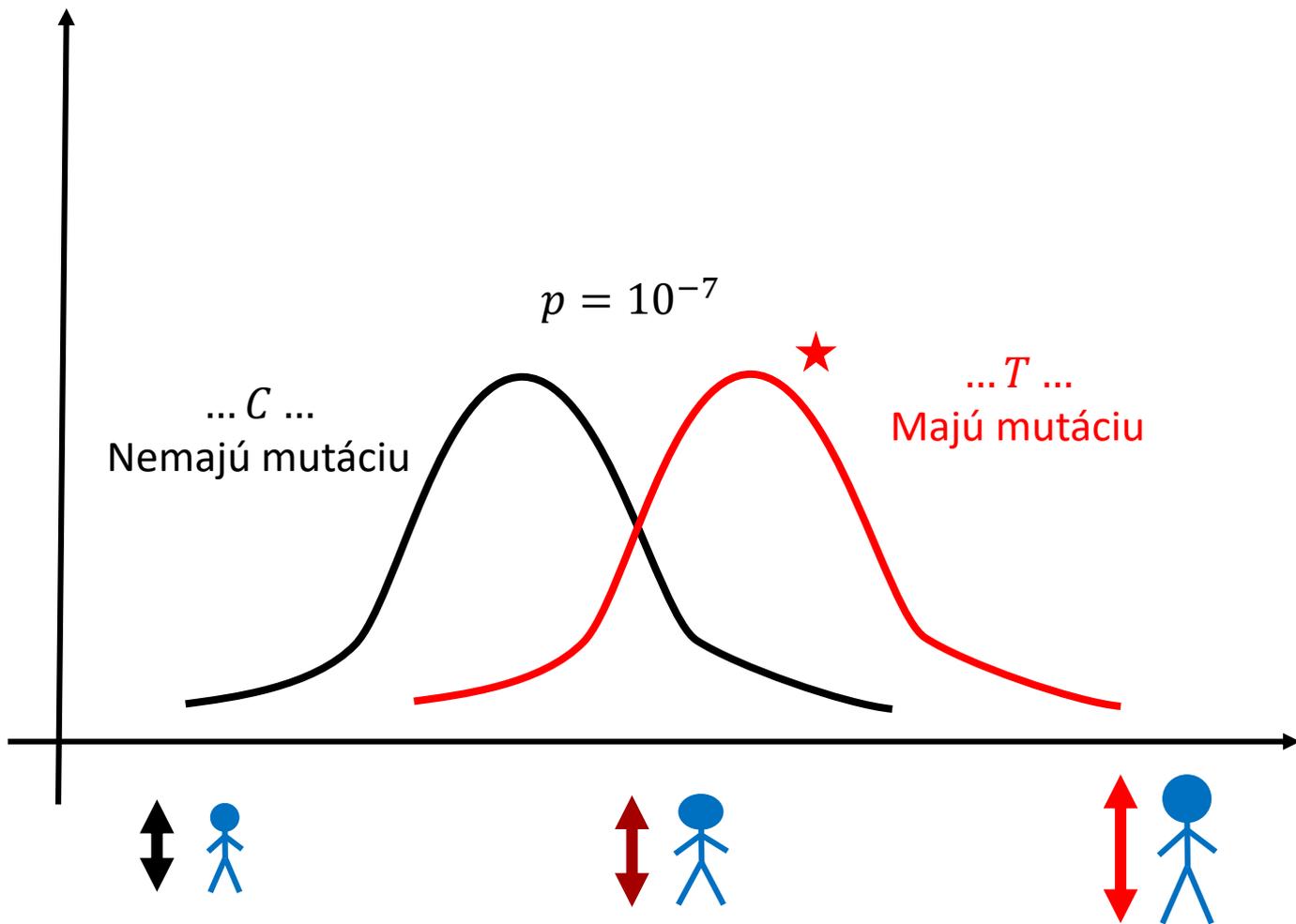
Peter M. Visscher, Matthew A. Brown, Mark I. McCarthy, and Jian Yang (2012). *Five Years of GWAS Discovery*. The American Journal of Human Genetics 90, 7–24

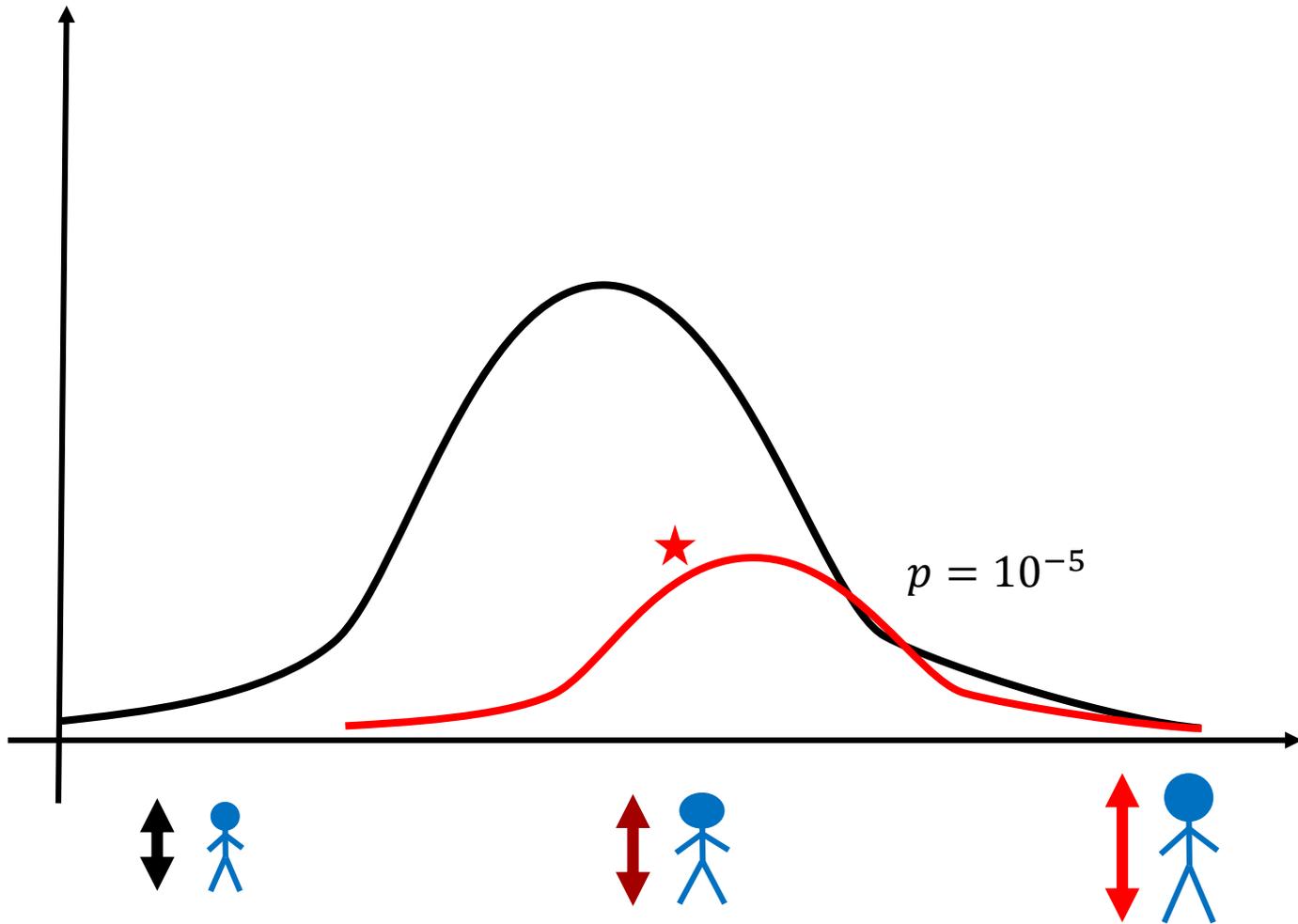
GWAS

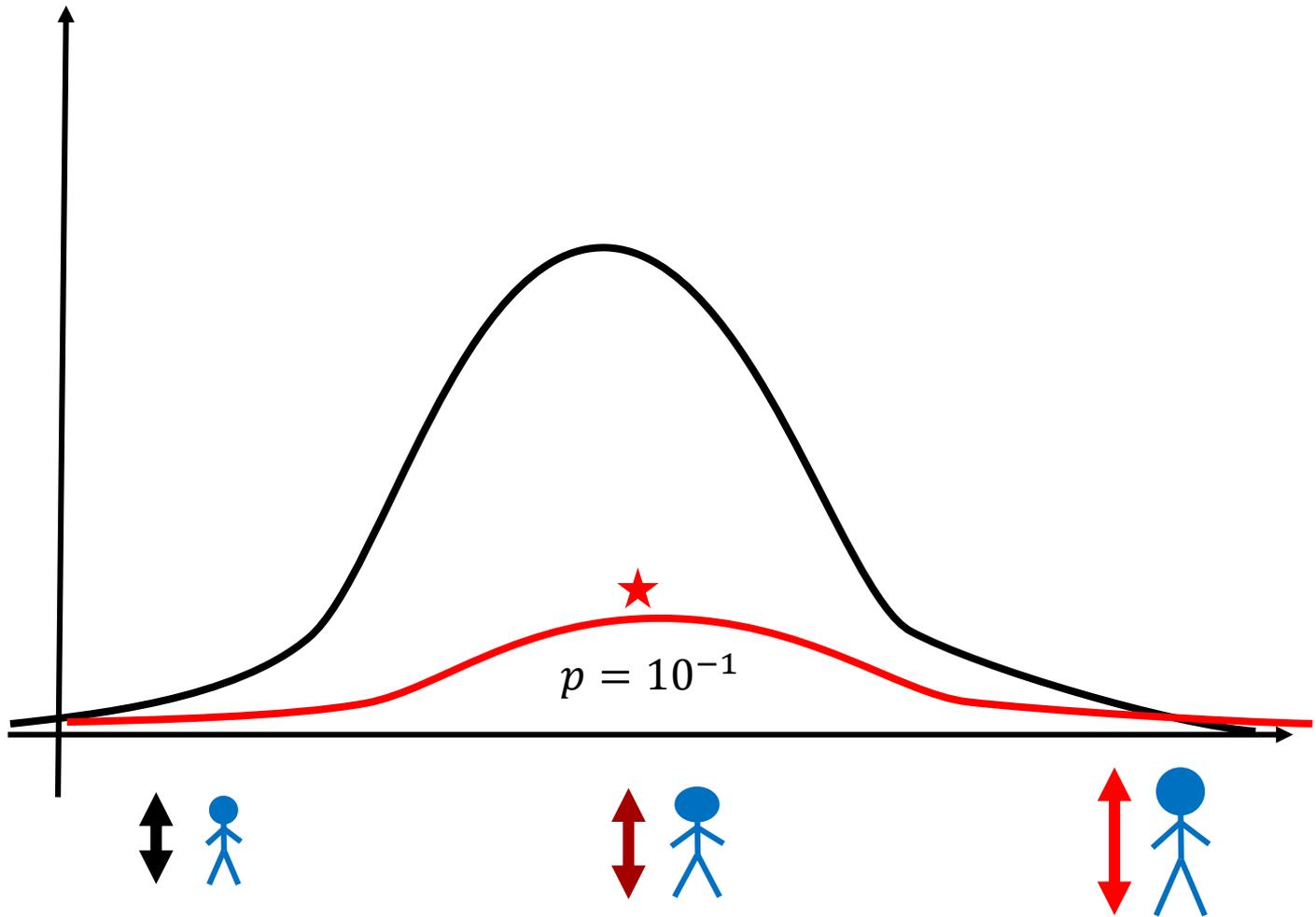
Genome-wide association studies

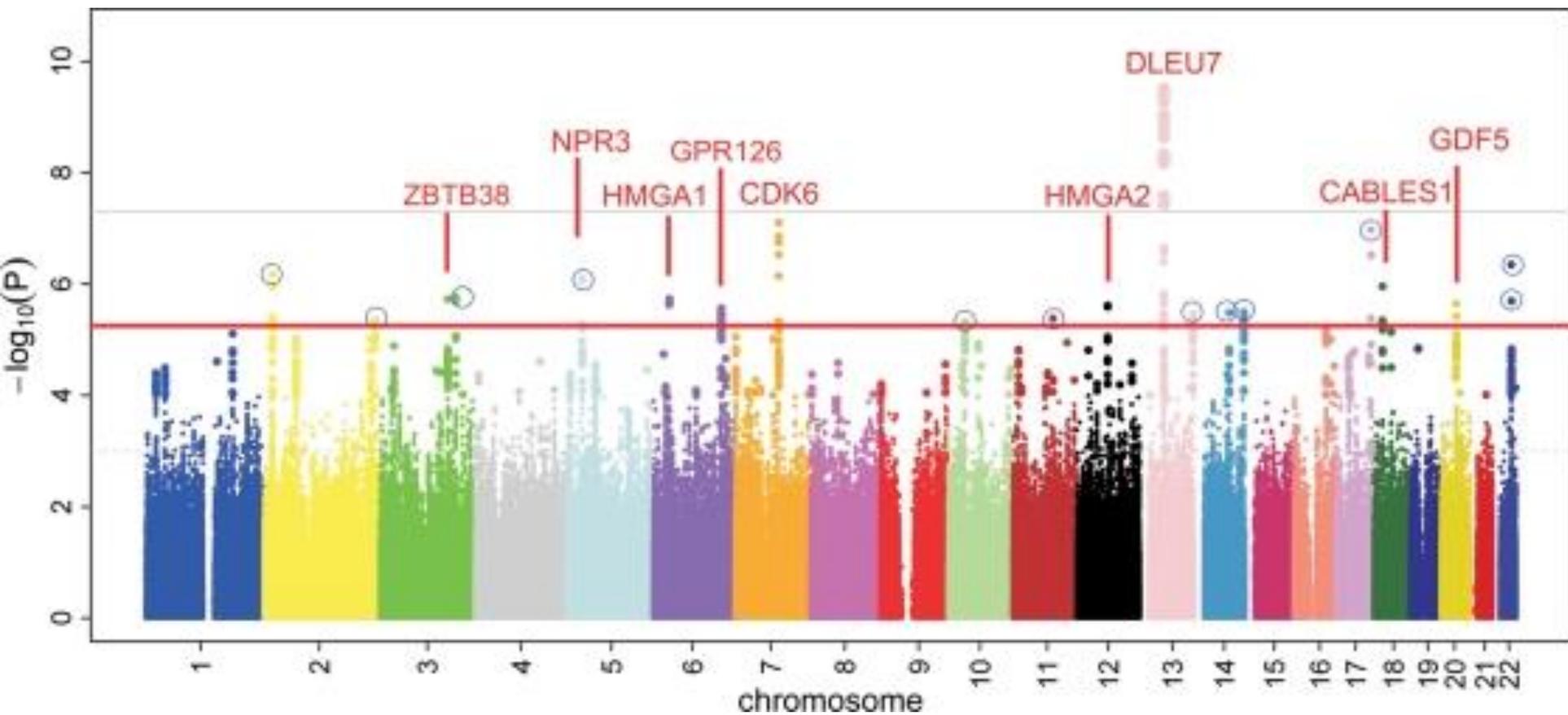




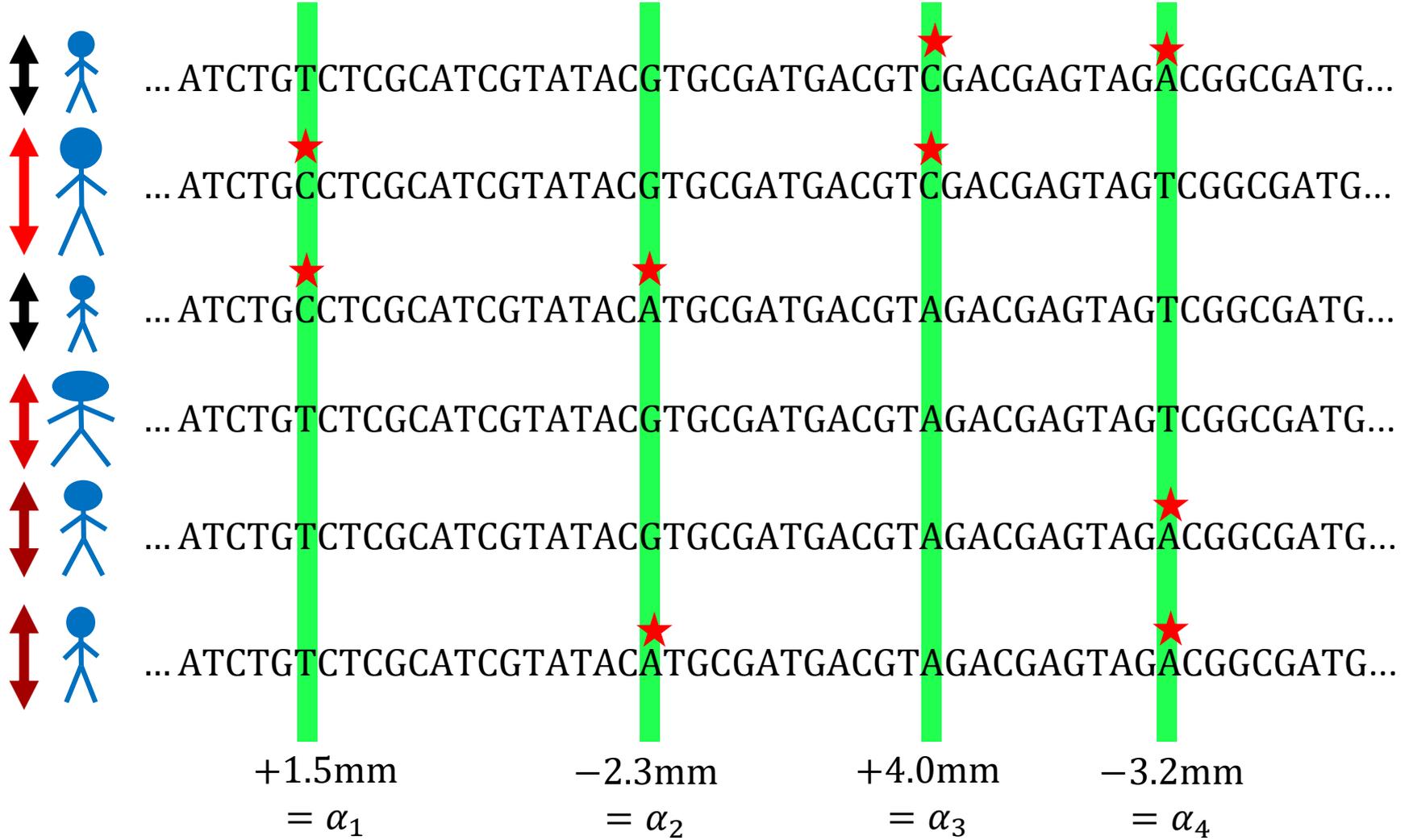








Estrada, Karol et al. (2009). A genome-wide association study of northwestern Europeans involves the C-type natriuretic peptide signaling pathway in the etiology of human height variation. *Human molecular genetics*. 18. 3516-24. 10.1093/hmg/ddp296.



$$\hat{z}_1 = 0\alpha_1 + 0\alpha_2 + 1\alpha_3 + 1\alpha_4 + \dots$$

$$\hat{z}_2 = 1\alpha_1 + 0\alpha_2 + 1\alpha_3 + 0\alpha_4 + \dots$$

⋮

$$\hat{z}_i = x_{i1}\alpha_1 + x_{i2}\alpha_2 + x_{i3}\alpha_3 + x_{i4}\alpha_4 + \dots$$

Table 1. Population Variation Explained by GWAS for a Selected Number of Complex Traits

Trait or Disease	h² Pedigree Studies	h² GWAS Hits^a	h² All GWAS SNPs^b
Type 1 diabetes	0.9 ⁹⁸	0.6 ^{99, c}	0.3 ¹²
Type 2 diabetes	0.3–0.6 ¹⁰⁰	0.05–0.10 ³⁴	
Obesity (BMI)	0.4–0.6 ^{101,102}	0.01–0.02 ³⁶	0.2 ¹⁴
Crohn's disease	0.6–0.8 ¹⁰³	0.1 ¹¹	0.4 ¹²
Ulcerative colitis	0.5 ¹⁰³	0.05 ¹²	
Multiple sclerosis	0.3–0.8 ¹⁰⁴	0.1 ⁴⁵	
Ankylosing spondylitis	>0.90 ¹⁰⁵	0.2 ¹⁰⁶	
Rheumatoid arthritis	0.6 ¹⁰⁷		
Schizophrenia	0.7–0.8 ¹⁰⁸	0.01 ⁷⁹	0.3 ¹⁰⁹
Bipolar disorder	0.6–0.7 ¹⁰⁸	0.02 ⁷⁹	0.4 ¹²
Breast cancer	0.3 ¹¹⁰	0.08 ¹¹¹	
Von Willebrand factor	0.66–0.75 ^{112,113}	0.13 ¹¹⁴	0.25 ¹⁴
Height	0.8 ^{115,116}	0.1 ¹³	0.5 ^{13,14}
Bone mineral density	0.6–0.8 ¹¹⁷	0.05 ¹¹⁸	
QT interval	0.37–0.60 ^{119,120}	0.07 ¹²¹	0.2 ¹⁴
HDL cholesterol	0.5 ¹²²	0.1 ⁵⁷	
Platelet count	0.8 ¹²³	0.05–0.1 ⁵⁸	

Defining the role of common variation in the genomic and biological architecture of adult human height

A full list of authors and affiliations appears at the end of the article.

These authors contributed equally to this work.

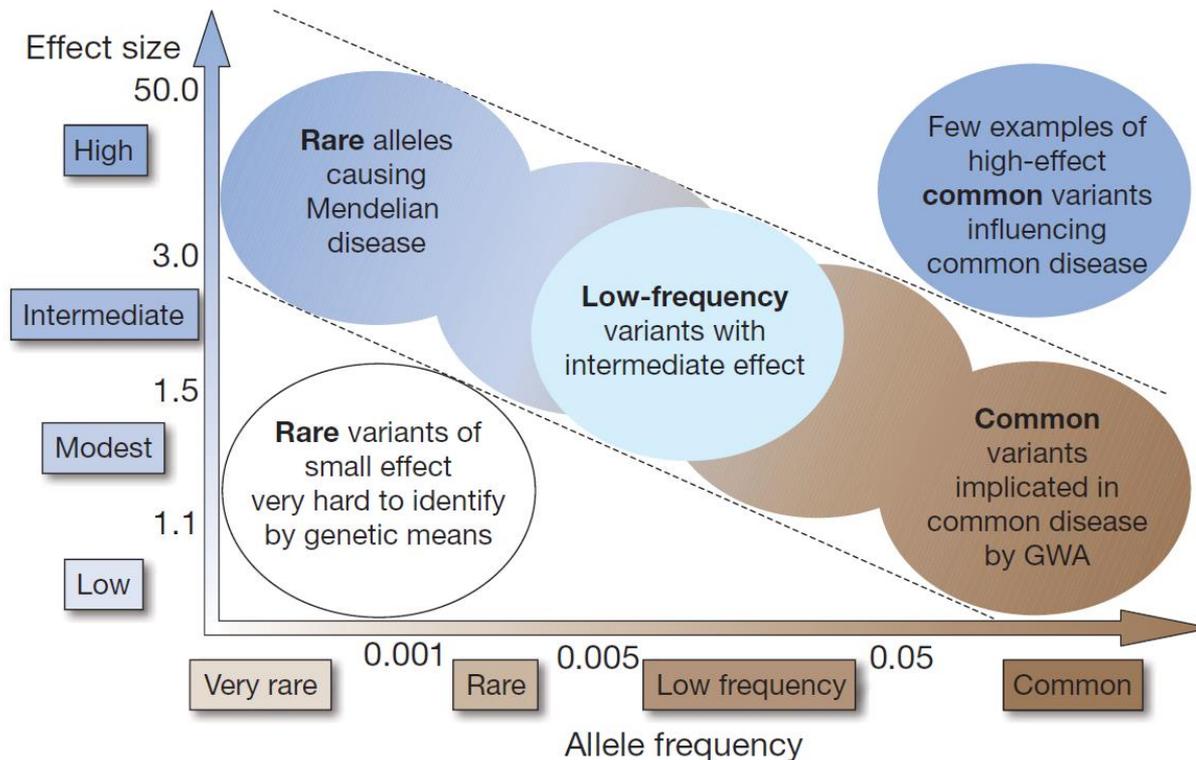
Abstract

Using genome-wide data from 253,288 individuals, we identified 697 variants at genome-wide significance that together explain one-fifth of heritability for adult height. By testing different numbers of variants in independent studies, we show that the most strongly associated ~2,000, ~3,700 and ~9,500 SNPs explained ~21%, ~24% and ~29% of phenotypic variance. Furthermore, all common variants together captured the majority (60%) of heritability. The 697 variants clustered in 423 loci enriched for genes, pathways, and tissue-types known to be involved in growth and together implicated genes and pathways not highlighted in earlier efforts, such as signaling by fibroblast growth factors, WNT/beta-catenin, and chondroitin sulfate-related genes. We identified several genes and pathways not previously connected with human skeletal growth, including mTOR, osteoglycin and binding of hyaluronic acid. Our results indicate a genetic architecture for human height that is characterized by a very large but finite number (thousands) of causal variants.

GWAS

Základné pravidlo biológie...

- **Veľmi veľa veľmi malých efektov**
- Neadaptívne efekty génov? → Deep learning?
- Prostredie, korelácie medzi mutáciami, ...



Teri A. Manolio, Francis S. Collins, Peter M. Visscher (2009). *Finding the missing heritability of complex diseases*. Nature vol. 461, pp. 747–753

Populačná genetika

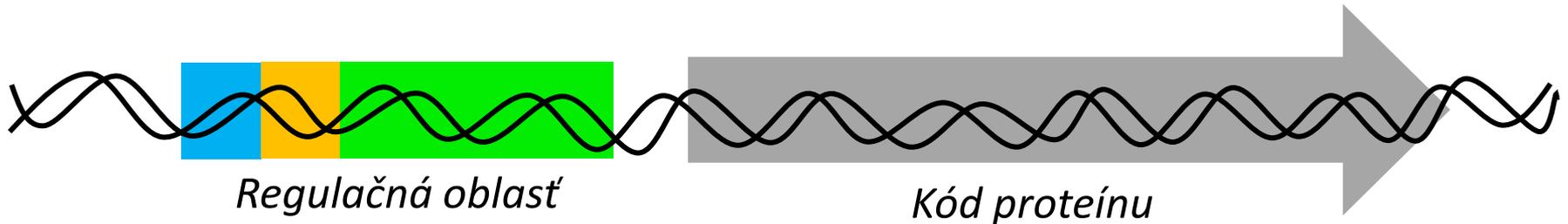
(simulácie v Mathematica)

Čo robia mutácie?

Expresia a regulácia génov

Centrálne dogma: DNA → RNA → Protein

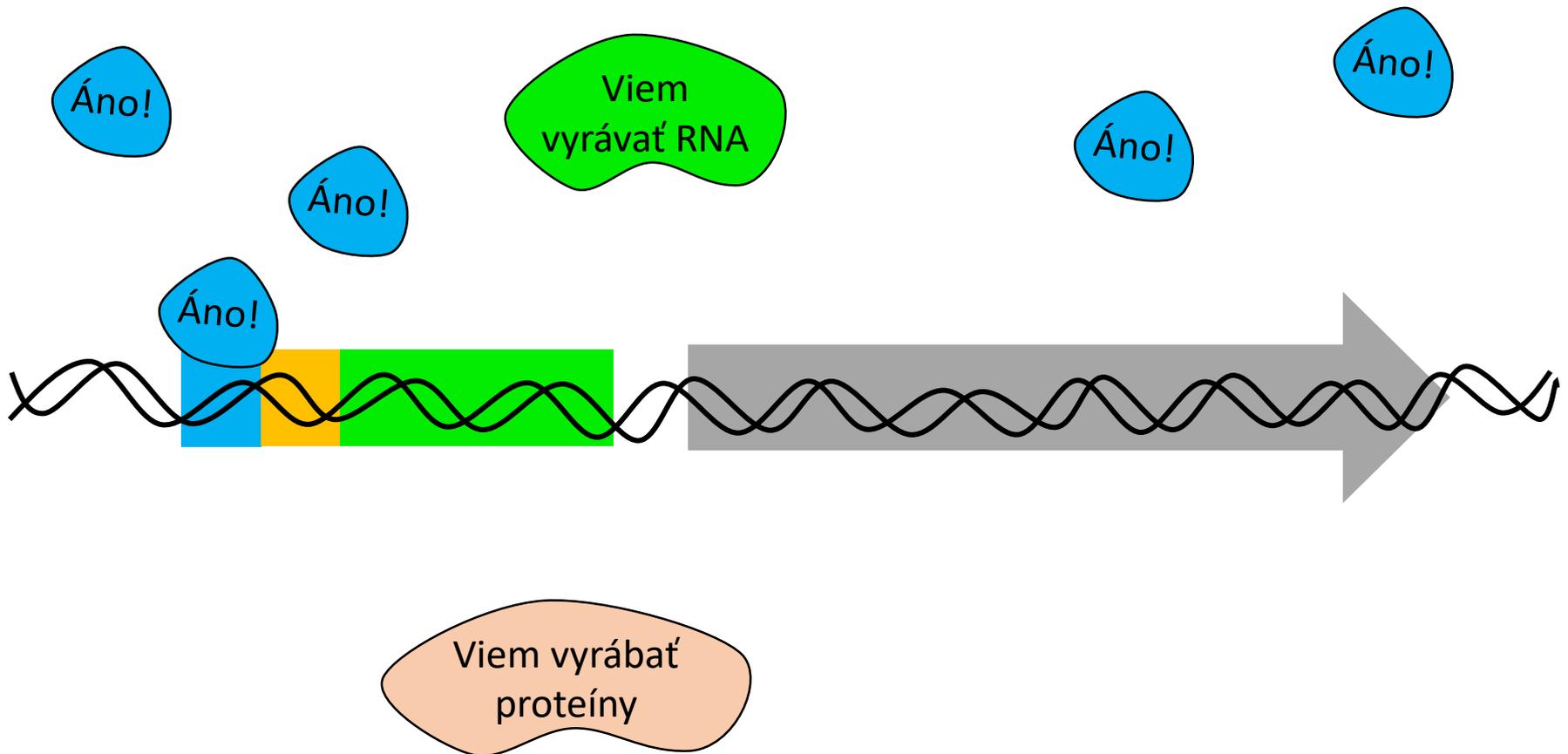
Ja viem
vyrábať RNA



Ja viem vyrábať
proteíny

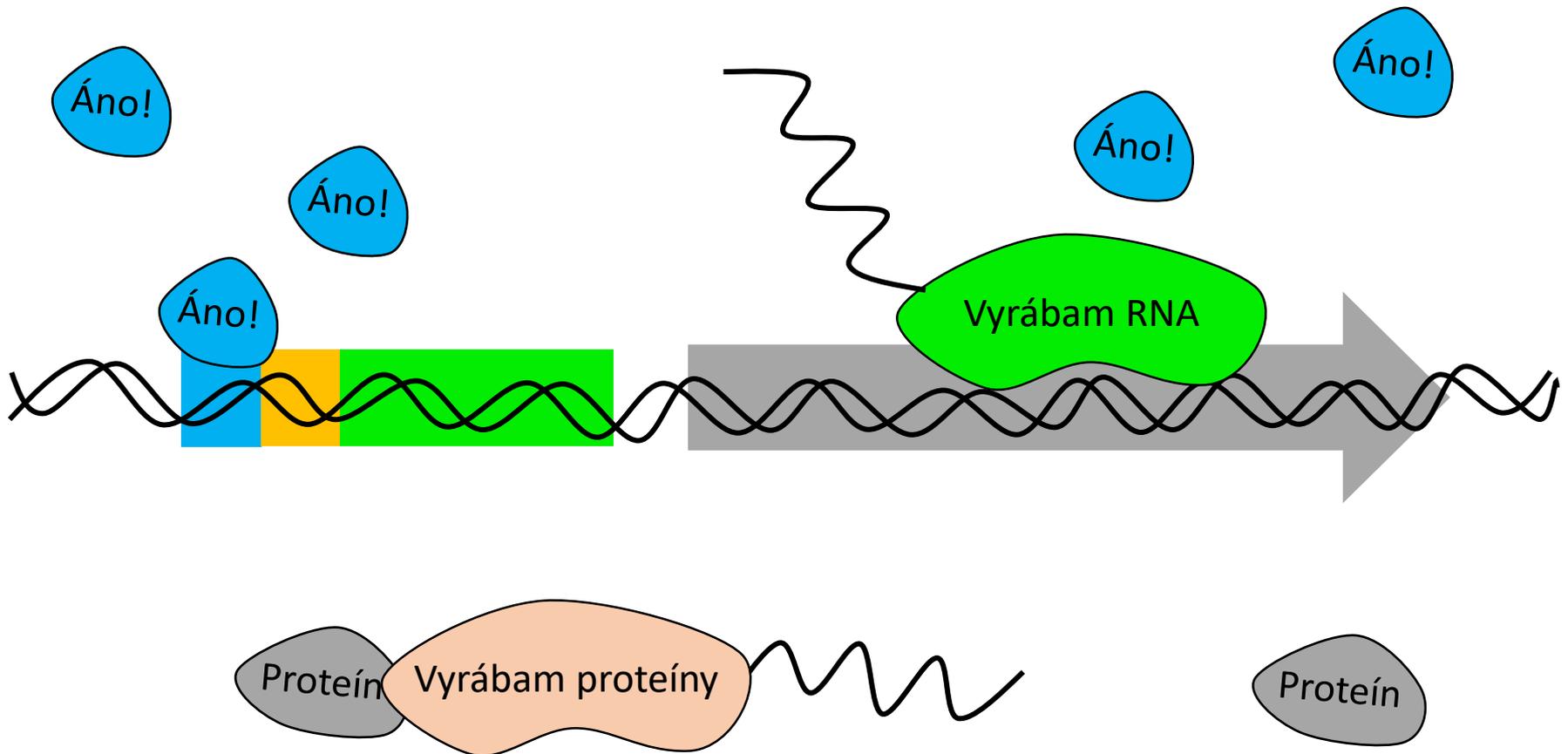
Expresia a regulácia génov

Centrálne dogma: DNA → RNA → Protein



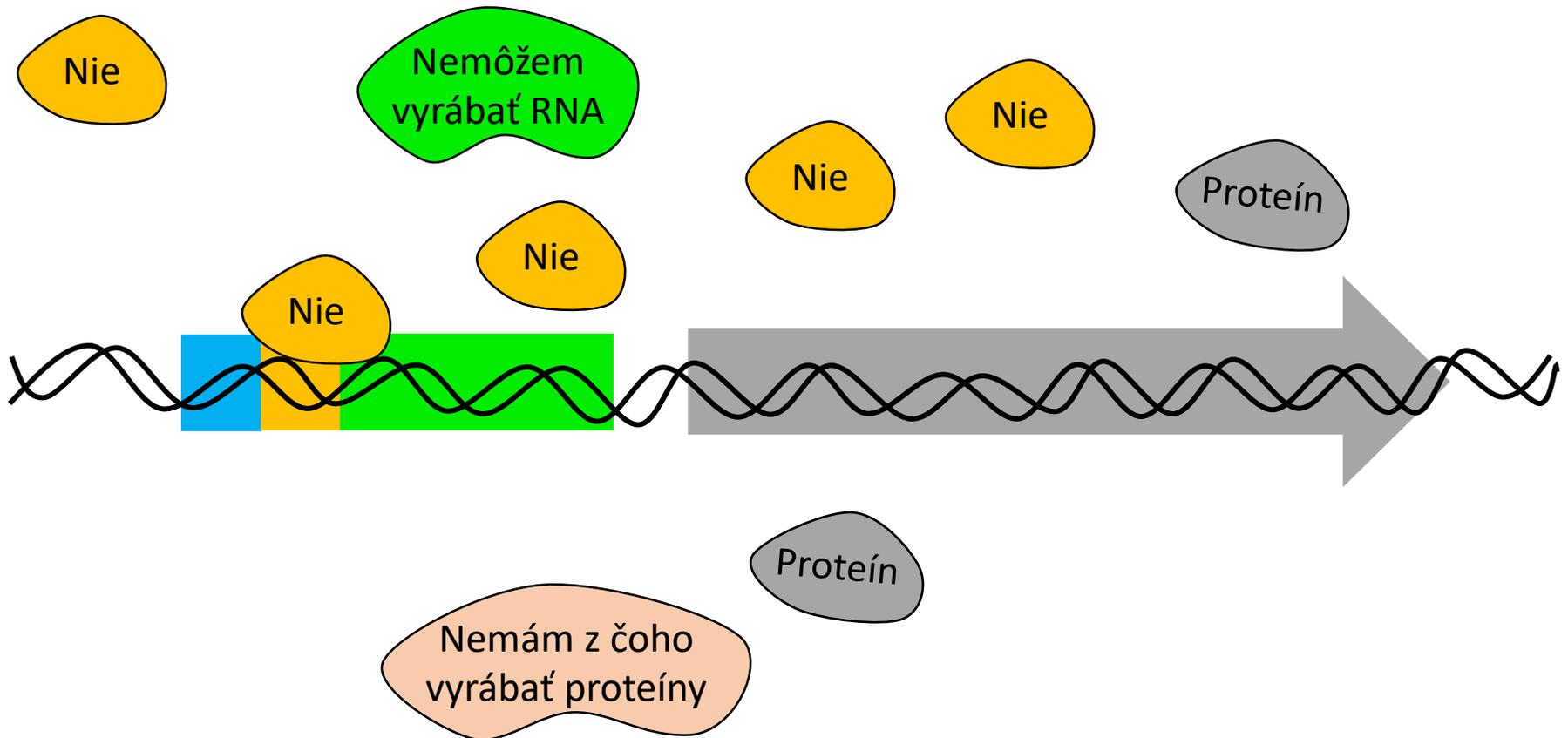
Expresia a regulácia génov

Centrálne dogma: DNA → RNA → Protein

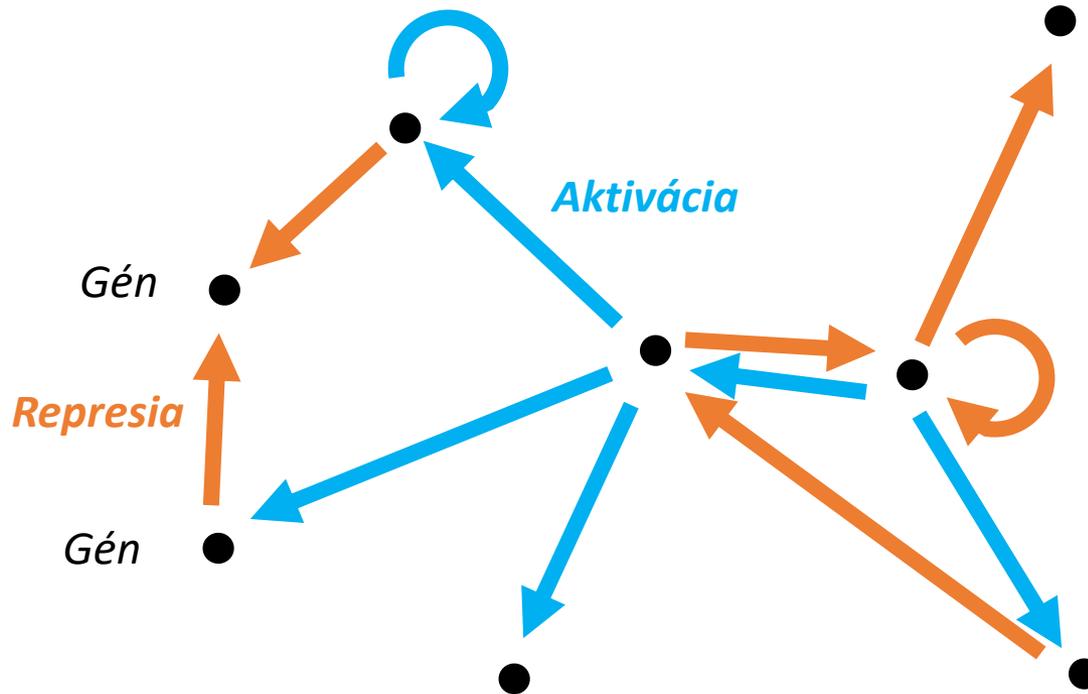


Expresia a regulácia génov

Centrálne dogma: DNA → RNA → Protein



Genetická regulačná sieť



a

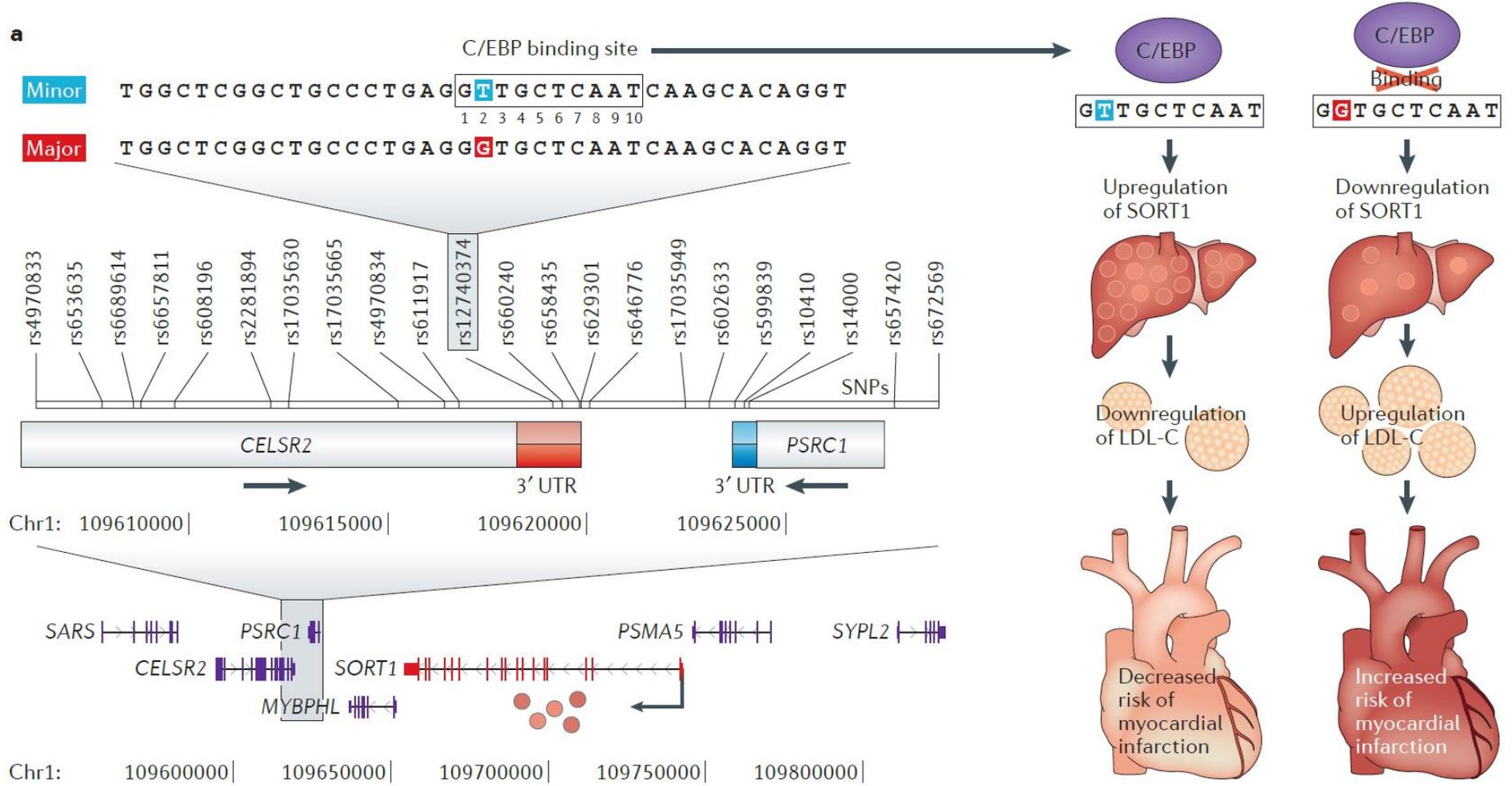
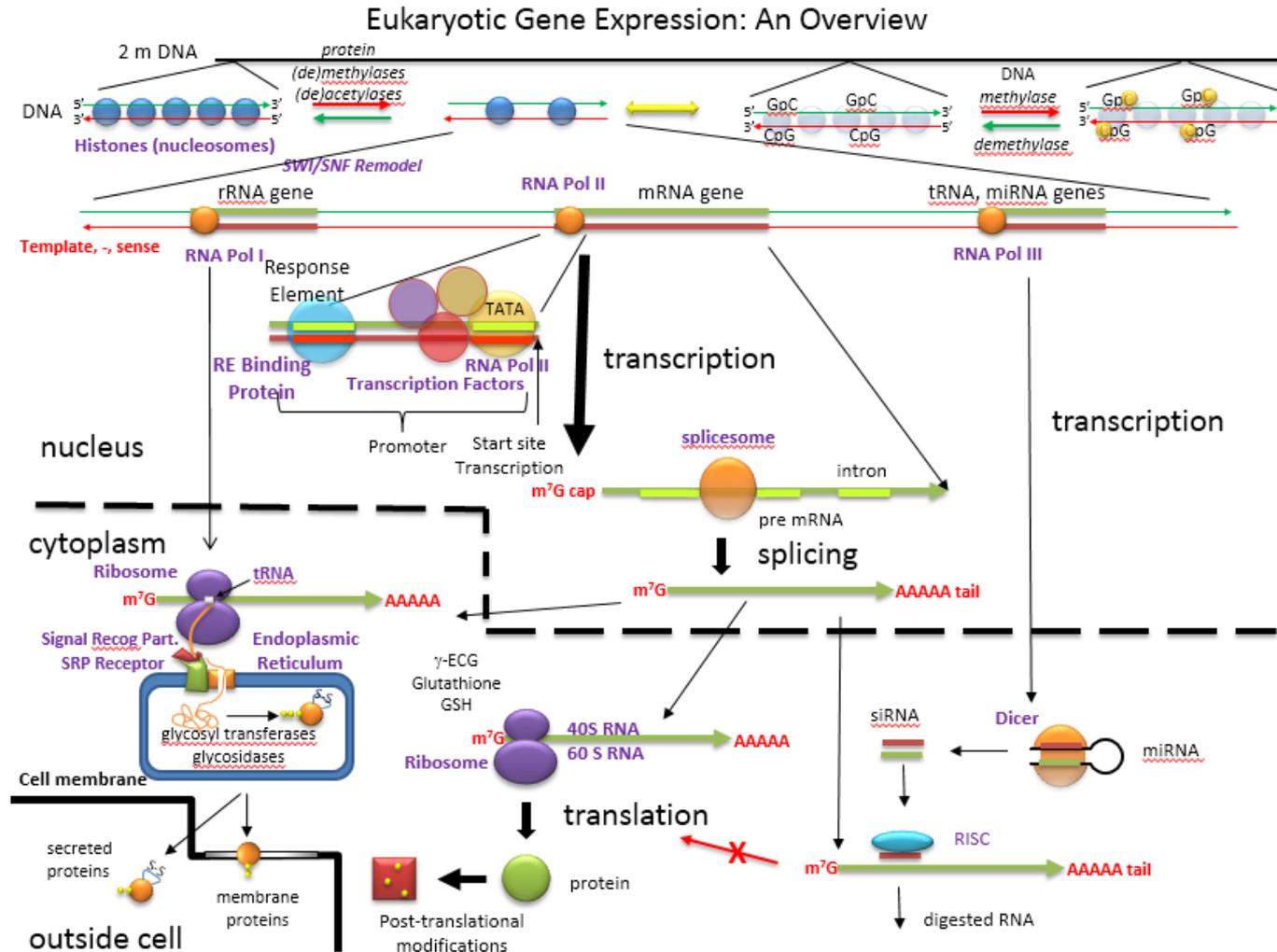


Table 1. Examples Linking Variable TF-DNA Binding to Phenotypic Variation Arranged by Date of Characterization

Phenotype	Affected Gene	Causal Variant Position Relative to TSS	Affected Binding Site	TFBS Outcome	Reference(s)
Hereditary persistence of fetal haemoglobin	<i>HBG</i>	-175 bp	GATA1; TAL1	Gain	(Martin et al., 1989; Wienert et al., 2015)
Haemophilia B Leyden	<i>F9</i>	-20 bp; 10 bp; -6 bp	HNF4a; C/EBPa; OC1/OC2	Loss	(Reijnen et al., 1992; Crossley and Brownlee, 1990; Funnell et al., 2013)
Haemophilia B Brandenburg	<i>F9</i>	-26 bp	AR	Loss	(Crossley et al., 1992)
Delta-thalassemia	<i>HBD</i>	-77 bp	GATA1	Loss	(Matsuda et al., 1992)
Duffy blood antigen/chemokine receptor expression	<i>DARC</i>	-46 bp	GATA1	Loss	(Tournamille et al., 1995)
Familial combined hyperlipidemia	<i>LPL</i>	-39 bp	OCT1	Loss	(Yang et al., 1995)
Bernard-Soulier syndrome	<i>GP1BB</i>	-133 bp	GATA1	Loss	(Ludlow et al., 1996)
Osteoporosis	<i>COL1A1</i>	+2 kb	Sp1	Gain	(Grant et al., 1996)
Maturity-onset diabetes of the young	<i>HNF1A</i>	-58 bp	HNF4A	Loss	(Gagnoli et al., 1997)
Asthma	<i>IL10</i>	-509 bp	YY1	Gain	(Hobbs et al., 1998)
Pyruvate kinase deficiency	<i>PKLR</i>	-72 bp	GATA1	Loss	(Manco et al., 2000)
Congenital erythropoietic porphyria	<i>UROS</i>	-70 bp; -90 bp	GATA1; CP2	Loss	(Solis et al., 2001)
Psoriasis	<i>SLC9A3R1</i>	-237 bp	RUNX1	Loss	(Helms et al., 2003)
Systemic lupus erythematosus	<i>FASLG</i>	-844 bp	CEBPB	Loss	(Wu et al., 2003)
Esophageal cancer	<i>COX-2</i>	-1195 bp	c-MYB	Gain	(Zhang et al., 2005)
Treacher Collins syndrome	<i>TCOF1</i>	-346 bp	YY1	Loss	(Masotti et al., 2005)
Alpha-thalassemia	<i>HBA</i>	-13 bp	GATA1	Gain	(De Gobbi et al., 2006)
Holoprosencephaly	<i>SHH</i>	-460 kb	SIX3	Loss	(Jeong et al., 2008)
Various cancers	<i>TERT</i>	-187 bp	ETS2	Loss	(Xu et al., 2008)
Nonsyndromic cleft lip	<i>IRF6</i>	-14 kb	AP2	Loss	(Rahimov et al., 2008)
Pierre Robin syndrome	<i>SOX9</i>	-1.44 Mb	MSX1	Loss	(Benko et al., 2009)

Bart Deplancke, Daniel Alpern, Vincent Gardeux (2016). *The Genetics of Transcription Factor DNA Binding Variation*. Cell 166/3, 538 – 554.

Základné pravidlo biológie...



Physical model of gene regulation

Reaction rate dynamics:

$$\frac{d}{dt} \underline{x_i} = \dot{x}_i = \underbrace{-\lambda_i x_i}_{\text{Degradation}} + \underbrace{m_i f_i(x)}_{\text{Synthesis rate}}$$

Product i concentration
Occupancy of active state

Monod-Wyman-Changeaux – binding in thermodynamic equilibrium:

$$f_i(x) = \frac{1}{1 + e^{-\Delta G_i(x)}} = \sigma(\Delta G_i(x))$$

Free energy difference: active vs. inactive state

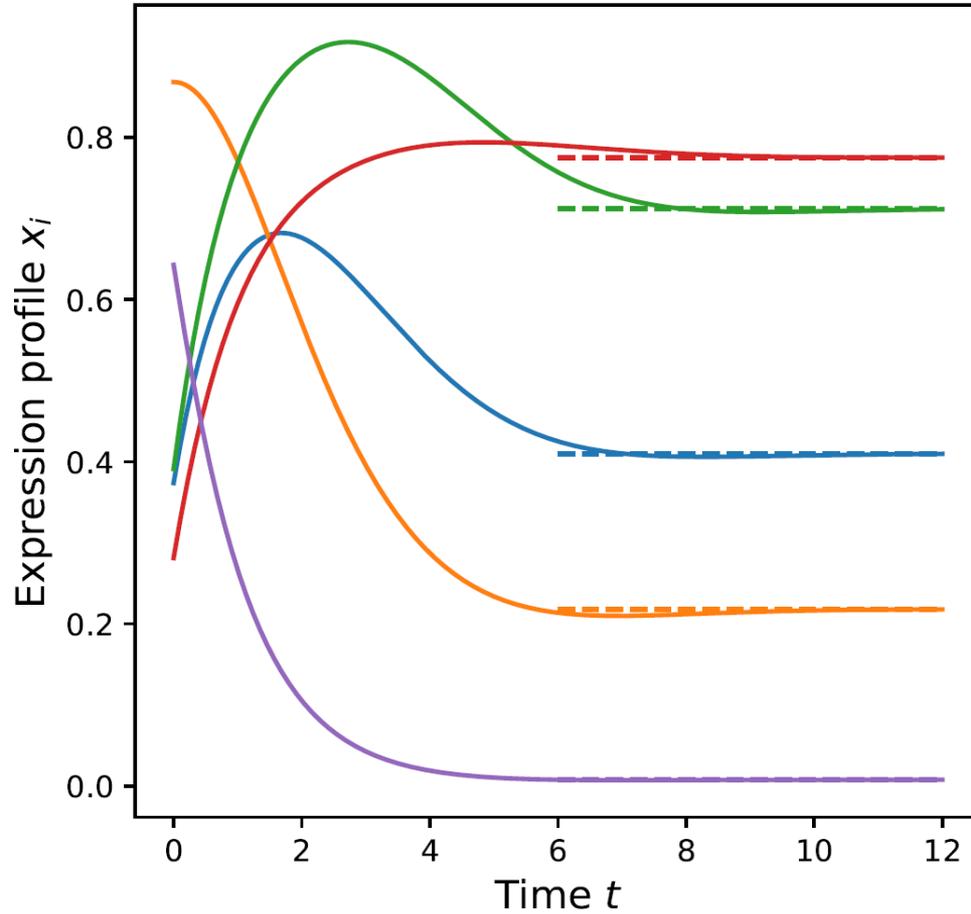
$$\Delta G_i(x) = \sum_{jk} \log \frac{1 + x_j / K_{ijk}^A}{1 + x_j / K_{ijk}^I} + \underline{\Delta G_i^0}$$

Other products
TF dissociation rate
gene i, TF j, BS k

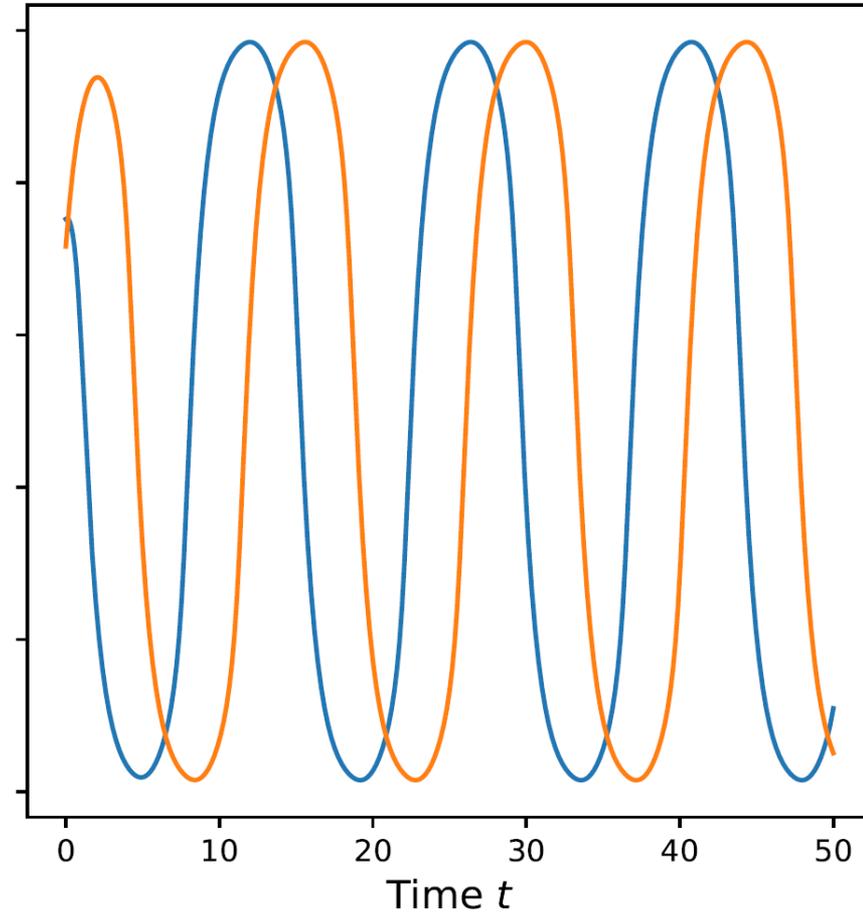
ΔG_i at zero TFs bound

Správanie genetických sietí

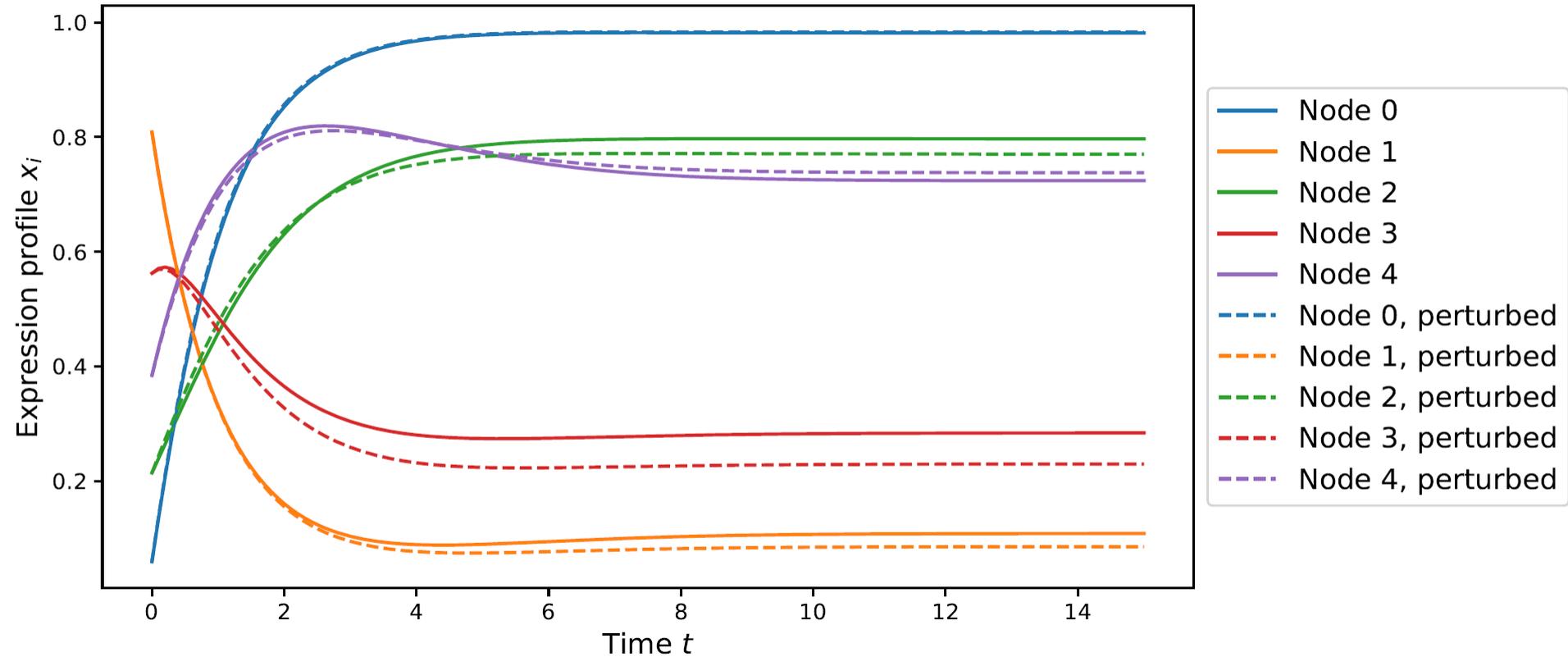
(A) 5 Genes; stable fixed point



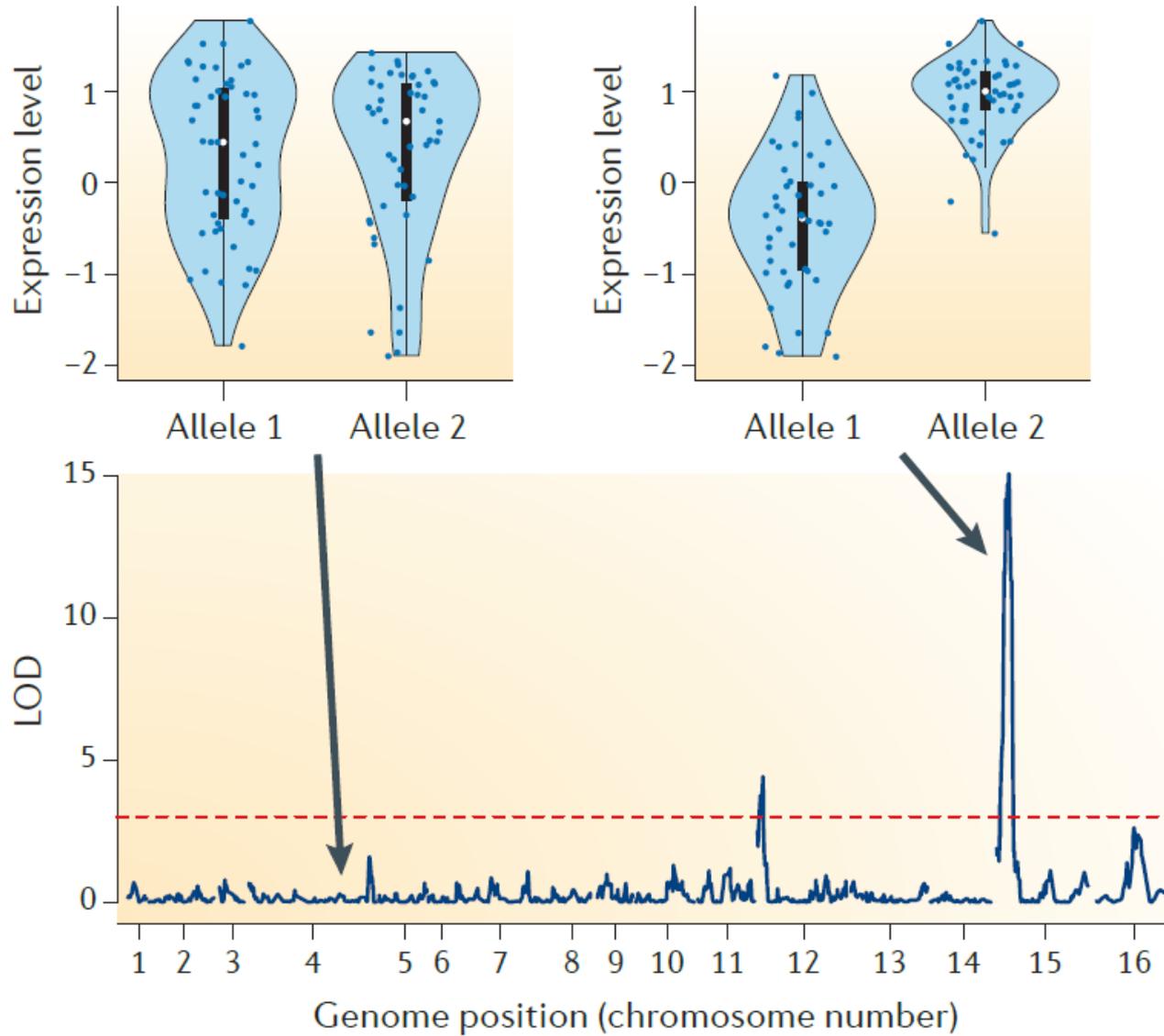
(B) 2 Genes; oscillations



Efekt mutácie

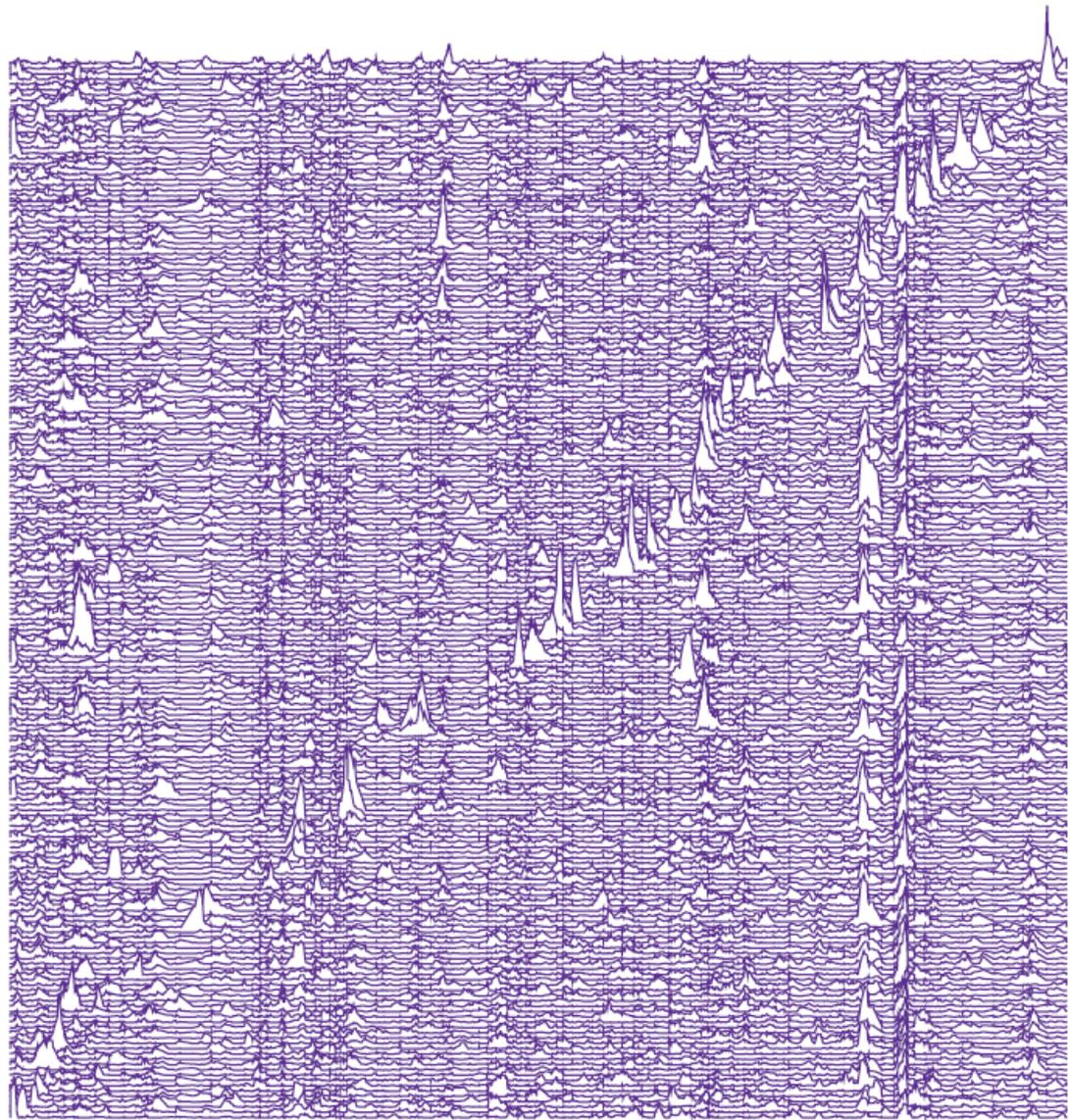


Mutácia: $R \rightarrow R + \delta R$
Expresia: $\tilde{x} \rightarrow \tilde{x} + \delta \tilde{x}(\delta R)$

a

b

Genes sorted by genomic position



1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16

Genome position (chromosome number)