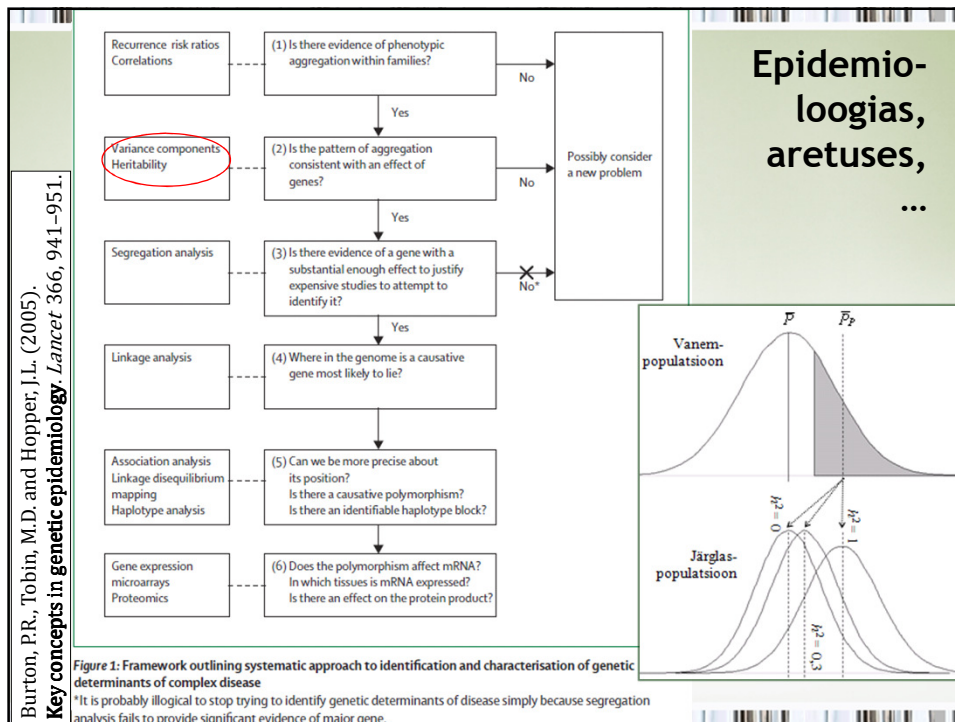


Päritavuskoefitsiendi hindamisest

Tanel Kaart

Statistilise geneetika seminar

04.02.2011



Geneetiline mudel

Fenotüüp = Genotüüp + Keskkond

$$P = \bar{P} + G + E$$

$$\text{var}(P) = \text{var}(G) + \text{var}(E) + 2 \text{cov}(G, E)$$

$$G = A + D + \dots$$

$$\text{var}(G) = \text{var}(A) + \text{var}(D) + \dots$$

G - genotüübiefekt, E - keskkonnaefekt, A - aretusväärtus, D - dominantsiefekt

$$A = \frac{1}{2}A_S + \frac{1}{2}A_D + MS$$

$$\text{var}(A) = \frac{1}{4}\text{var}(A_S) + \frac{1}{4}\text{var}(A_D) + \text{var}(MS)$$

Definitsioon

Päritavus (päritavuskoefitsient; *heritability*) on indiviididevahelise geneetilise erinevusega seletatav osa populatsiooni üldisest fenotüübilisest varieeruvusest.

$$h^2 = \text{var}(G) / \text{var}(P)$$

(**päritavus laiemas mõttes** peegeldab kogu võimalikku geneetilist mõju)

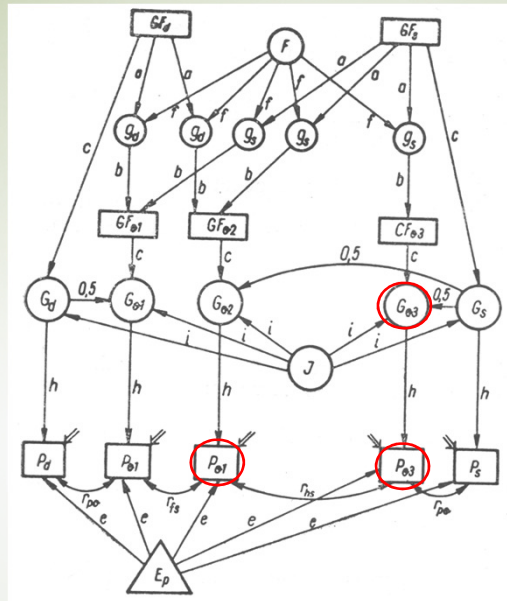
$$h^2 = \text{var}(A) / \text{var}(P)$$

(**päritavus kitsamas mõttes** mõõdab üksnes alleelide aditiivsest toimest tingitud varieeruvuse osakaalu)

$$A = h^2 \times P$$

$$(b_{P \rightarrow A} = \text{cov}(A, P) / \text{var}(P) = \text{var}(A) / \text{var}(P) = h^2)$$

Tähistus



Tähistused h ja h^2 tulenevad Sewall Wright'i töödest.

$$r(P, G) = h$$

(korrelatsioon fenotüübi- ja genotüübiväärtuste vahel)

$$r(P_{o1}, P_{o3}) = r_{hs} = 0,25 \times h^2$$

(korrelatsioon poolõvede fenotüübiväärtuste vahel)

Kasutusel on ka H^2 ja h_w^2 (päritavus laiemas mõttes)

Omadused

- ▣ Et $0 \leq \text{var}(A) \leq \text{var}(G) \leq \text{var}(P)$, siis $0 \leq h^2 \leq h_w^2 \leq 1$.
- ▣ Päritavuskoeffitsient näitab tunnuse geneetilise muutlikkuse osa antud geneetilise struktuuriga populatsioonis konkreetsetes keskkonnatingimustes.
- ▣ Päritavuskoeffitsient ei näita tunnuse päriliku tingituse määra ega mehhanismi üksik-indiviidide arengus.



▣ ...



Eeldused

- Matemaatilised/epidemioloogilised eeldused, (kasutatava statistilise mudeli vastavus andmetele):
 - kõik kasutatavad andmed on usaldusväärsed ja nende struktuur vastavab sobitavale mudelile;
 - vaatlused on sõltumatud (tinglikustatult mudeli liikmete suhtes);
 - uuritavate tunnuste ja hinnatavate efektide jaotused on korrektsed.
- Bioloogilised (geneetilised) eeldused:
 - kehtib Hardy-Weinbergi seadus;
 - rakendatav geneetiline mudel on korrektne (keskkond ja genotüüp on sõltumatud jmt);
 - registreeritud ja kasutatavad sugulussidemed on täielikud ja korrektsed;
 - ...

Keskkonna ja genotüübi sõltumatus

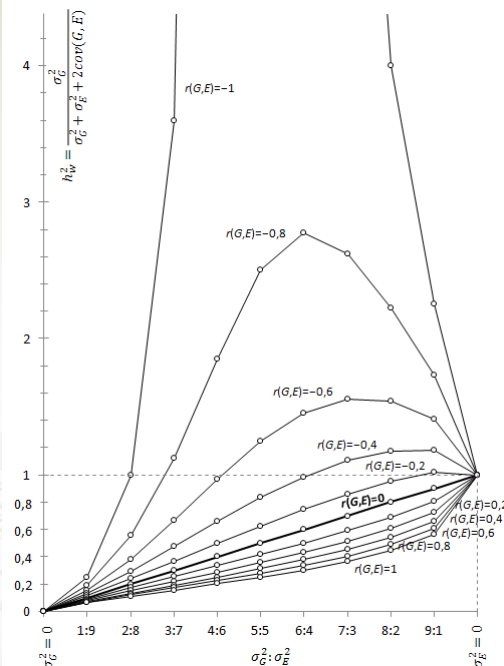
Oletame lihtsuse mõttes, et $\text{var}(G) = \text{var}(E) = \sigma^2$.

Siis keskkonna ja genotüübi sõltumatuse korral

$$\begin{aligned} \text{cov}(G, E) &= 0, \\ \text{var}(P) &= \text{var}(G) + \text{var}(E) \\ &\quad + 2 \text{cov}(G, E) = 2\sigma^2 \\ \text{ja } h_w^2 &= 1/2. \end{aligned}$$

Keskkonna ja genotüübi sõltuvuse korral aga

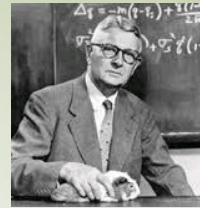
$$\begin{aligned} h_w^2 &= \sigma^2 / 2[\sigma^2 + \text{cov}(G, E)] \\ &= 1 / [2(1 + r_{GE})], \\ \text{sest } \text{cov}(G, E) &= r_{GE} \sigma^2. \end{aligned}$$



Hindamismeetodid

Kaks peamist koolkonda

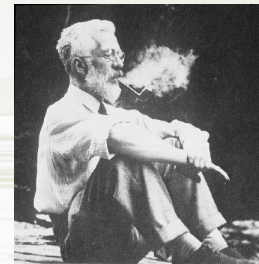
- Regressioon- ja korrelatsioonanalüüs (Sewall Wright, Jay Laurence Lush; rajakoeffitsientide meetod, seleksiooniindeksid)



Idee: sugulased on omavahel sarnasemad!

$$h^2 = b_{X \rightarrow Y} / 2f_{XY} \text{ või } h^2 = r_{XY} / 2f_{XY}$$

- Dispersioonanalüüs ja segamudelid (Sir Ronald Aylmer Fisher; dispersioonikomponendid, intraklasskorrelatsioonikordaja)



Idee: sugulased erinevad üksteisest vähem!

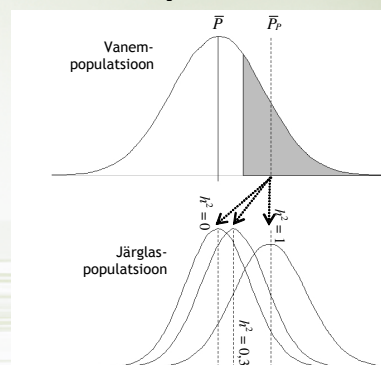
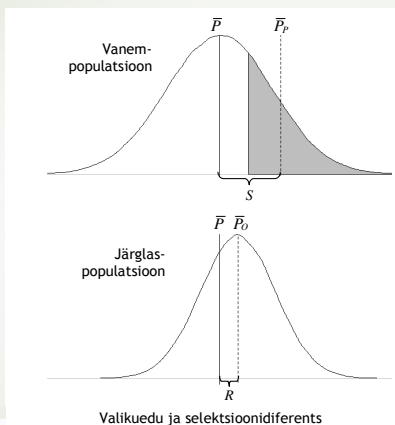
$$h^2 = t / 2f_{XY}, \quad t = \sigma_B^2 / (\sigma_B^2 + \sigma_W^2)$$

Seleksiooniekspereimendid

Seleksiooniefekti (R) ja seleksioonidiferentsi (S) suhe annab realiseerunud e efektiivse päritavuse:

$$h^2 = \frac{\bar{P}_O - \bar{P}}{\bar{P}_P - \bar{P}} = \frac{R}{S}$$

$$R = h^2 \times S$$



Statistiline mudel vs geneetiline mudel

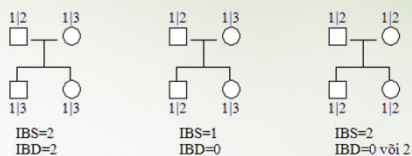
Statistiline mudel (<i>observational components</i>)	Geneetiline mudel (<i>causal components</i>)
$b_{X \rightarrow Y}; r_{XY}; \sigma^2; \dots$	$\text{var}(A); \text{var}(G); \text{var}(E_C); \dots$
↓	↓
$\widehat{h}^2; \dots$	$h^2; \dots$

Vaja, et mõista hinnatud päritavuskoefitsiendi olemust, ning et hinnata/kahtlustada/...

- dominantsiefekti,
- püsivat keskkonnamõju,
- emapoolset geneetilist efekti,
- imprintingut,
- epigeneetilist päritavust,
- ...

Indiviidide vaheline sugulus

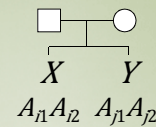
Indiviidide vaheline **suguluskoefitsient** (*coancestry, kinship coefficient, ...*) on tõenäosus, et ühe indiviidi juhuslikult valitud lookusest juhuslikult valitud alleel on **päritolult identne (IBD)** teise indiviidi vastavast lookusest juhuslikult valitud alleeliga.



Olgu indiviidid X ja Y genotüüpidega $A_{11}A_{12}$ ning $A_{21}A_{22}$.

$$\begin{aligned}
 f_{XY} &= P(2 \text{ juhuslikult valitud alleeli on } IBD) \\
 &= \sum_{i,j} P(A_i \equiv A_j \mid \text{juhuslikult valitud alleelid on } A_i \text{ ja } A_j) \\
 &\quad \times P(\text{juhuslikult valiti alleelid } A_i \text{ ja } A_j) \\
 &= \frac{1}{4} [P(A_{11} \equiv A_{11}) + P(A_{11} \equiv A_{22}) + P(A_{22} \equiv A_{11}) + P(A_{22} \equiv A_{22})]
 \end{aligned}$$

Indiviidide vaheline sugulus



Olgu X ja Y täisõved (FS). Siis

$$f_{XY} = P(2 \text{ juhuslikult valitud alleeli on IBD})$$

$$= \frac{1}{4} [\underbrace{P(A_{11} \equiv A_{11})}_{\frac{1}{2} \times \frac{1}{2}} + \underbrace{P(A_{11} \equiv A_{22})}_{\frac{1}{2} \times \frac{1}{2}} + \underbrace{P(A_{22} \equiv A_{11})}_{\frac{1}{2} \times \frac{1}{2}} + \underbrace{P(A_{22} \equiv A_{22})}_{\frac{1}{2} \times \frac{1}{2}}] = \frac{1}{4}$$

Sugulusaste	Suguluskoeffitsient f (või ϕ , $\frac{1}{2}r$, ...)
Indiviid ise	$\frac{1}{2}$
Ühemunarakukaksikud (MZ)	$\frac{1}{2}$
Täisõved (FS)	$\frac{1}{4}$
Poolõved (HS)	$\frac{1}{8}$
Vanem ja järglane (PO)	$\frac{1}{4}$
Vanavanem ja lapselaps	$\frac{1}{8}$

Indiviidide vaheline geneetiline kovariatsioon

Olgu indiviidid X ja Y genotüüpidega $A_{11}A_{22}$ ning $A_{11}A_{22}$.

Tähistagu a_{11} , a_{22} , a_{11} ja a_{22} alleeli- ning $d_{11,22}$ ja $d_{11,22}$ dominantsiefekte.

$$G_X = a_{11} + a_{22} + d_{11,22} \text{ ja } G_Y = a_{11} + a_{22} + d_{11,22}$$

$$\begin{aligned} \text{cov}(G_X, G_Y) &= \text{cov}(a_{11}, a_{11}) + \text{cov}(a_{11}, a_{22}) + \text{cov}(a_{11}, d_{11,22}) \\ &\quad + \text{cov}(a_{22}, a_{11}) + \text{cov}(a_{22}, a_{22}) + \text{cov}(a_{22}, d_{11,22}) \\ &\quad + \text{cov}(d_{11,22}, a_{11}) + \text{cov}(d_{11,22}, a_{22}) + \text{cov}(d_{11,22}, d_{11,22}). \end{aligned}$$

$$\text{cov}(a, d) = 0;$$

$$\text{cov}(a_x, a_y) = \begin{cases} 0, & \text{kui } x \neq y \text{ } (\neg IBD) \\ \text{var}(A)/2, & \text{kui } x \equiv y \text{ } (IBD) \end{cases}$$

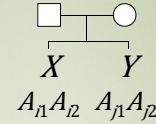
sest $\text{var}(A) = \text{var}(a_1 + a_2) = 2\text{var}(a_1)$, mistõttu $\text{var}(a_1) = \text{cov}(a_1, a_1) = \text{var}(A)/2$

$$\text{cov}(d_{xy}, d_{wz}) = \begin{cases} 0, & \text{kui } xy \neq wz \text{ (mõlemad alleelid } \neg IBD) \\ \text{var}(D), & \text{kui } xy \equiv wz \text{ (mõlemad alleelid } IBD) \end{cases}$$

Indiviidide vaheline geneetiline kovariatsioon

Olgu X ja Y genotüübiväärtustega

$$G_X = a_{11} + a_{12} + d_{112} \text{ ja } G_Y = a_{j1} + a_{j2} + d_{j1j2}$$



täisõved (FS).

Siis on võimalikud kolm varianti:

Variant	Tõenäosus	$cov(G_X G_Y)$
#IBD = 0	$\frac{1}{4}$	0
#IBD = 1	$\frac{1}{2}$	$\text{var}(A) / 2$
#IBD = 2	$\frac{1}{4}$	$\text{var}(A) + \text{var}(D)$

$$\begin{aligned} \text{cov}(G_X G_Y) &= \frac{1}{2} \times \text{var}(A) / 2 + \frac{1}{4} \times [\text{var}(A) + \text{var}(D)] \\ &= \frac{1}{2} \times \text{var}(A) + \frac{1}{4} \times \text{var}(D) \end{aligned}$$

Indiviidide vaheline geneetiline kovariatsioon

Üldine valem:

$$\text{cov}(G_X G_Y) = 2f_{XY} \text{var}(A) + g_{XY} \text{var}(D)$$

Sugulusaste	f_{XY} (või ϕ_{XY} ...)	g_{XY} (või Δ_{XY} ...)	$\text{cov}(G_X G_Y)$
Indiviid ise	$\frac{1}{2}$	1	$\text{var}(A) + \text{var}(D)$
Ühem. kaksikud	$\frac{1}{2}$	1	$\text{var}(A) + \text{var}(D)$
Täisõved	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2} \text{var}(A) + \frac{1}{4} \text{var}(D)$
Poolõved	$\frac{1}{8}$	0	$\frac{1}{4} \text{var}(A)$
Vanem ja järglane	$\frac{1}{4}$	0	$\frac{1}{2} \text{var}(A)$
V.vanem ja l.laps	$\frac{1}{8}$	0	$\frac{1}{4} \text{var}(A)$
Abikaasad	0	0	0

Vanemate-järglaste vaheline regressioon

Järglane ja üks vanem:

$$b_{P \rightarrow O} = \frac{\text{cov}(P_P, P_O)}{\text{var}(P_P)} = \frac{1/2 \text{var}(A)}{\text{var}(P)} = 1/2 h^2 \Rightarrow h^2 = 2 \times b_{P \rightarrow O}$$

$$r_{PO} = \frac{\text{cov}(P_P, P_O)}{\sqrt{\text{var}(P_P) \times \text{var}(P_O)}} = \frac{1/2 \text{var}(A)}{\text{var}(P)} = 1/2 h^2, \text{ kui } \text{var}(P_P) = \text{var}(P_O)$$

Järglane ja vanemate keskmise:

$$b_{\bar{P} \rightarrow O} = \frac{\text{cov}(P_{\bar{P}}, P_O)}{\text{var}(P_{\bar{P}})} = \frac{1/2 \text{var}(A)}{1/2 \text{var}(P)} = h^2 \Rightarrow h^2 = b_{\bar{P} \rightarrow O}, \text{ kui } \text{var}(P_{P1}) = \text{var}(P_{P2})$$

$$r_{\bar{P}O} = \frac{\text{cov}(P_{\bar{P}}, P_O)}{\sqrt{\text{var}(P_{\bar{P}}) \times \text{var}(P_O)}} = \frac{1/2 \text{var}(A)}{\sqrt{1/2 \text{var}(P)}} = \sqrt{1/2} h^2, \text{ kui ...}$$

Järglaste keskmise ja üks vanem:

$$b_{P \rightarrow \bar{O}} = \frac{\text{cov}(P_P, P_{\bar{O}})}{\text{var}(P_P)} = \frac{1/2 \text{var}(A)}{\text{var}(P)} = 1/2 h^2 \Rightarrow h^2 = 2 \times b_{P \rightarrow \bar{O}}$$

$$r_{P\bar{O}} = \frac{\text{cov}(P_P, P_{\bar{O}})}{\sqrt{\text{var}(P_P) \times \text{var}(P_{\bar{O}})}} = \frac{1/2 \text{var}(A)}{\sqrt{1/n \text{var}(P)}}, \text{ kui ...}$$

Õvede vaheline korrelatsioon

Poolõved:

$$r_{HS} = \frac{\text{cov}(HS)}{\sqrt{\text{var}(P) \times \text{var}(P)}} = \frac{1/4 \text{var}(A)}{\text{var}(P)} = 1/4 h^2 \Rightarrow h^2 = 4 \times r_{HS}$$

Täisõved (=kahemunarakukaksikud):

$$r_{FS} (= r_{DZ}) = \frac{\text{cov}(FS)}{\sqrt{\text{var}(P) \times \text{var}(P)}} = \frac{1/2 \text{var}(A) + 1/4 \text{var}(D) + \text{var}(E_C)}{\text{var}(P)} \geq 1/2 h^2$$

Ühemunarakukaksikud:

$$r_{MZ} = \frac{\text{cov}(MZ)}{\sqrt{\text{var}(P) \times \text{var}(P)}} = \frac{\text{var}(G) + \text{var}(E_C)}{\text{var}(P)} \geq h_w^2$$

Kaksikute uuringuis:

$$h_w^2 = \frac{\text{var}(G)}{\text{var}(P)} \approx 2(r_{MZ} - r_{DZ})$$

$$c^2 = \frac{\text{var}(E_C)}{\text{var}(P)} \approx 2r_{DZ} - r_{MZ}$$

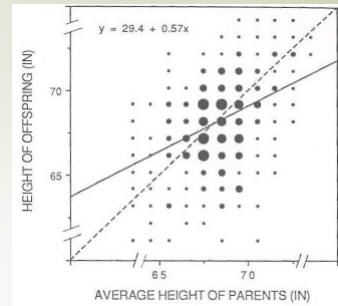
Vanemate-järglaste vaheline regressioon

Sir Francis Galton'i (1889) uuring esitamaks järglaste pikkust (928 indiviidi) vanemate keskmise pikkuse kaudu (205 vanempaari).

Wikipedia

Regressioonikordaja on alati < 1 ,
 st et järglaste fenotüüp on keskmiselt populatsiooni keskmisele lähemal (võrreldes vanemate fenotüübiga).

Regression – tagasilangus, regressioon

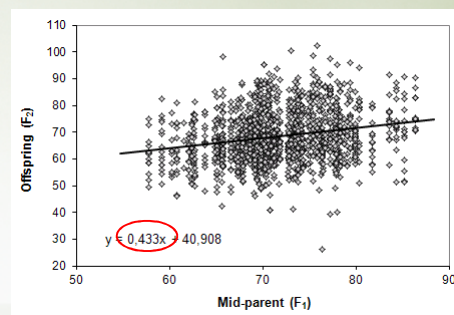


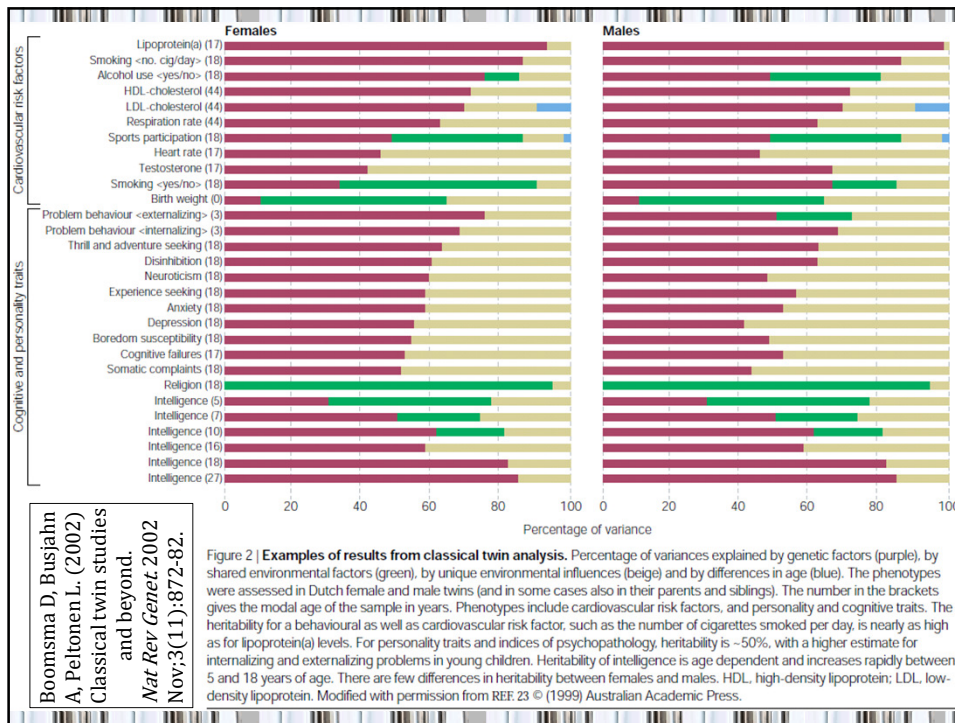
Vanemate-järglaste vaheline regressioon

Näide. Uurimisobjekt: võsavaksik; uuritav tunnus: nuku mass.

$F_1 \rightarrow F_2$	h^2
<u>$(S+D)/2 \rightarrow O$</u>	<u>0,433</u>
$S \rightarrow O$	0,581
$D \rightarrow O$	0,449
$(S+D)/2 \rightarrow O_m$	0,322
$S \rightarrow O_m$	0,472
$D \rightarrow O_m$	0,346
$(S+D)/2 \rightarrow O_f$	0,569
$S \rightarrow O_f$	0,721
$D \rightarrow O_f$	0,559

S - sire (isa), D - dam (ema),
 O_m - male offspring (isane järglane),
 O_f - female offspring (emane järglane)





Poolõvede dispersioonanalüüs

$$y_{ij} = \mathbf{x}'_{ij}\boldsymbol{\beta} + s_i + e_{ij}, \quad \sigma_y^2 = \sigma_s^2 + \sigma_e^2,$$

Et ühelt poolt

$$\text{cov}(y_{ij}; y_{ij'}) = \text{cov}(HS) = \frac{1}{4} \text{var}(A)$$

ja teiselt poolt

$$\text{cov}(y_{ij}; y_{ij'}) = \text{cov}(s_i + e_{ij}; s_i + e_{ij'}) = \text{cov}(s_i; s_i) = \sigma_s^2$$

siis järelikult

$$\sigma_s^2 = \frac{1}{4} \text{var}(A),$$

$$\sigma_e^2 = \frac{3}{4} \text{var}(A) + \text{var}(D) + \dots + \text{var}(E)$$

ja

$$h^2 = \frac{4\sigma_s^2}{(\sigma_s^2 + \sigma_e^2)}$$

$$P_{ij} = \bar{P} + \underbrace{\frac{1}{2} A_{S_i}}_{s_i} + \underbrace{\frac{1}{2} A_{D_{ij}} + MS_{ij} + D_{ij} + \dots + E_{ij}}_{e_{ij}}$$

$$\text{var}(P) = \underbrace{\frac{1}{4} \text{var}(A)}_{\sigma_s^2} + \underbrace{\frac{1}{4} \text{var}(A) + \frac{1}{2} \text{var}(A) + \text{var}(D) + \dots + \text{var}(E)}_{\sigma_e^2}$$

Täisõvede dispersioonanalüüs

$$y_{ijk} = \mathbf{x}'_{ijk} \boldsymbol{\beta} + s_i + d_{ij} + e_{ijk},$$

$$\sigma_y^2 = \sigma_s^2 + \sigma_d^2 + \sigma_e^2$$

$$\sigma_s^2 = \text{cov}(\text{HS}) = \frac{1}{4} \text{var}(A)$$

$$\sigma_e^2 = \sigma_y^2 - \underbrace{\text{cov}(\text{FS})}_{\frac{1}{2} \text{var}(A) + \frac{1}{4} \text{var}(D) + \text{var}(E_c)} = \frac{1}{2} \text{var}(A) + \frac{3}{4} \text{var}(D) + \text{var}(E)$$

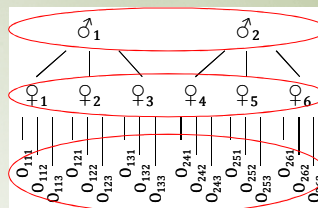
$$\begin{aligned} \sigma_d^2 &= \sigma_y^2 - \sigma_s^2 - \sigma_e^2 \\ &= \text{cov}(\text{FS}) - \text{cov}(\text{HS}) = \frac{1}{4} \text{var}(A) + \frac{1}{4} \text{var}(D) + \text{var}(E_c) \end{aligned}$$

Seega

$$h^2 = 4\sigma_s^2 / (\sigma_s^2 + \sigma_d^2 + \sigma_e^2)$$

$$h^2 \leq 2(\sigma_s^2 + \sigma_d^2) / (\sigma_s^2 + \sigma_d^2 + \sigma_e^2)$$

$$c^2 = (\sigma_d^2 - \sigma_s^2) / \sigma_y^2, \text{ kui } \text{var}(D) = 0; d^2 = 4(\sigma_d^2 - \sigma_s^2) / \sigma_y^2, \text{ kui } \text{var}(E_c) = 0; \dots$$



Täisõvede dispersioonanalüüs

Näide. Uurimisobjekt: võsavaksik; uuritav tunnus: nuku mass.

Covariance Parameter Estimates				
Cov Parm	Estimate	Standard Error	Z Value	Pr > Z
isaF1	5.1016	3.1779	1.61	0.0542
emaF1(isaF1)	10.8303	3.0598	3.54	0.0002
Residual	43.3495	1.7027	25.46	<.0001

$$\begin{aligned} h^2 &= 4\sigma_s^2 / (\sigma_s^2 + \sigma_d^2 + \sigma_e^2) \\ &= 4 \times 5,10 / (5,10 + 10,83 + 43,35) = 0,344 \end{aligned}$$

$$\begin{aligned} d^2 &= 4(\sigma_d^2 - \sigma_s^2) / \sigma_y^2 \\ &= 4 \times (10,83 - 5,10) / (5,10 + 10,83 + 43,35) \\ &= 0,387 \end{aligned}$$



Mäletatavasti:

F ₁ → F ₂	h ²
(S+D)/2 → O	0,433
S → O	0,581
D → O	0,449
(S+D)/2 → O _m	0,322
S → O _m	0,472
D → O _m	0,346
(S+D)/2 → O _f	0,569
S → O _f	0,721
D → O _f	0,559

S - sire (isa), D - dam (ema),
O_m - male offspring (isane järglane),
O_f - female offspring (emane järglane)

Looma mudel

Geneetiline mudel

$$P_i = \bar{P} + A_i + E_i$$

Statistiline mudel

$$y_i = \mathbf{x}'_i \boldsymbol{\beta} + a_i + e_i$$

$$\sigma_y^2 = \sigma_a^2 + \sigma_e^2$$

$$\text{var} \begin{pmatrix} \mathbf{a} \\ \mathbf{e} \end{pmatrix} = \begin{pmatrix} \mathbf{A}\sigma_a^2 & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_N\sigma_e^2 \end{pmatrix}$$

Näide: (1) ---- [2] ---- (3)
 | |
 [4] ---- (5)
 |
 (6)

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & 0.5 & 0 & 0.25 \\ 0 & 1 & 0 & 0.5 & 0.5 & 0.5 \\ 0 & 0 & 1 & 0 & 0.5 & 0.25 \\ 0.5 & 0.5 & 0 & 1 & 0.25 & 0.625 \\ 0 & 0.5 & 0.5 & 0.25 & 1 & 0.625 \\ 0.25 & 0.5 & 0.25 & 0.625 & 0.625 & 1.125 \end{pmatrix}$$

Maatriks \mathbf{A}

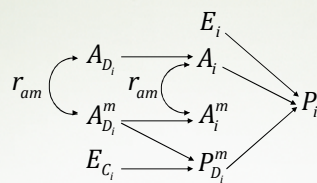
(aditiivgeneetilise suguluse maatriks)

- omab üht rida ja üht veergu iga sugupuus sisalduva indiviidi tarvis;
- elemendid on aditiivgeneetilise suguluse kordajad ($2f_{XY}$) + inbriiding

$$h^2 = \sigma_a^2 / \sigma_y^2$$

Maternaalse efektiga looma mudel

Geneetiline mudel



Päritavus: $h^2 = \sigma_a^2 / \sigma_y^2$

Maternaalne päritavus: $m^2 = \sigma_m^2 / \sigma_y^2$

Püsiva keskkonnamõju osakaal: $c^2 = \sigma_c^2 / \sigma_y^2$

Statistiline mudel

$$y_{ijk} = \mathbf{x}'_{ijk} \boldsymbol{\beta} + a_i + m_j + c_k + e_{ijk}$$

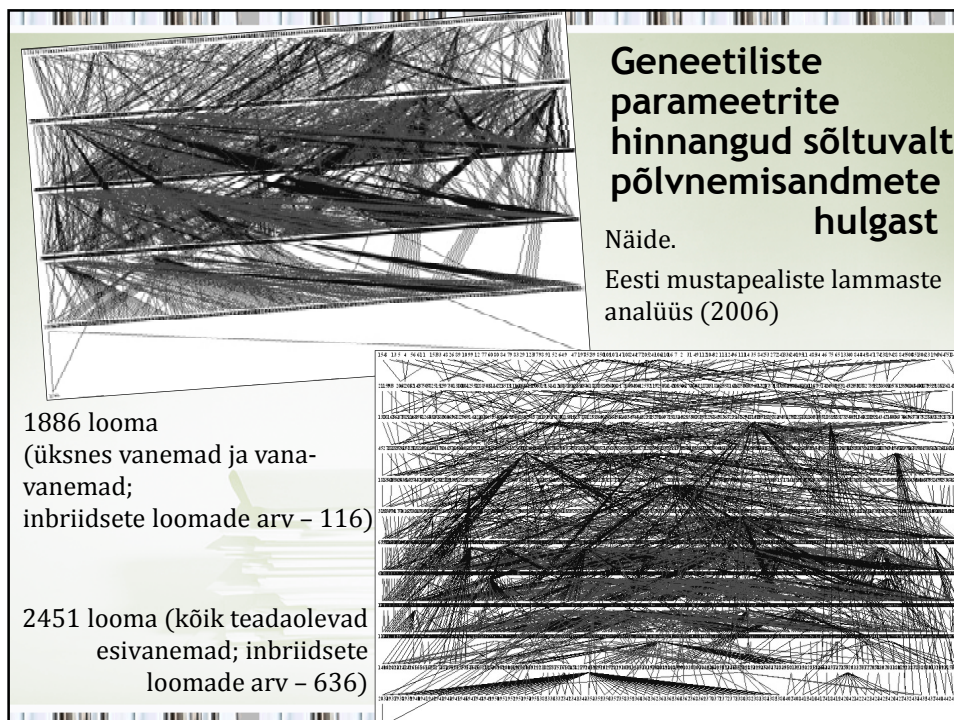
$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\mathbf{a} + \mathbf{W}\mathbf{m} + \mathbf{S}\mathbf{c} + \mathbf{e}$$

$$\sigma_y^2 = \sigma_a^2 + \sigma_m^2 + 2\sigma_{am} + \sigma_c^2 + \sigma_e^2$$

$$\text{var} \begin{pmatrix} \mathbf{a} \\ \mathbf{m} \\ \mathbf{c} \\ \mathbf{e} \end{pmatrix} = \begin{pmatrix} \mathbf{A}\sigma_a^2 & \mathbf{A}\sigma_{am} & \mathbf{0} & \mathbf{0} \\ \mathbf{A}\sigma_{am} & \mathbf{A}\sigma_m^2 & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I}_b\sigma_c^2 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{I}_N\sigma_e^2 \end{pmatrix}$$

Otsese ja maternaalse geneetilise efekti vaheline korrelatsioon:

$$r_{am} = \sigma_{am} / \sqrt{\sigma_a^2 \cdot \sigma_m^2}$$

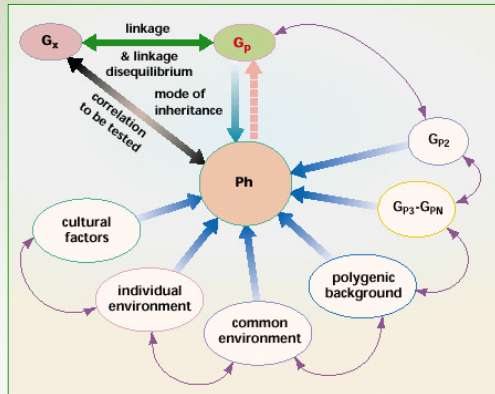


Geneetiliste parameetrite hinnangud sõltuvalt mudelist ja põlvnemisandmete hulgast

Eesti mustapealiste lammaste võõrutusmass. Mudel: $y_{ijk} = \mathbf{x}'\boldsymbol{\beta} + a_i + m_j + c_k + e_{ijk}$

	Põlvnemis- andmed	Hinnatud geneetilised parameetrid				
		$\hat{\sigma}_e^2$	\hat{h}^2	\hat{m}^2	\hat{r}_{am}	\hat{c}^2
Mudel 1	isad	48,67	0,256	-	-	-
Mudel 2	isad+emad	44,82	0,274	-	-	-
Mudel 3	≤ 2	43,03	0,162	-	-	-
Mudel 4	eellasgen.	41,67	0,118	-	-	0,065
Mudel 5		41,74	0,139	0,043	0,446	-
Mudel 6		41,30	0,127	0,012	0,922	0,048
Mudel 7	≤ 12	43,00	0,165	-	-	-
Mudel 8	eellasgen.	41,62	0,129	-	-	0,066
Mudel 9		41,97	0,125	0,055	0,087	-
Mudel 10		41,66	0,118	0,024	0,260	0,041

QTL-efektiga looma mudel



Schematic model of trait aetiology - Weiss and Terwilliger (2000) *Nature genet.* **26**: 151-157

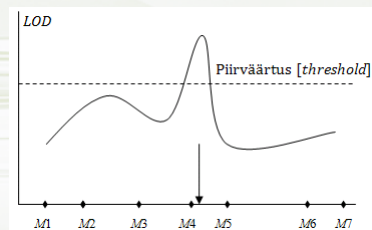
$$h_{QTL}^2 = \frac{\sigma_a^2 + \sigma_q^2}{\sigma_a^2 + \sigma_q^2 + \sigma_e^2}$$

$$P_i = \bar{P} + Q_i^p + Q_i^m + A_i + E_i$$

$$y = X\beta + Za + ZTq + e$$

$$\sigma_y^2 = \sigma_a^2 + \sigma_q^2 + \sigma_e^2$$

$$\text{var} \begin{pmatrix} \mathbf{a} \\ \mathbf{q} \\ \mathbf{e} \end{pmatrix} = \begin{pmatrix} \mathbf{A}\sigma_a^2 & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{Q}_{jM,p}\sigma_q^2 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I}_n\sigma_e^2 \end{pmatrix}$$

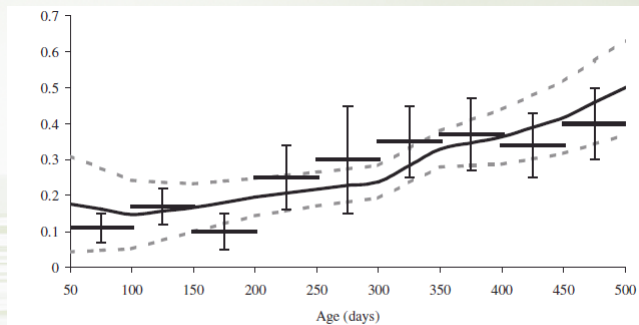


Ajas/ruumis/... muutuv päritavus

$$P_i(t) = \bar{P} + A_i(t) + E_i(t)$$

$$y_{it} = \mathbf{x}_i' \boldsymbol{\beta} + a_{it} + e_{it}$$

$$\sigma_y^2(t) = \sigma_a^2(t) + \sigma_e^2(t)$$



Fischer TM, Gilmour AR, van der Werf JH. Computing approximate standard errors for genetic parameters derived from random regression models fitted by average information REML. *Genet Sel Evol.* 2004 May-Jun; 36(3):363-9.

Figure 2. Heritability estimates of weight over time \pm standard errors from random regression (continuous line) and univariate (discrete lines) analysis.

Ajas/ruumis/... muutuv päritavus

$$\begin{aligned}
 a_t &= a_0 + a_1 \times t, & \text{cov}(a_t, a_s) \\
 &= \text{cov}(a_0 + a_1 t, a_0 + a_1 s) \\
 &= \text{var}(a_0) + t \text{cov}(a_1, a_0) + s \text{cov}(a_0, a_1) + st \text{var}(a_1) \\
 &= \begin{pmatrix} 1 & t \\ 1 & s \end{pmatrix} \text{var} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ t & s \end{pmatrix} = \Phi \mathbf{K} \Phi^T
 \end{aligned}$$

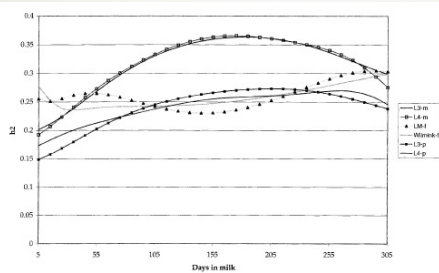


Figure 1. Heritability (h^2) for milk (m), fat (f), and protein (p) yield as a function of DIM. Third (L3) and fourth (L4) order Legendre polynomials and a function proposed by Lidauer and Mantysaari (1999) (LM) functions were fitted in an animal model for milk and protein, and the LM and the Wilmink function were fitted for fat (Modified from Jakobsen, 2000).

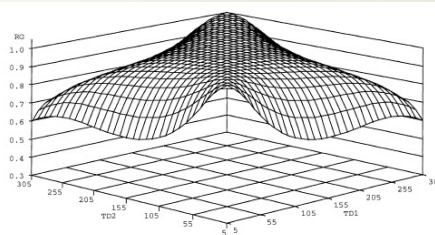
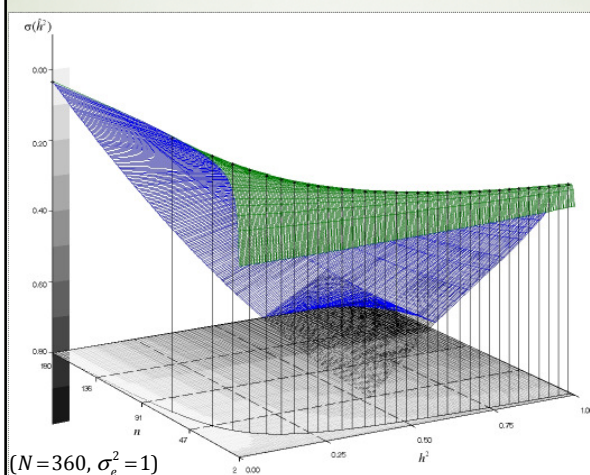


Figure 2. Genetic correlations (r_g) between test-day milk yields on one test day (TD_1) and another test day (TD_2) using a 4 order Legendre polynomial on 8075 Holstein cows (Modified from Jakobsen, 2000).

Jensen J. Genetic evaluation of dairy cattle using test-day models. *J Dairy Sci.* 2001 Dec; 84(12):2803-12.

Optimaalne disain

- ▣ Regressioonanalüüs vanemate keskmise baasil ($h^2 = b_{\bar{P} \rightarrow O}$).
- ▣ Korrelatsioonanalüüs poolõvede baasil ($h^2 = 4 \times r_{HS}$)



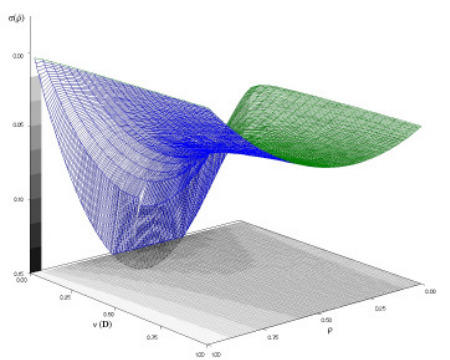
- ▣ Poolõvede dispersioonanalüüsil

$$\text{var}(\hat{h}^2) \rightarrow \min,$$

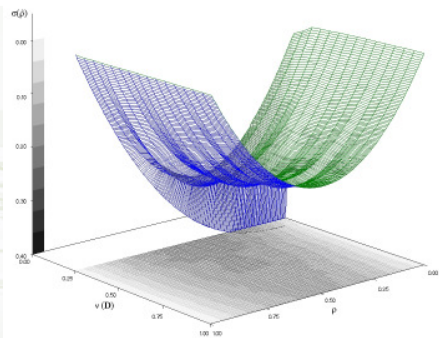
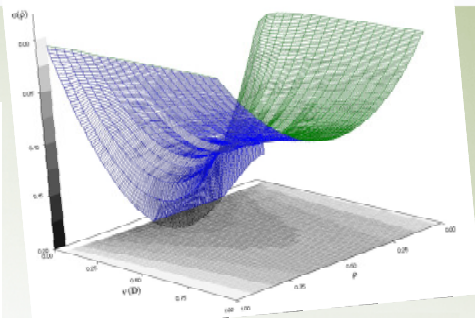
$$\text{kui } n = \frac{N(\tau+1)+1}{N\tau+2}$$

$$(\tau = \sigma_s^2 / \sigma_e^2)$$

Sõltuvus andmete tasakaalulisusest



Intraklasskorrelatsioonikordaja (ρ) (=päritavuskoeffitsiendi) hinnangute täpsus poolõvede analüüsil sõltuvalt andmete tasakaalulisusest (gruppide/isade arv $a = 90, 20, 4$; $N = 360, \sigma_e^2 = 1$)



Geneetiliste parameetrite hinnangute täpsus sõltuvalt sugulussidemete tugevusest

$\text{Var}(\hat{\sigma}_{a,REML}^2) \times 10$		
$h^2 = 0.8$	$h^2 = 0.25$	$h^2 = 0.01$
5.0505	0.3591	0.2061
4.4824	0.3502	0.2060
3.5541	0.3372	0.2059
4.2435	0.3542	0.2061
0.2534	0.2131	0.2027
0.2020	0.2020	0.2020
0.2038	0.1995	0.2020

Model	Description	Scheme / $\text{Var}(\mathbf{a}) = \sigma^2 \mathbf{A}$ (for symmetric matrices only upper triangular block is shown)
M1	All 100 observed individuals unrelated.	$\text{Var}(\mathbf{a}) = \sigma^2 \mathbf{I}_{100}$
M2	25 individuals are half-sibs, 75 are unrelated. Parents of half-sibs are not under study.	$\text{Var}(\mathbf{a}) = \sigma^2 \begin{pmatrix} \mathbf{G} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_{75} \end{pmatrix}$, where $\mathbf{G}_{10} = \begin{pmatrix} 1 & 0.25 & \dots & 0.25 \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{pmatrix}$
M3	25 individuals are full-sibs, 75 are unrelated. Parents of full-sibs are not under study.	$\text{Var}(\mathbf{a}) = \sigma^2 \begin{pmatrix} \mathbf{G}_f & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_{75} \end{pmatrix}$, where $\mathbf{G}_f = \begin{pmatrix} 1 & 0.5 & \dots & 0.5 \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{pmatrix}$
M4	2 families, where 2 half-sibs are crossed to produce 2 offspring (thus in both families are 5 related individuals – 1 grandparent, 2 parents and 2 offspring); the other 90 individuals are unrelated.	$\text{Var}(\mathbf{a}) = \sigma^2 \begin{pmatrix} \mathbf{G} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{G}_f & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I}_{90} \end{pmatrix}$, where $\mathbf{G} = \begin{pmatrix} 1 & 0.5 & 0.5 & 0.5 & 0.5 \\ & 1 & 0.25 & 0.625 & 0.625 \\ & & 1 & 0.625 & 0.625 \\ & & & 1.125 & 0.625 \\ & & & & 1.125 \end{pmatrix}$
M5	2 unrelated individuals crossed to produce 2 offspring who are crossed to produce 2 offspring etc.; 100 th individual has inbreeding coefficient equal to 0.99996.	$\text{Var}(\mathbf{a}) = \sigma^2 \begin{pmatrix} 1 & 0 & 0.5 & 0.5 & 0.5 & \dots & 0.5 \\ & 1 & 0.5 & 0.5 & 0.5 & \dots & 0.5 \\ & & 1 & 0.5 & 0.75 & \dots & 0.75 \\ & & & 1 & 0.75 & \dots & 0.75 \\ & & & & 1.25 & \dots & 1 \\ & & & & & \ddots & \\ & & & & & & 1.99996 \end{pmatrix}$
M6	1 offspring of unrelated individuals and her/his 99 identical clones (thus, the additive genetic relationship between all individuals equals to 1 but there is no inbreeding). Parents of the first individual are not under study, they are used just to have the correct genetic relationships.	$\text{Var}(\mathbf{a}) = \sigma^2 \mathbf{J}_{100}$
M7	1 offspring of genetically identical (pure line) individuals and her/his 99 identical clones (thus the additive genetic relationship between individuals and the inbreeding both equal to 1). Parents of the first individual are not under study, they are used just to have the correct genetic relationships.	$\text{Var}(\mathbf{a}) = 2\sigma^2 \mathbf{J}_{100}$

Päritavuskoefitsiendi väärtus sõltuvalt kasutatavast mudelist

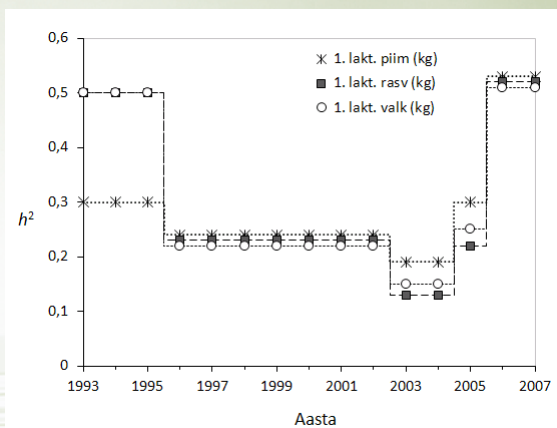
Jõudluskontrollikeskus

1993-1995 Isa mudel

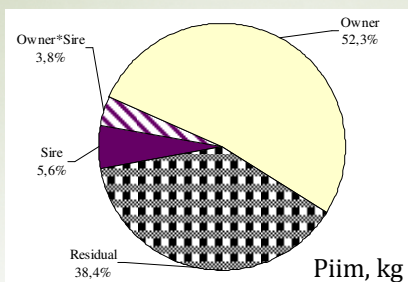
1996-2002 Looma mudel

2003-2005 Kontrollpäeva looma mudel

2006-... Juhuslike regressiooni-kordajatega kontrollpäeva looma mudel (*random regression test day animal model*)



Genotüübi-keskkonna interaktsioon



Eesti holsteini tõugu lehmade uuring 2003. aastal;
loomade arv – 13194;
isade arv – 231;
omanike arv – 1029

