



Decision support tools – from the cradle to the grave

Donagh Berry¹, Margaret Kelleher², & Fíona Dunne¹

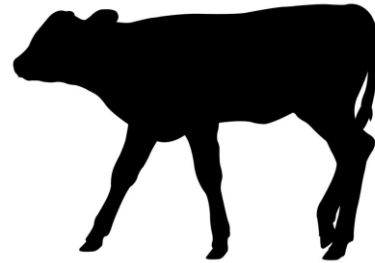
¹Teagasc, Moorepark, Ireland

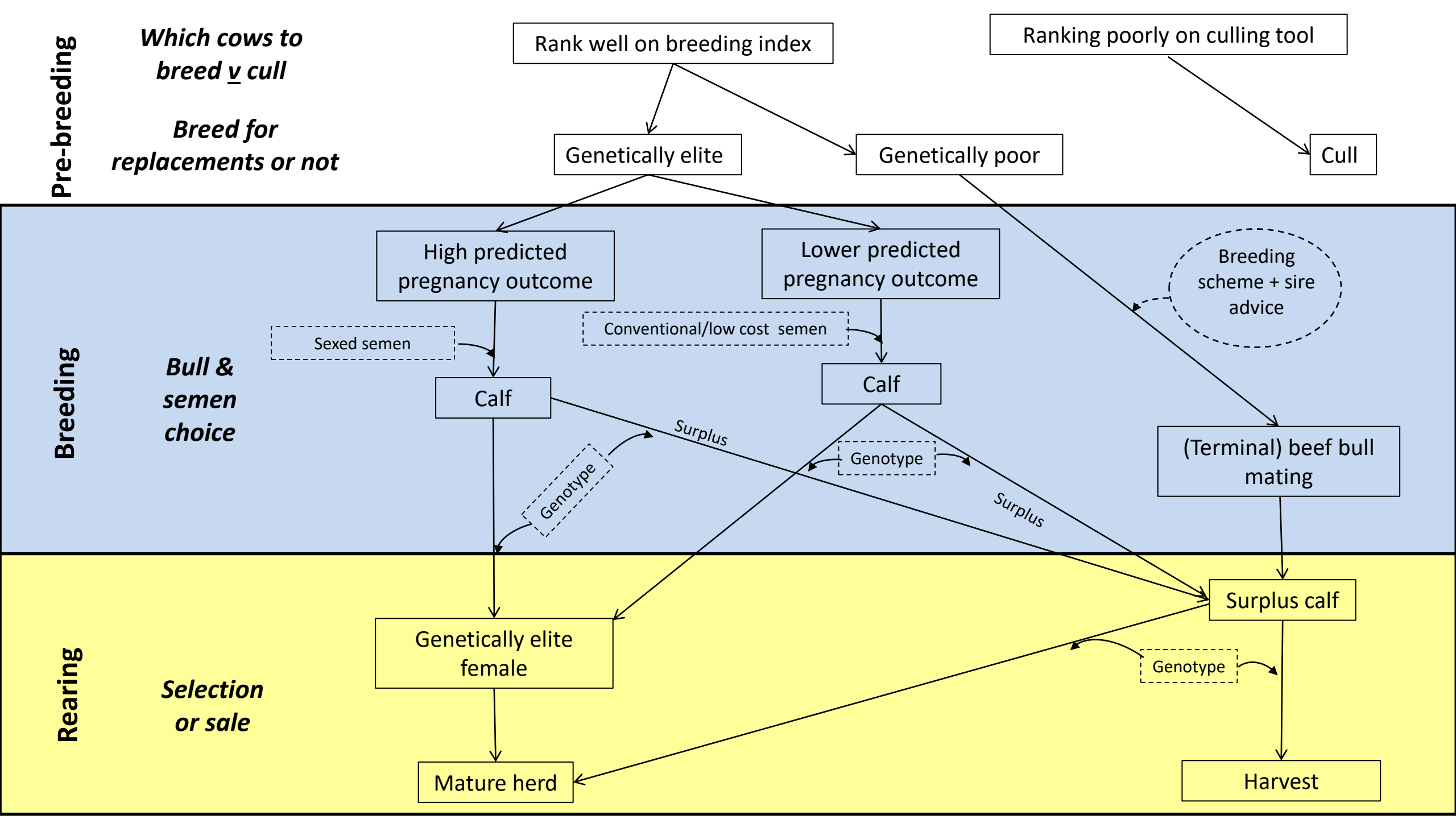
²Irish Cattle Breeding Federation, Ireland

donagh.berry@teagasc.ie

EAAP, Decemeber 2020

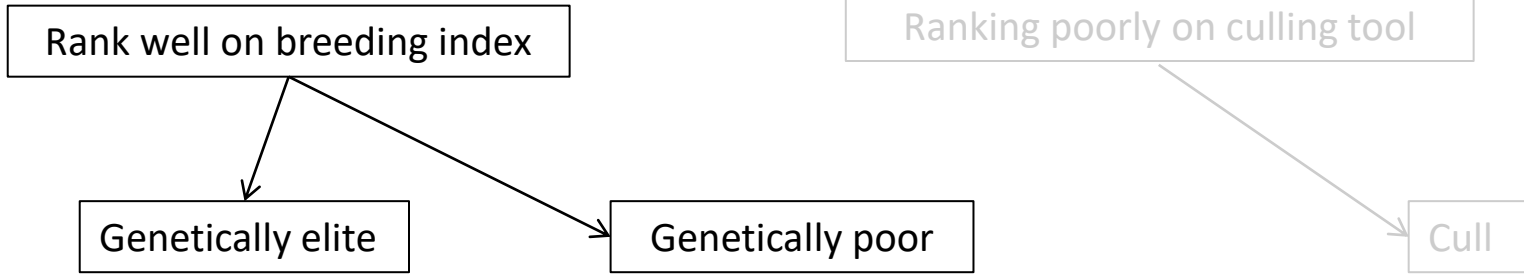






Which cows to breed v cull

Breed for replacements or not



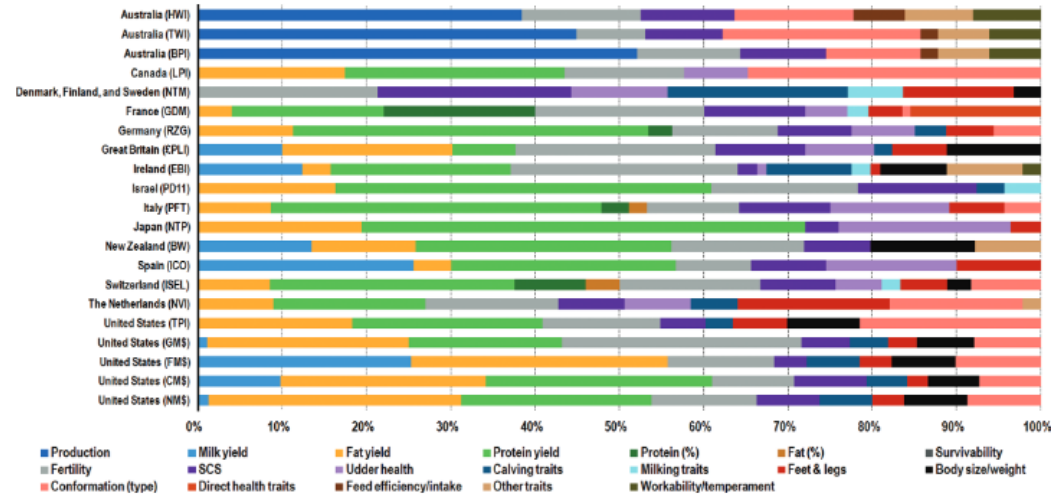
J. Dairy Sci. 101:3686–3701
<https://doi.org/10.3168/jds.2017-13335>
 © American Dairy Science Association[®], 2018.

Symposium review: Possibilities in an age of genomics: The future of selection indices¹

J. B. Cole² and P. M. VanRaden



Livestock Production Science 67 (2001) 223–239



Breeding objectives for beef cattle in Ireland

P.R. Amer^{a,*}, G. Simm^b, M.G. Keane^c, M.G. Diskin^d, B.W. Wickham^e

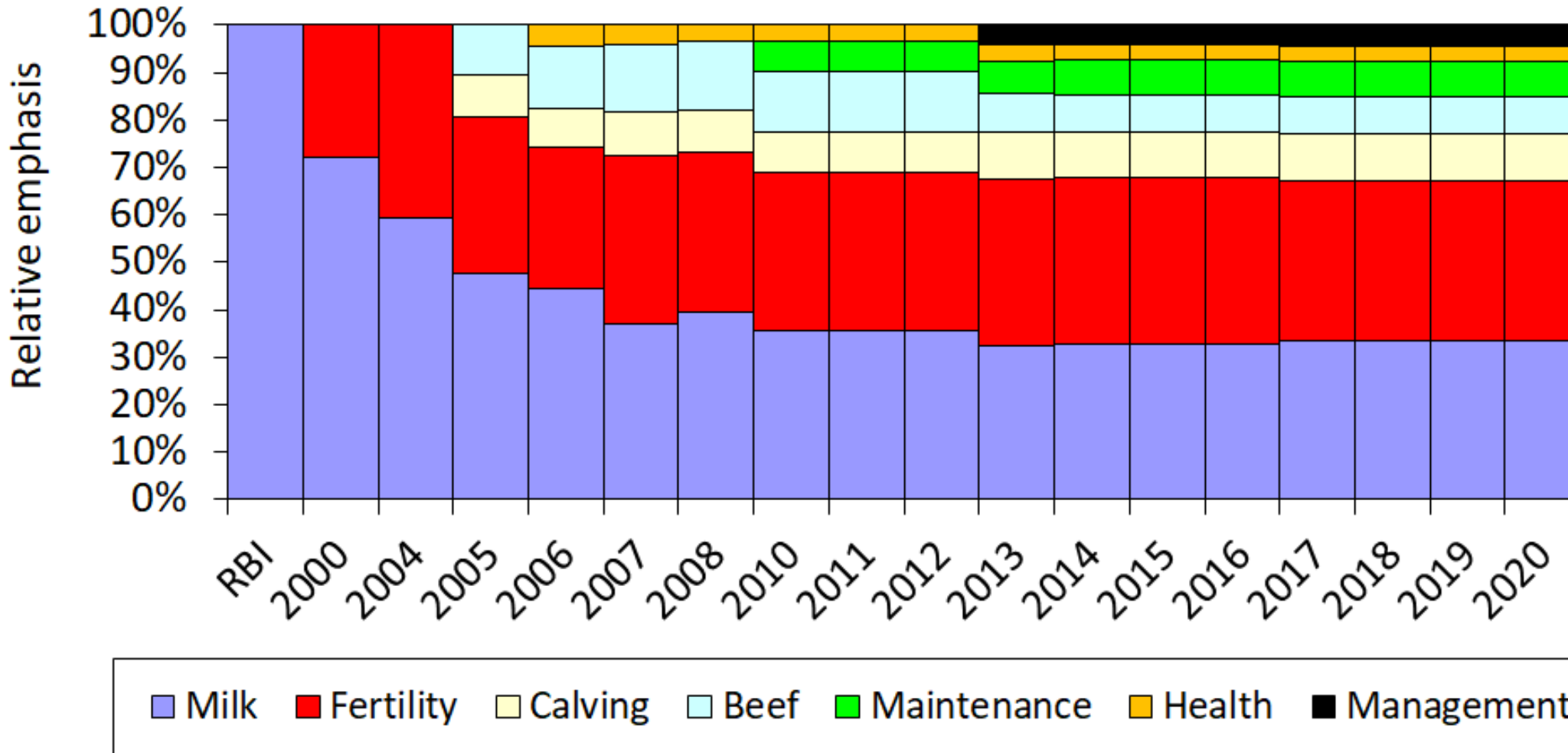
$$H = a_1 \cdot GM_1 + a_2 \cdot GM_2 + \dots + a_n \cdot GM_n$$

$$H = \text{Sub-index}_1 + \text{Sub-index}_2 + \dots + \text{Sub-index}_n$$

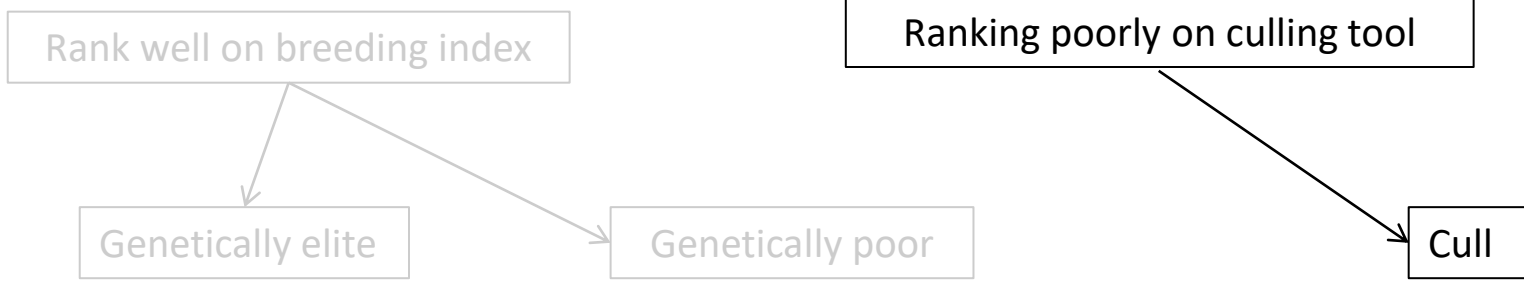
1. Must be important
2. Must exhibit genetic variability
3. Must be measurable (or correlated with a measurable trait)



Evolution of Irish Dairy Breeding Goal (Economic Breeding Index)




Which cows to breed v cull
Breed for replacements or not



An index framework founded on the future profit potential of female beef cattle to aid the identification of candidates for culling

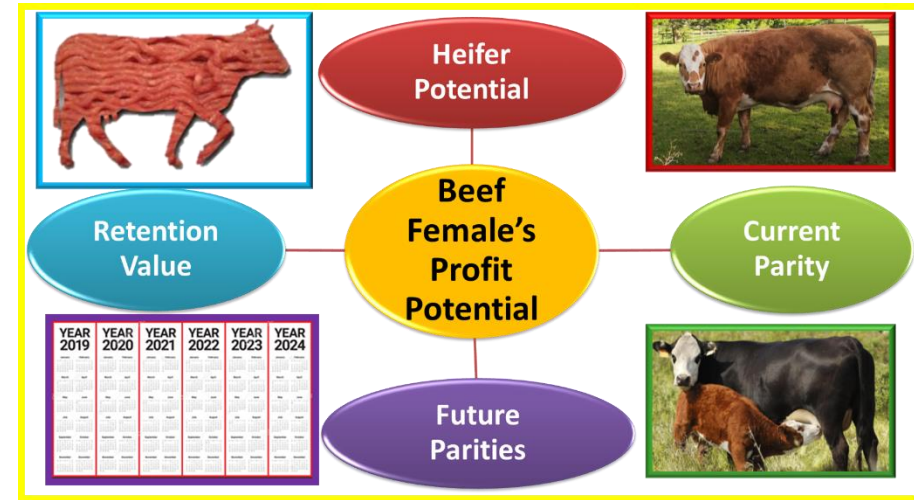
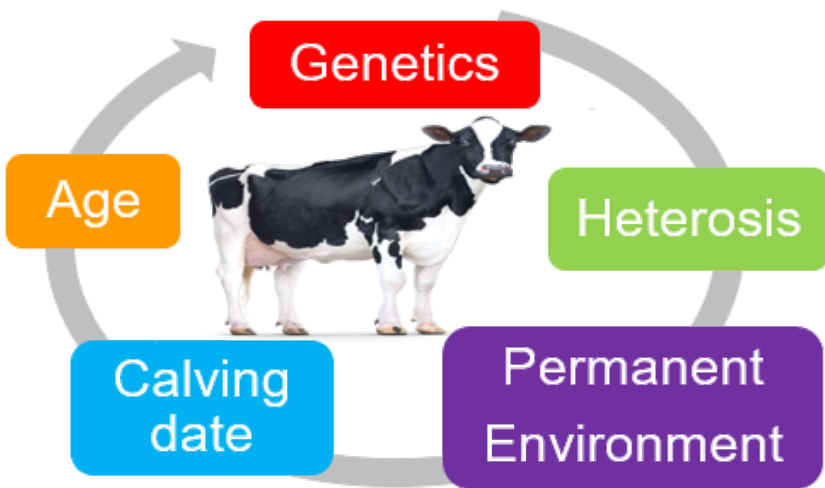
F L Dunne, D P Berry, M M Kelleher, R D Evans, S W Walsh, P R Amer

Journal of Animal Science, skaa334, <https://doi.org/10.1093/jas/skaa334>

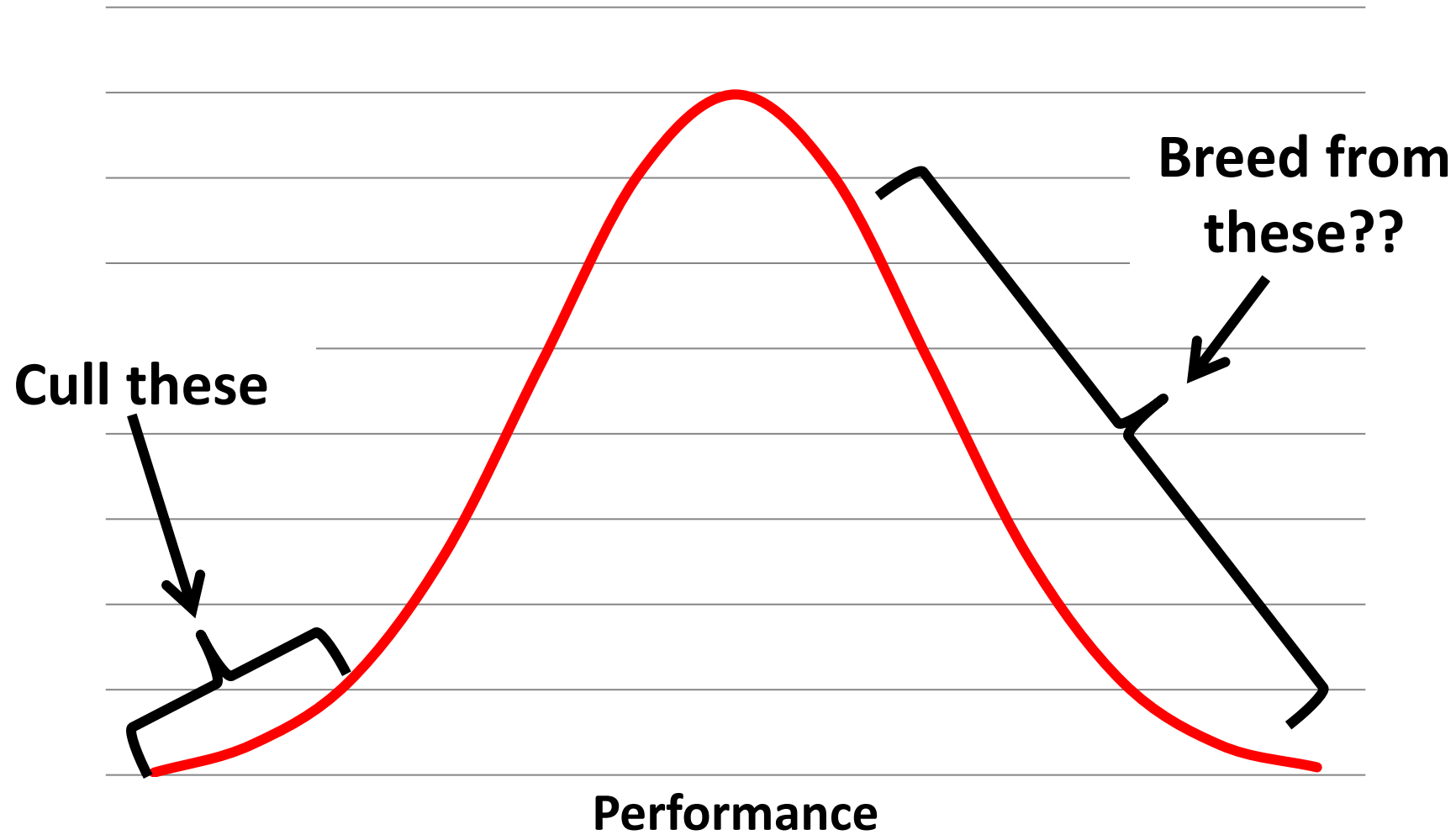
 J. Dairy Sci. 98:4225–4239
<http://dx.doi.org/10.3168/jds.2014-9073>
© American Dairy Science Association®, 2015.

Development of an index to rank dairy females on expected lifetime profit

M. M. Kelleher,*† P. R. Amer,‡ L. Shalloo,* R. D. Evans,§ T. J. Byrne,‡ F. Buckley,* and D. P. Berry*¹



Achieving performance gains



Performance



Phenotype = environmental effects + genetic effects + noise

Milk = parity*age + het + “herd-season” + **genetics** + perm. envir. + e



EBV/PTA



Performance



Phenotype = environmental effects + genetic effects + noise

$$\text{Milk} = \text{parity}^* \text{age} + \text{het} + \text{"herd-season"} + \text{genetics} + \text{perm. envir.} + e$$



Estimated Production Values (EPV)



COW index



C.O.W

=

Current Lactation

- Production
- Management
- Health (SCC)
- Maintenance
- Fertility (calving date)

Net Replacement Cost

- Cull cow value
- Replacement cost

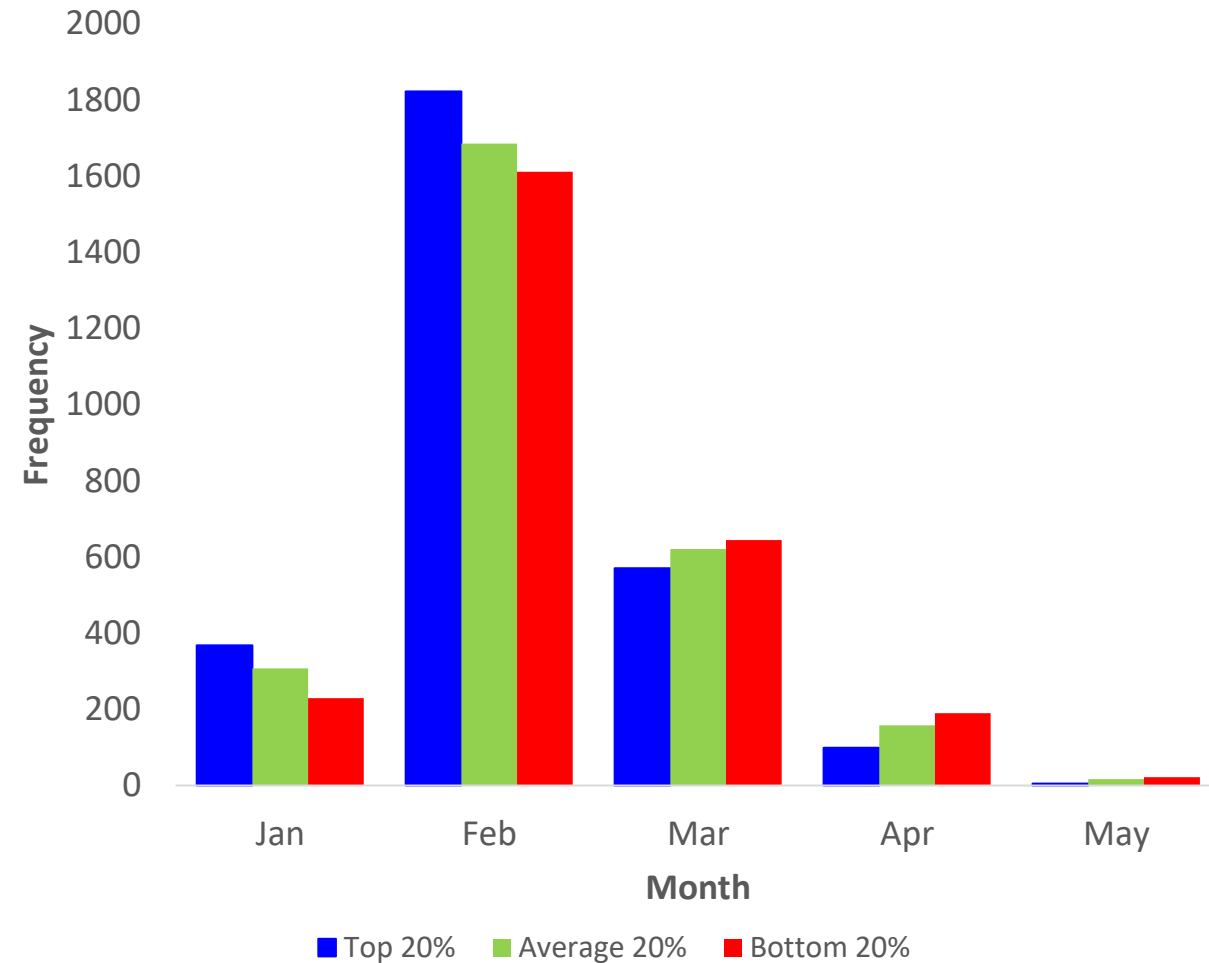
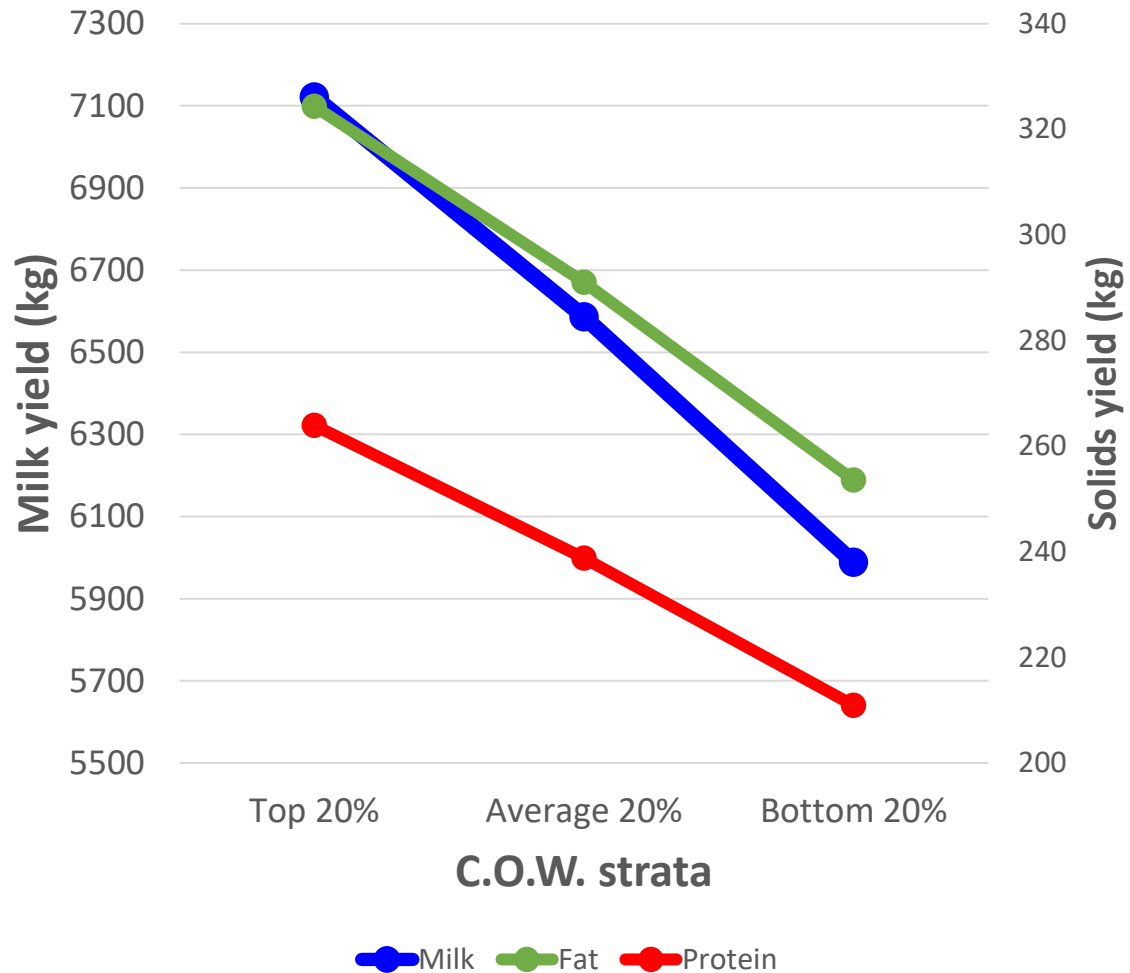
Future Lactations

- Production
- Management
- Health
- Maintenance
- Beef
- Fertility
- Calving
- Descendants

+ *predictions on fertility, survival and SCC performance*



Proof in the pudding - dairy



Deployment

C.O.W. (Cow's Own Worth) Profile Click on an Animal Number or Jumbo to view more details.

Help PDF

Help

Record Events

Missing Sire

7 Missing

Heat & AI/Serve

206 Cows Served

Pregnancy Diagnosis

211 Cows Scanned
18% Empty

Dry-Off

0 Cows

Mastitis & Lameness

0 Mastitis Case(s)
0 Lameness Case(s)

Milking Temperament

0 Scored

Mark For Culling

0 Cows Marked



Eligible Animals

Missing Animals

Please ensure to record health events (particularly mastitis & lameness) to improve the accuracy of your C.O.W.

Showing 1 to 203 of 203 entries

Hide filters

Excel

PDF

Print

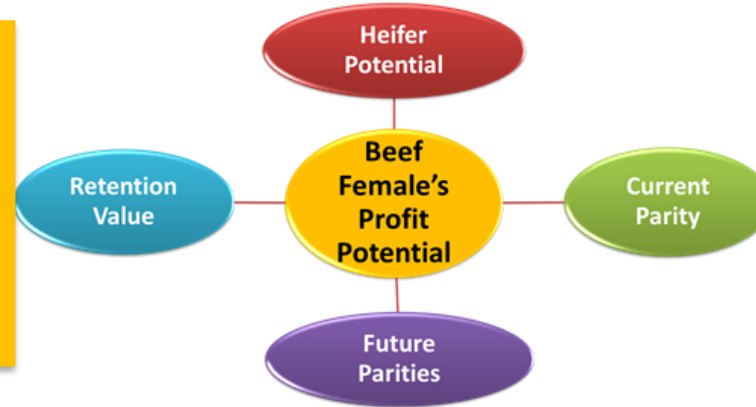
From To Jumbo Animal Number From To From To From To From To From To From To From Date To Date Preg Diag From Date To Date Marked For From To

Rank	Animal Details			C.O.W. Components				EBI	Fertility				SCC
C.O.W. Rank	Jumbo	Animal Number	Lact	C.O.W. (€)	Current Lact Profit (€)	Future Lacts Profit (€)	Net Replace Cost (€)	EBI (€)	Last Calving Date	Preg Diag	Expected Calving Date	Marked For Culling	SCC (,000 c/ml)
1	2127	IE151721782127	2	1895	216	873	805	136	31-JAN-17	IN CALF	07-FEB-18	NO	27
2	2098	IE151721742098	2	1820	190	826	804	152	20-FEB-17	IN CALF	13-FEB-18	NO	20
3	2332	IE151721772332	1	1788	127	754	907	151	08-FEB-17	IN CALF	01-FEB-18	NO	50
4	2262	IE151721732262	1	1779	116	754	909	172	15-FEB-17	IN CALF	11-FEB-18	NO	18
5	1968	IE151721751968	3	1770	210	859	702	130	26-JAN-17	IN CALF	21-FEB-18	NO	14
6	1857	IE151721711857	3	1726	223	802	701	110	21-FEB-17	IN CALF	14-FEB-18	NO	10
7	1733	IE151721711733	4	1643	222	823	598	153	18-FEB-17	IN CALF	28-MAR-18	NO	29
8	2232	IE151721762232	1	1600	106	586	908	155	10-FEB-17	IN CALF	13-APR-18	NO	25

Beef Female's Profit Potential



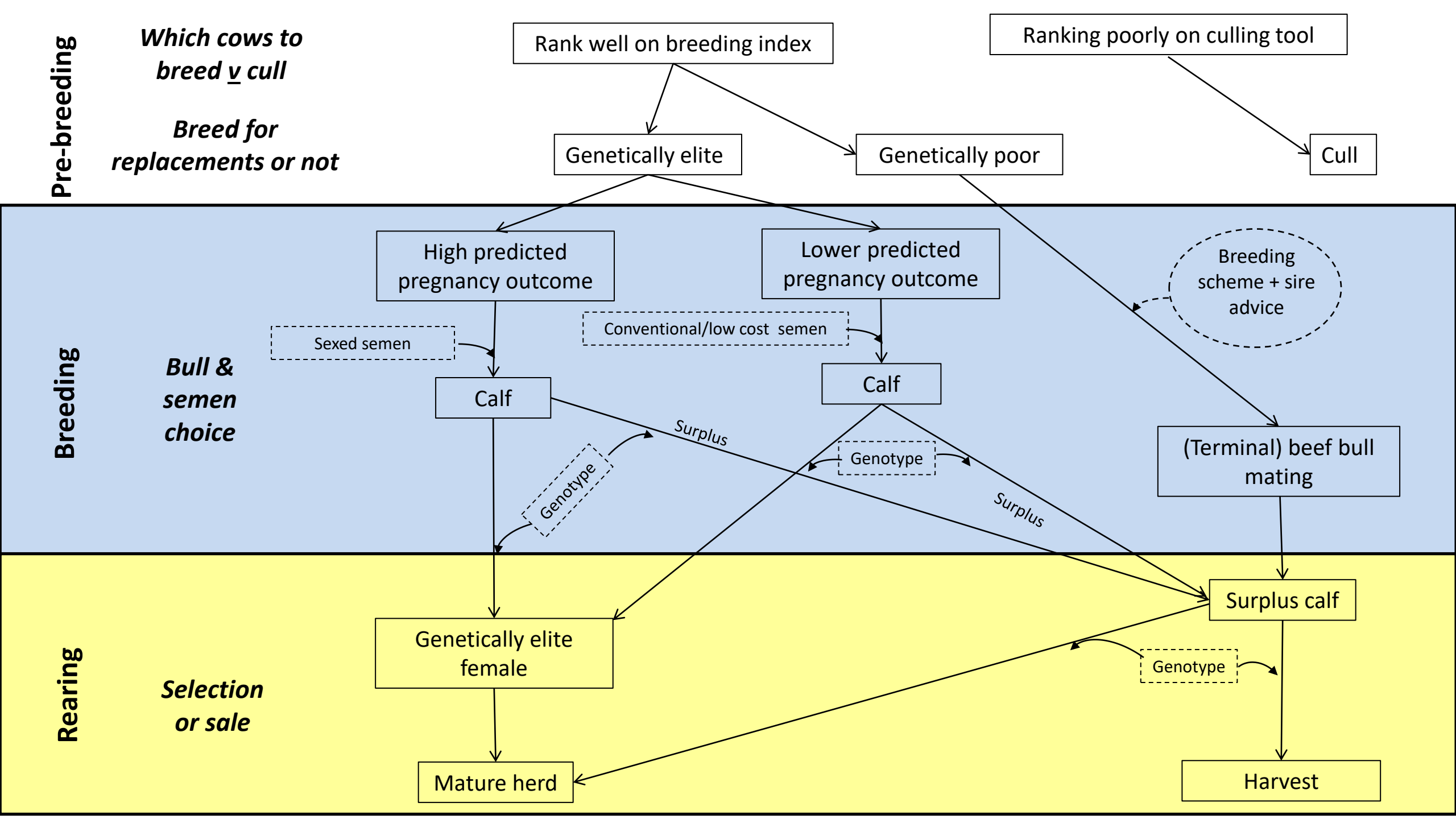
Stratified animals within herd based on BFPP value



		Top 25%	50% to 75%	25% to 50%	Bottom 25%
Cows traits	Calving date	6 th Apr	16 th Apr	28 th Apr	14 th May
	Survival (0 to 1)	1.62	1.49	1.33	1.00

Carcass		Top 25%	50% to 75%	25% to 50%	Bottom 25%
Progeny traits	Weight (kg)	398.46	398.48	396.61	394.29
	Conformation (EUROP)	7.19	7.14	7.07	6.94
	Fat (EUROP)	7.94	7.99	8.04	8.04





Pre-breeding

Which cows to breed v cull
Breed for replacements or not

Rank well on breeding index

Ranking poorly on culling tool

Genetically elite

Genetically poor

Cull

Breeding

Bull & semen choice

Animal (2016), 10:5, pp 736–745 © The Animal Consortium 2016
doi:10.1017/S1751731115002827



Genetic differences based on a beef terminal index are reflected in future phenotypic performance differences in commercial beef cattle

S. M. Connolly^{1,2}, A. R. Cromie³ and D. P. Berry^{1†}



J. Dairy Sci. 102:10056–10072
<https://doi.org/10.3168/jds.2019-16912>
© American Dairy Science Association®, 2019.

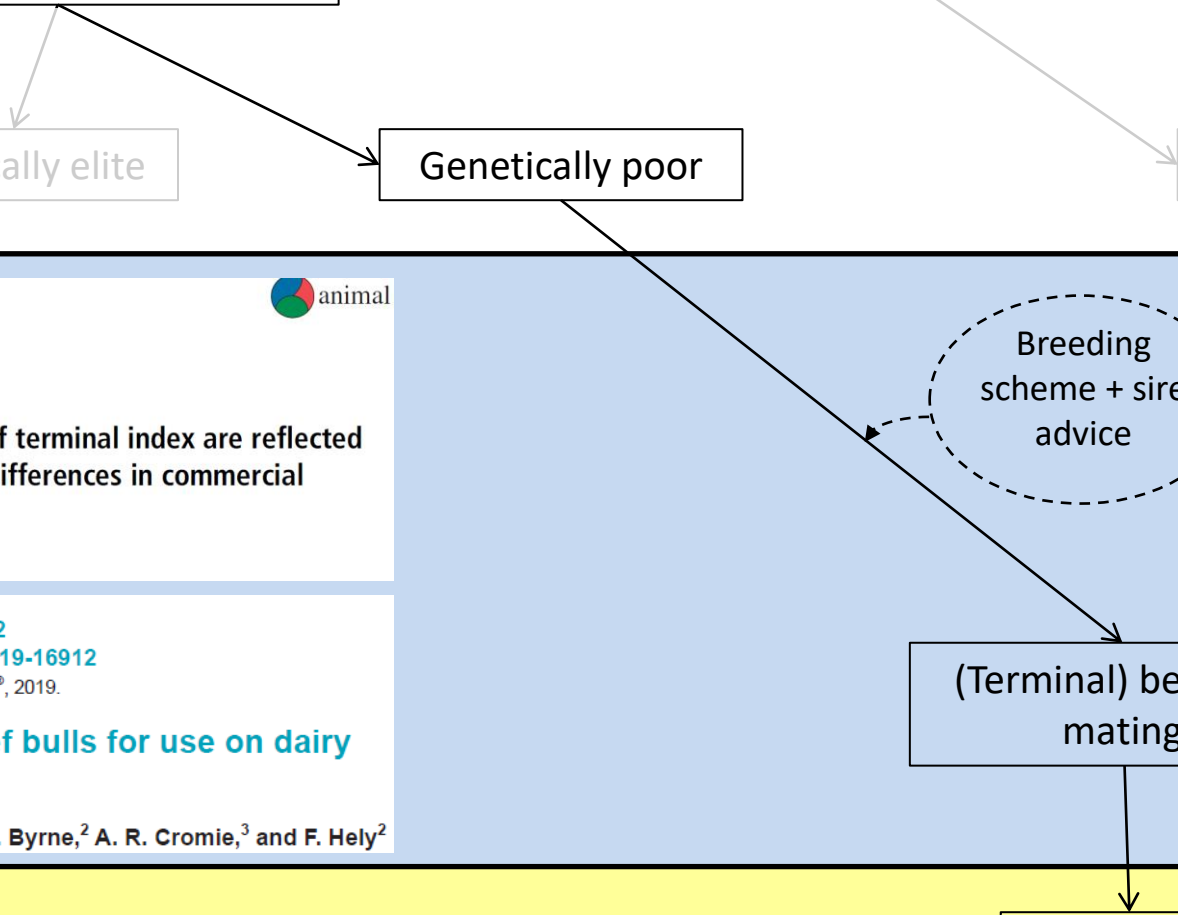
A breeding index to rank beef bulls for use on dairy females to maximize profit

D. P. Berry,^{1*} P. R. Amer,² R. D. Evans,³ T. Byrne,² A. R. Cromie,³ and F. Hely²

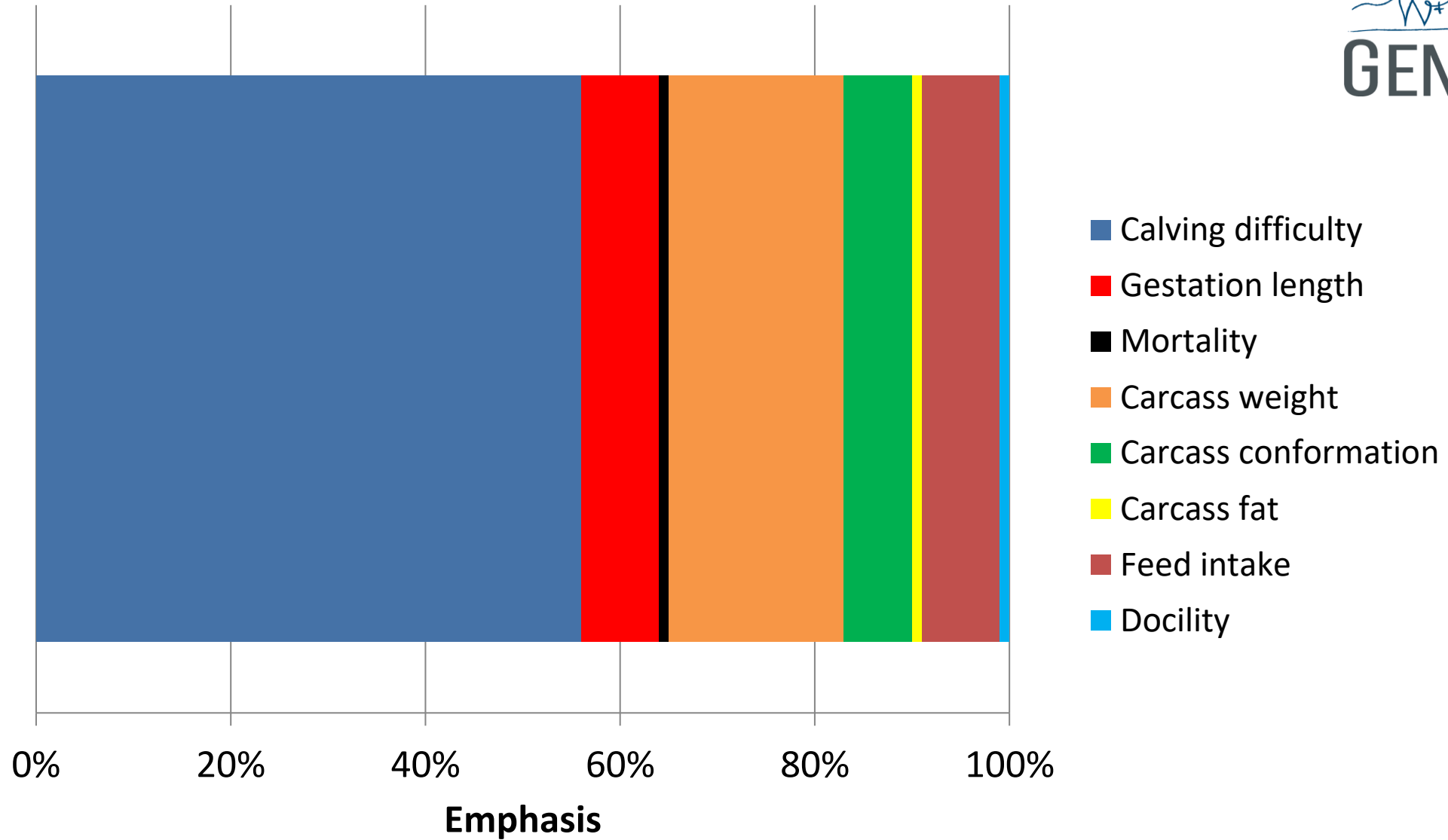
Breeding scheme + sire advice

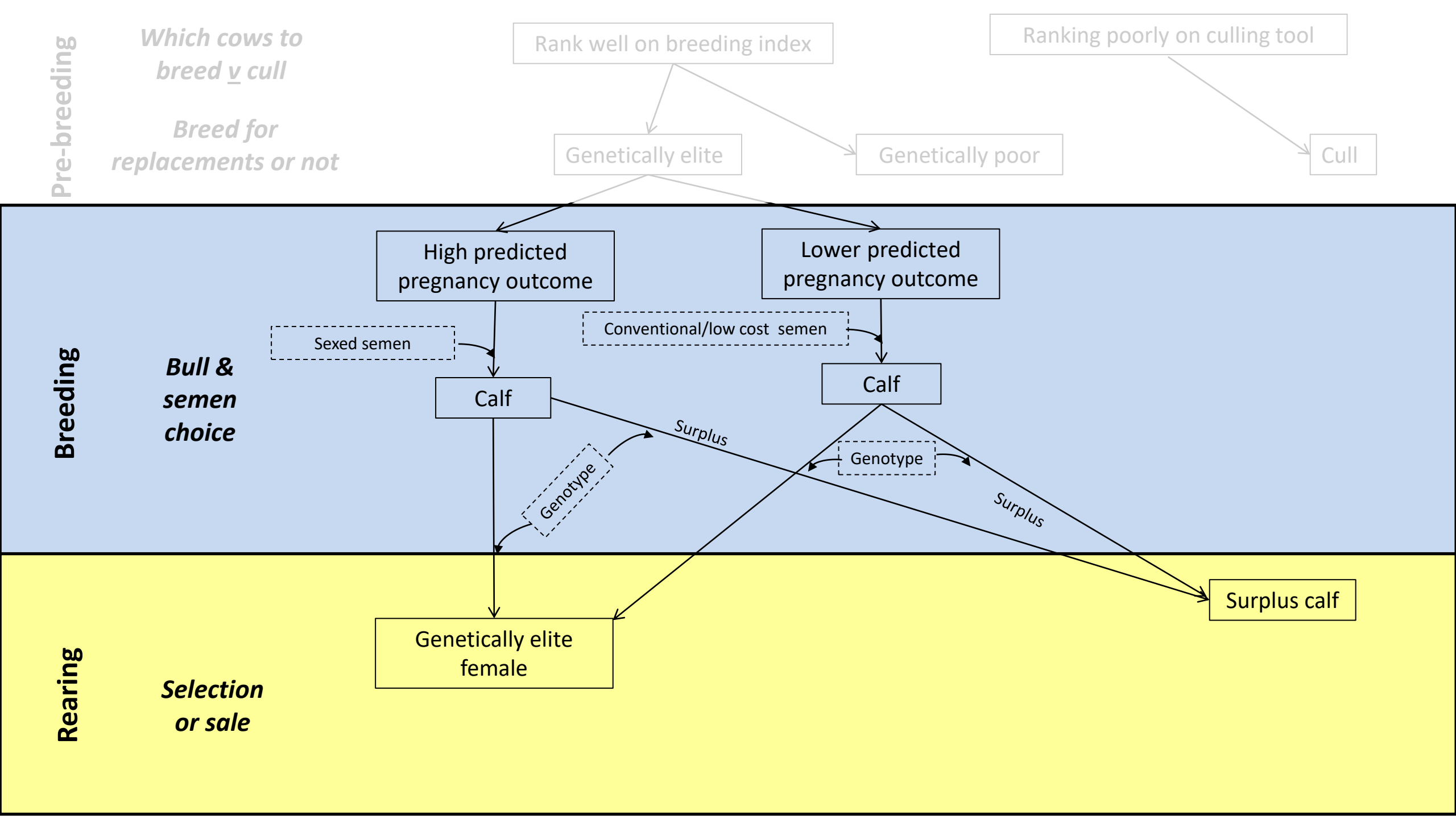
(Terminal) beef bull mating

Surplus calf



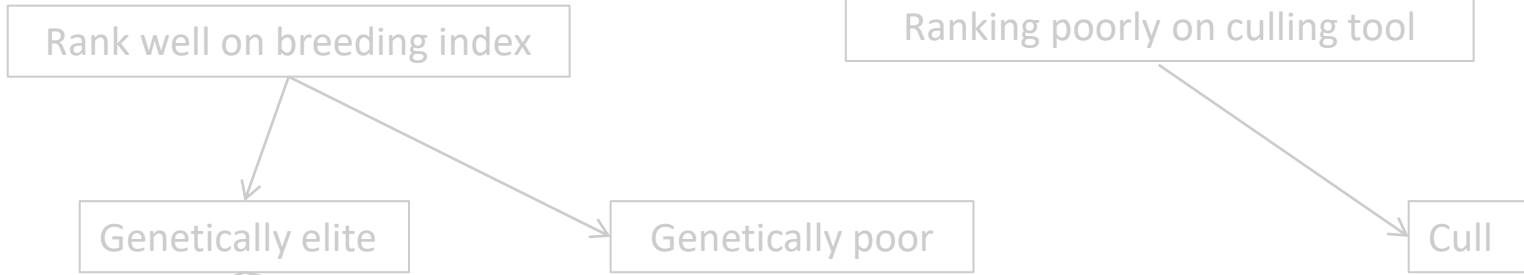
Dairy-beef index – relative emphasis





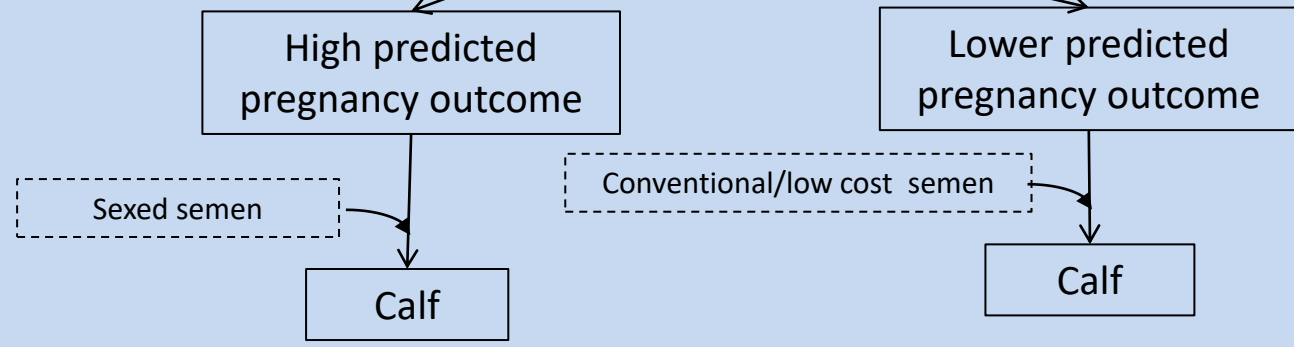
Pre-breeding

Which cows to breed v cull
Breed for replacements or not



Breeding

Bull & semen choice



J. Dairy Sci. 101:1648–1660
<https://doi.org/10.3168/jds.2016-12453>
© American Dairy Science Association®, 2018.

Estimating probability of insemination success using milk progesterone measurements

P. Blavy,^{*1} N. C. Friggens,^{*} K. R. Nielsen,[†] J. M. Christensen,[†] and M. Derks[†]§#



J. Dairy Sci. 100:5550–5563
<https://doi.org/10.3168/jds.2016-11830>
© American Dairy Science Association®, 2017.

The creation and evaluation of a model predicting the probability of conception in seasonal-calving, pasture-based dairy cows

Caroline Fenlon,^{*1,2} Luke O'Grady,^{†1} Michael L. Doherty,[†] John Dunning,^{*} Laurence Shalloo,[‡] and Stephen T. Butler[‡]



J. Dairy Sci. 98:5262–5273
<http://dx.doi.org/10.3168/jds.2014-8984>
© American Dairy Science Association®, 2015.

Machine learning algorithms for the prediction of conception success to a given insemination in lactating dairy cows

K. Hempstalk,^{*1} S. McParland,[†] and D. P. Berry[†]

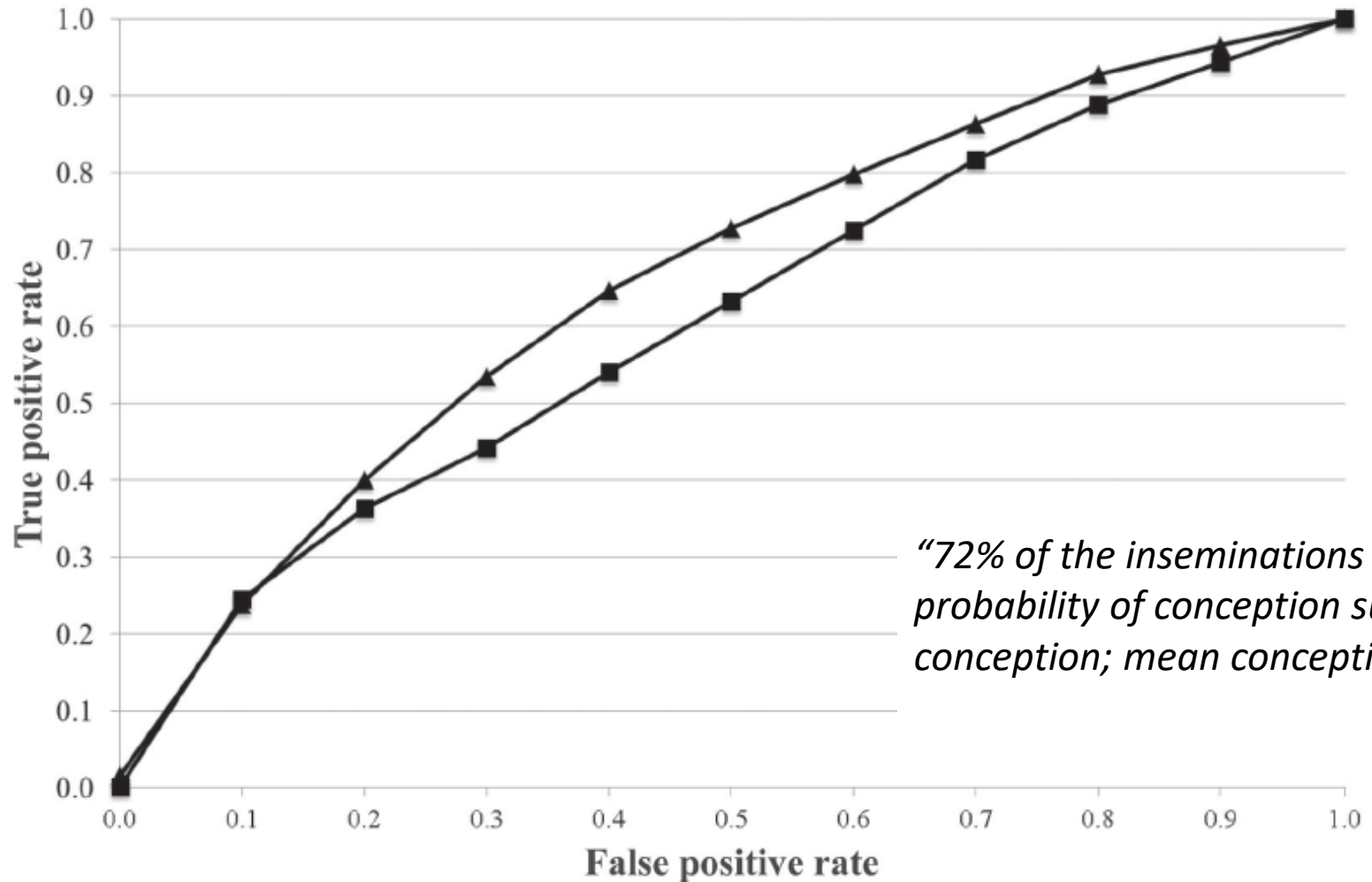


J. Dairy Sci. 102:10460–10470
<https://doi.org/10.3168/jds.2019-16412>
© 2019, The Authors. Published by FASS Inc. and Elsevier Inc. on behalf of the American Dairy Science Association®.
This is an open access article under the CC BY-NC-ND license (<http://creativecommons.org/licenses/by-nc-nd/4.0/>).

Classifying the fertility of dairy cows using milk mid-infrared spectroscopy

P. N. Ho,^{1*} V. Bonfatti,² T. D. W. Luke,^{1,3} and J. E. Pryce^{1,3}

Can we predict pregnancy??



(Future) noise

- Technician (x time)
- Ejaculate
- Bull x cow interactions
- (Herd) temporal
- (True) cow ovulation status

“72% of the inseminations in the top 5% of predicted probability of conception success resulted in a successful conception; mean conception rate...was 54.6%.”

Hempstalk et al (2015)



Sire advice



J. Dairy Sci. 102:8210–8220
<https://doi.org/10.3168/jds.2019-16283>
© American Dairy Science Association®, 2019.

A mating advice system in dairy cattle incorporating genomic information

T. R. Carthy,^{1*} J. McCarthy,² and D. P. Berry¹



J. Dairy Sci. 102:5279–5294
<https://doi.org/10.3168/jds.2018-15971>

© 2019, The Authors. Published by FASS Inc. and Elsevier Inc. on behalf of the American Dairy Science Association®.
This is an open access article under the CC BY-NC-ND license (<http://creativecommons.org/licenses/by-nc-nd/4.0/>).

Variance of gametic diversity and its application in selection programs

[D. J. A. Santos](#),^{1,2*} [J. B. Cole](#),³ [T. J. Lawlor Jr.](#),⁴ [P. M. VanRaden](#),³ [H. Tonhati](#),² and [L. Ma](#)^{1*}

15:45 Prediction of gametic variance and its use in breeding programs

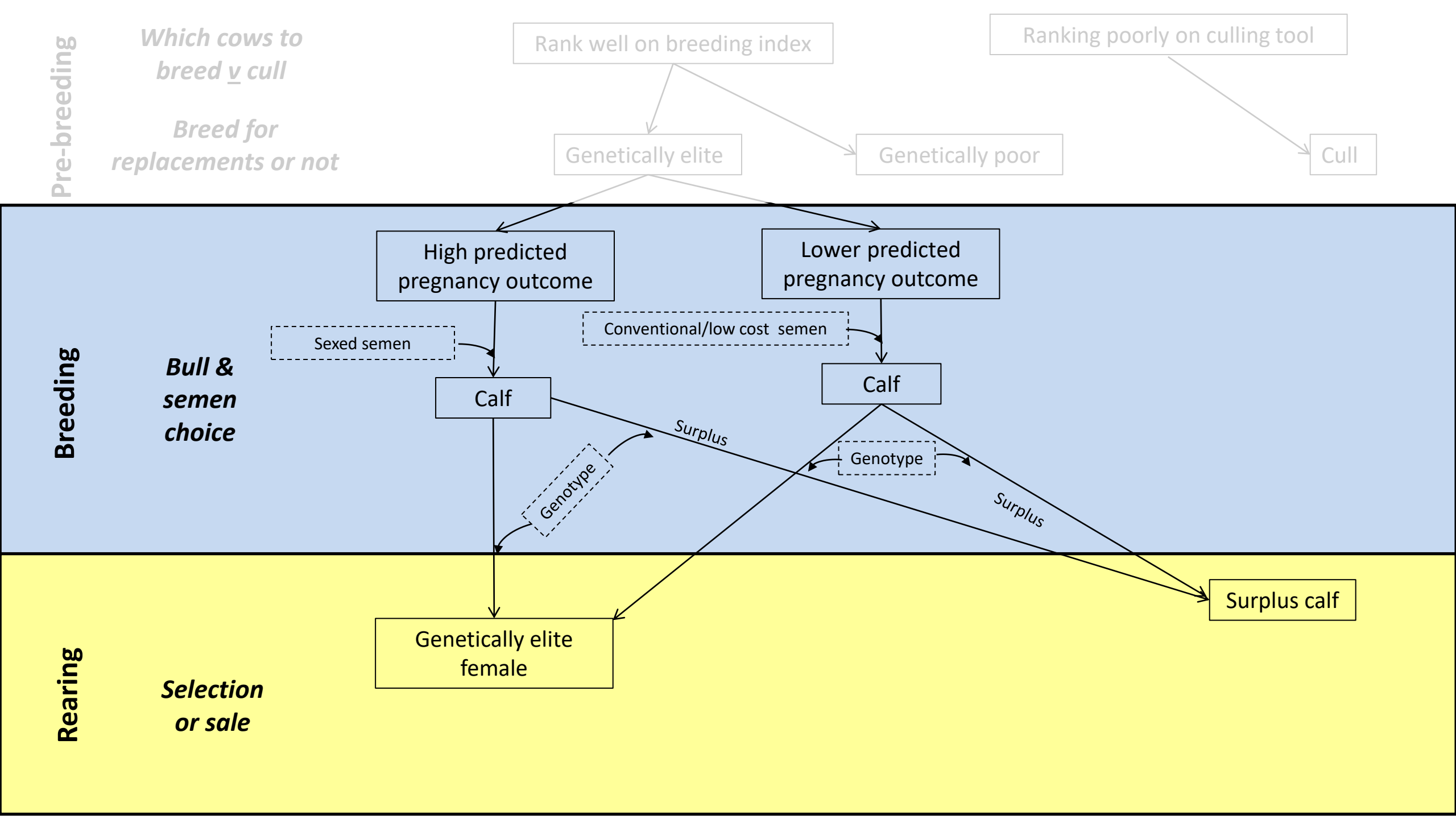
C. Hoze, A. Baur, S. Fritz and D. Boichard

Session 41. Thursday 15:45

Considerations

- Genetic complementarity
 - Major genes
- Coancestry (not for dairy-beef)
- Cognisance of female phenotypic features
- Homogeneity/risk





Pre-breeding

Which cows to breed v cull
Breed for replacements or not

Rank well on breeding index

Ranking poorly on culling tool

Genetically elite

Genetically poor

Cull

Breeding

Bull & semen choice

High predicted pregnancy outcome

Lower predicted pregnancy outcome

Sexed semen

Conventional/low cost semen

Calf

Calf

Genotype

Genotype

Rearing

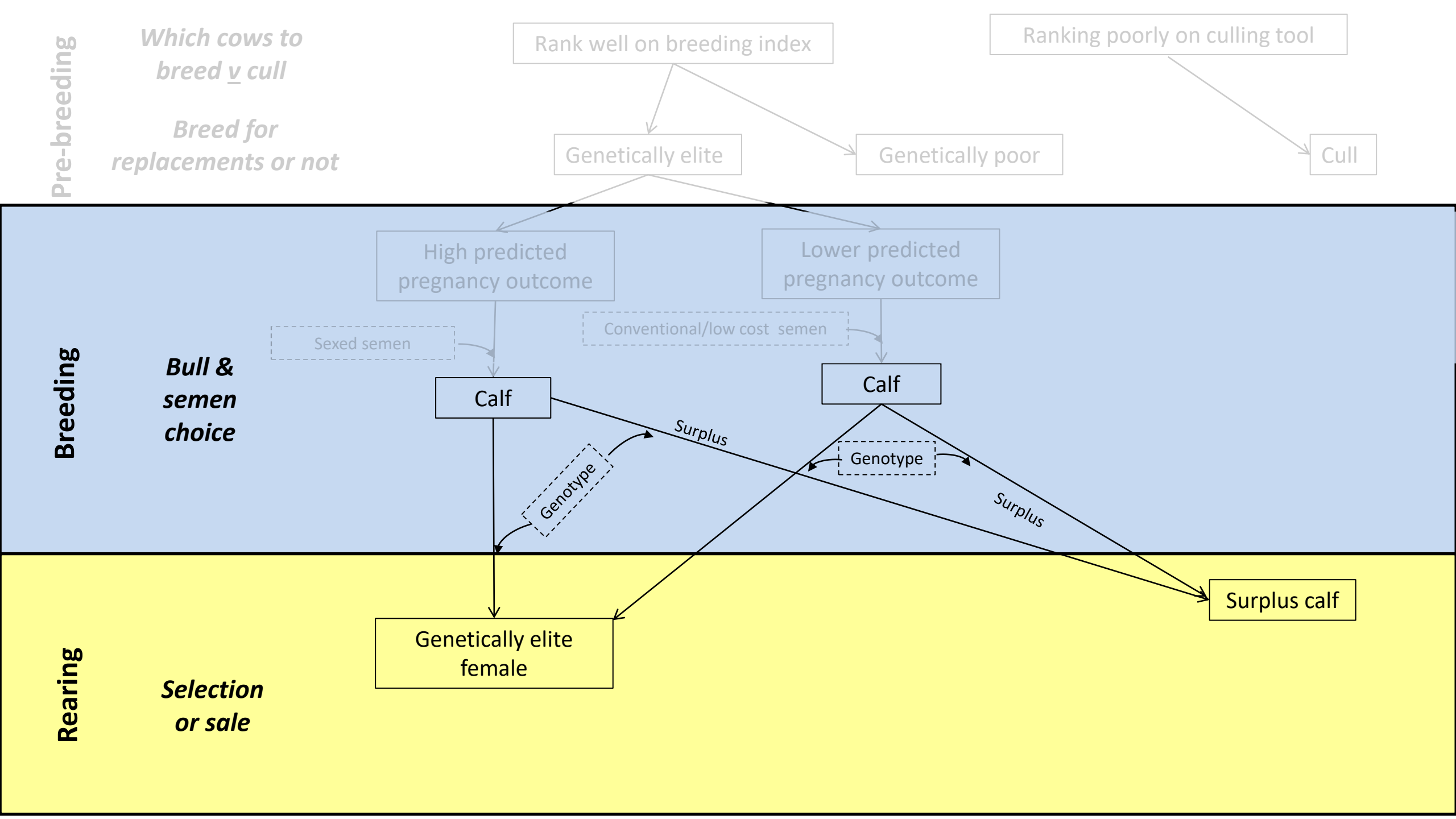
Selection or sale

Genetically elite female

Surplus calf

Surplus

Surplus



Return on investment for genotyping

Animal, page 1 of 11 © The Animal Consortium 2020
doi:10.1017/S1751731120000208

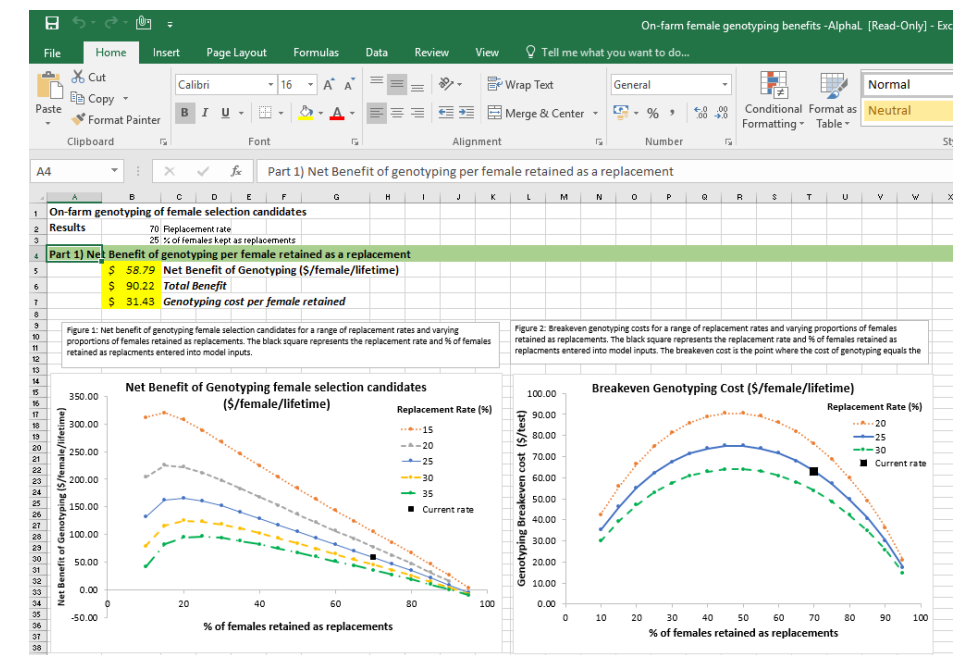


On-farm net benefit of genotyping candidate female replacement cattle and sheep

J. E. Newton and D. P. Berry[†]

Model inputs

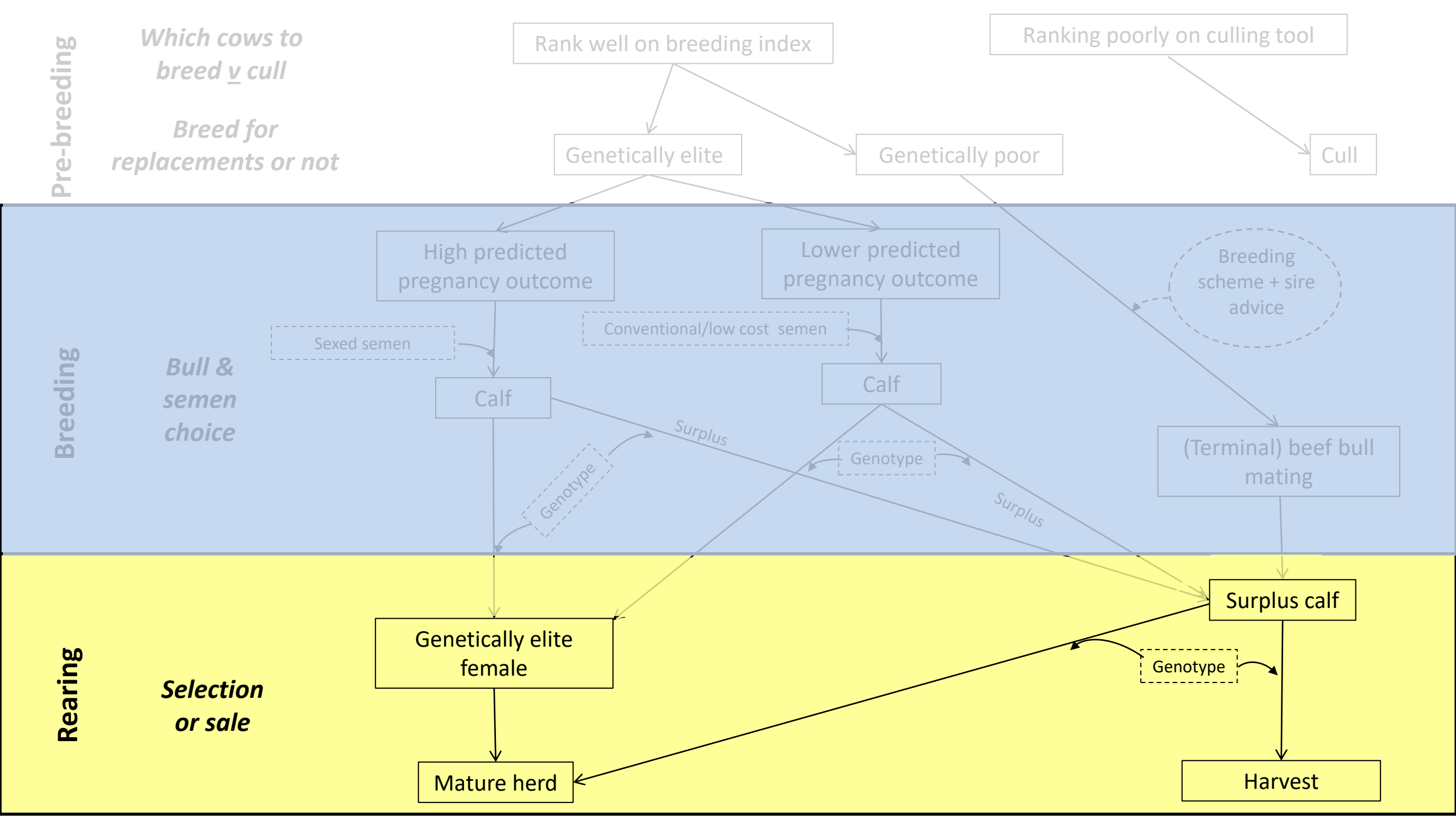
- Proportion of female progeny kept as replacements
- Replacement rate
- Reliability of genomic test
- Sire parentage error rate
- Reliability of traditional estimated breeding values
- Heritability of index
- Genomic test cost - per female
- Actual standard deviation of index



Model outputs

- Net Benefit of Genotyping (\$/female/lifetime)
- Breakeven cost of genotyping

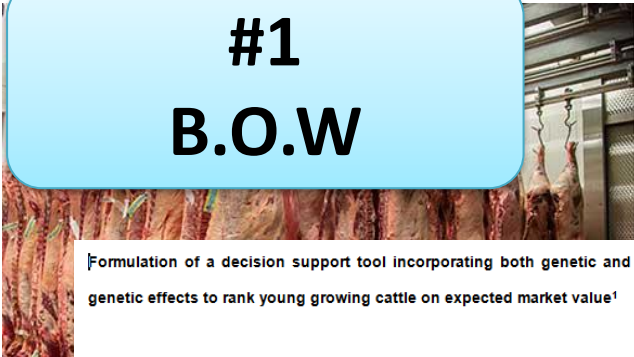




F1 dairy crossbreds



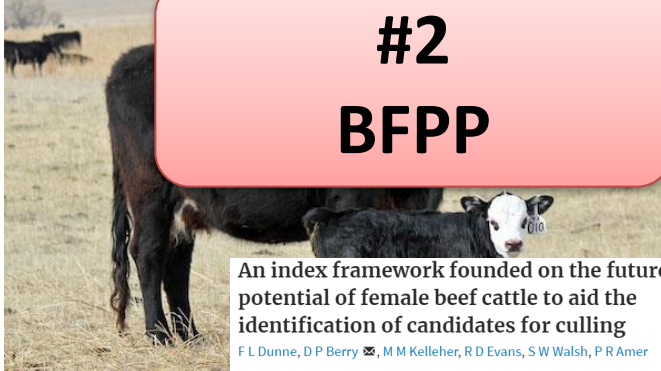
#1
B.O.W



Formulation of a decision support tool incorporating both genetic and non-genetic effects to rank young growing cattle on expected market value¹

F.L. Dunne^{1,2}, R. D. Evans³, M.M. Kelleher³, S.W. Walsh² and D. P. Berry¹

#2
BFPP



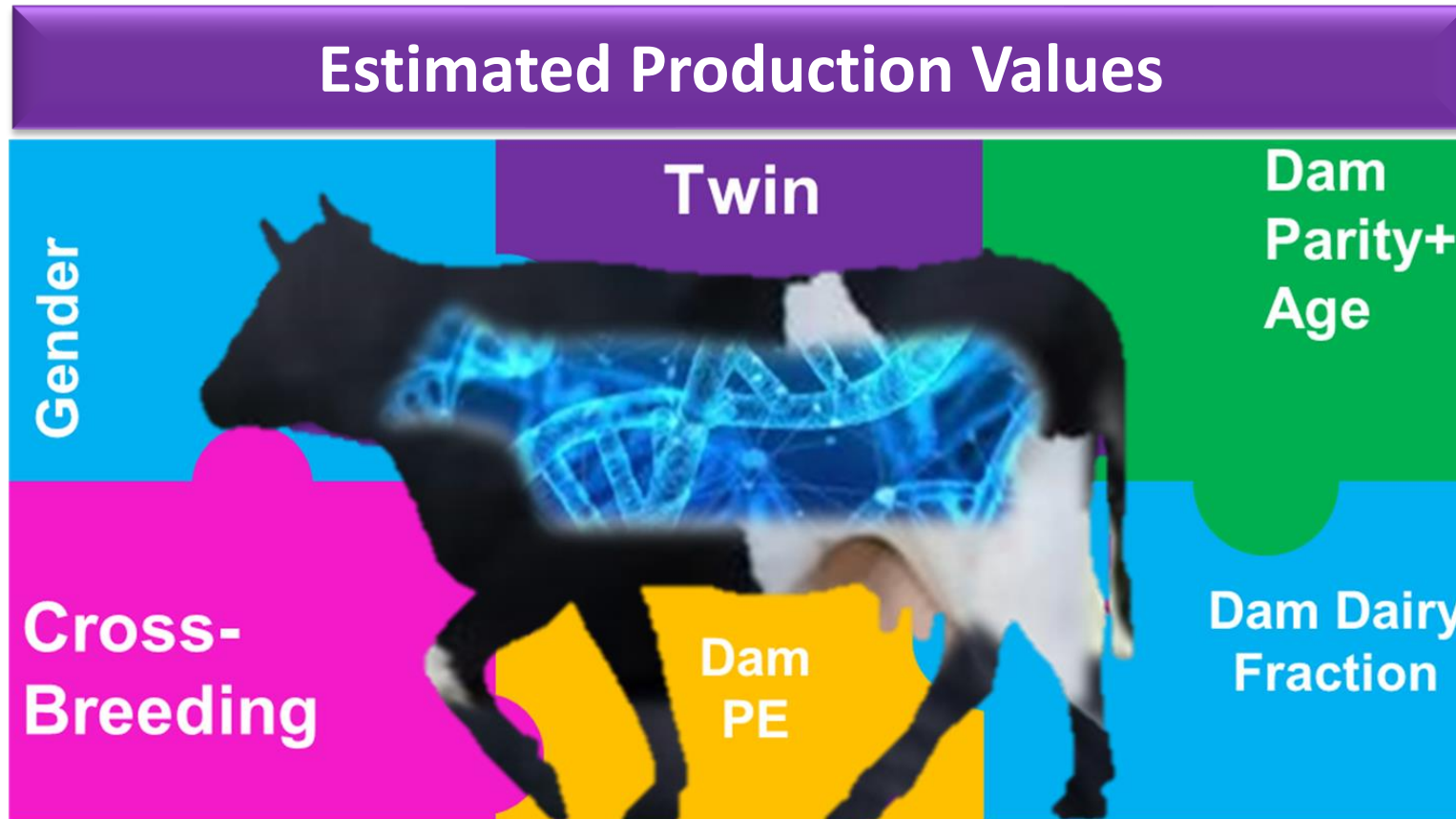
An index framework founded on the future profit potential of female beef cattle to aid the identification of candidates for culling

F L Dunne, D P Berry ✉, M M Kelleher, R D Evans, S W Walsh, P R Amer

Journal of Animal Science, skaa334, <https://doi.org/10.1093/jas/skaa334>



Beef's Own Worth (BOW)



Beef's Own Worth (BOW)



Traits included:

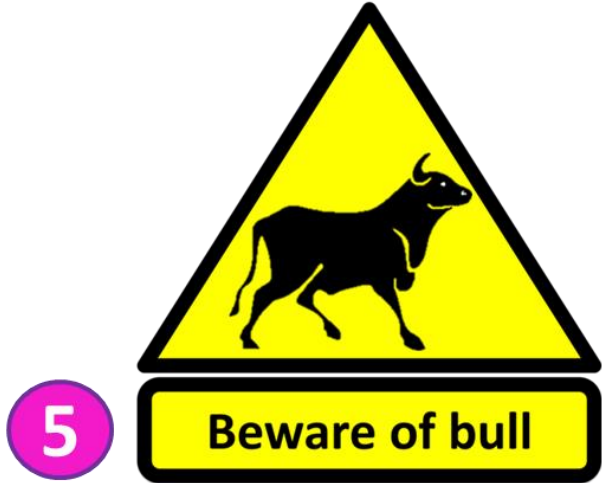


EUROP Classification

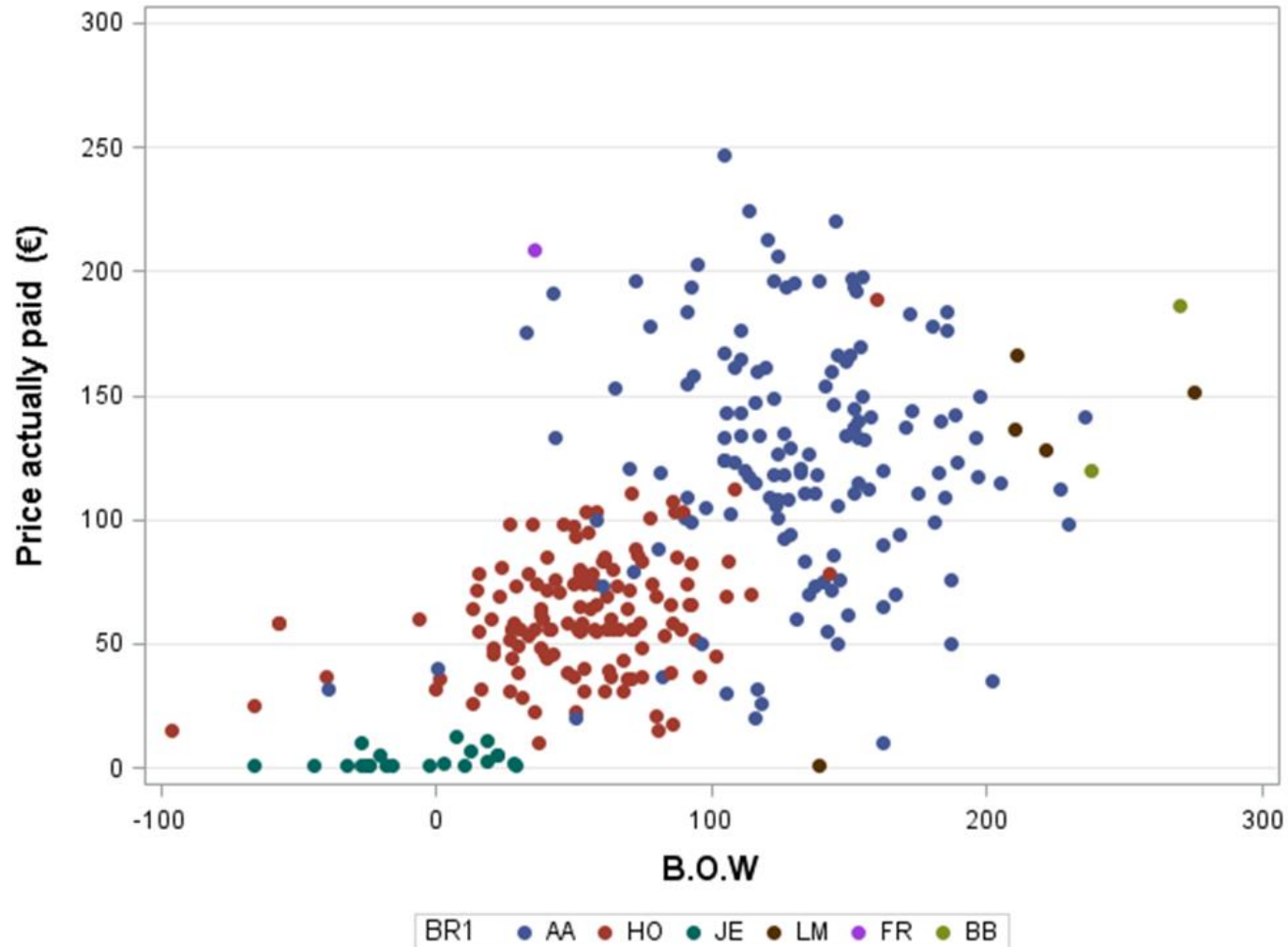
Fat	E+	E=	E-	U+	U=	U-	R+	R=	R-	O+	O=	O-	P+	P=	P-
1-	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
1=	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
1+	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
2-	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
2=	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
2+	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
3-	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
3=	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
3+	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
4-	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
4=	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
4+	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
5-	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
5=	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352
5+	442	436	430	424	418	412	406	400	400	382	376	370	364	358	352

2

3



BOW in practice



Data

Irish auction sales 2019

Calves 10 to 42 days old

Number of animals = 439



Decision support tools



- Modular framework
 - Selection index theory (familiarity)
- Exploits currently available data sources
 - Augmented with additional data
 - Free-notypes
- “Cross compliance” across tools
- Use of herd BLUEs to make more bespoke

