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`Genetics' to `Genomics' ↓ 10-100 DNA >>1000 DNA markers markers

Application for fisheries management

- •Define management units
- •Quantify the extent of divergence and connectivity between stock units
- •Allow mixed-stock analysis with substantially increased resolution relative to morphology based data
- •Species identification in time and space (including prey species)

Genomic analysis example: Atlantic salmon



- Genome: 2.97 billion bases –ACTGCTGAAGGT.....
- •46.598 genes (37.206 coding for a specific protein)
- Links between genes and life histories







Example from Atlantic salmon

Size matters - sometimes



doi:10.1038/nature16062

Sex-dependent dominance at a single locus maintains variation in age at maturity in salmon

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Males and females share many traits that have a common genetic basis; however, selection on these traits often differs between the sexes, leading to sexual conflict^{1,2}. Under such sexual antagonism, theory predicts the evolution of genetic architectures that resolve this sexual conflict^{2–5}. Yet, despite intense theoretical and empirical interest, the specific loci underlying sexually

known genetic architecture means that the evolutionary consequences of sexual conflict, particularly its importance in maintaining adaptive variation^{3,6,16}, remains largely unknown^{14,16}.

The age at which an individual reproduces is a critical point in its life history. Age at maturity affects fitness traits including survival, size at maturity and lifetime reproductive success¹⁷. Age at maturity



Genetics and size in Atlantic salmon



- A single gene (VGLL3) explains 39% of the variation in age at maturation (= cize at river return)
- Ger Implications for management
- Ger and stocking programmes

Barson et al. 2015

7



Types of population/stock structure – a theoretical example



Laikre et al. 2005



Genetic differences arise due to random and evolutionary processes

How are fish stocks managed?





Greenland cod fishery



ICES Advice 2011

Wieland & Paulsen 2004

DTU





Sara Bonanomi







Historical commercial catches of Atlantic cod (*Gadus morhua*) in Greenlandic waters



DNA from 100 archived otoliths for each year



SCIENTIFIC **Reports**

OPEN Archived DNA reveals fisheries and climate induced collapse of a major fishery

Received: 17 March 2015 Sara Bonanomi^{1,2,†}, Loïc Pellissier^{3,†}, Nina Overgaard Therkildsen^{2,4}, Rasmus Berg Accepted: 16 September 2015 Hedeholm^{2,5}, Anja Retzel^{2,5}, Dorte Meldrup¹, Steffen Malskær Olsen⁶, Anders Nielsen⁷, Published: xx xx xxxx Christophe Pampoulie⁸, Jakob Hemmer-Hansen¹, Mary Susanne Wisz^{9,‡}, Peter Grønkjær^{2,10,‡}







Cod example 2



Atlantic cod in the Baltic Sea





Atlantic cod in the Baltic Sea

DTU

Scientific cruises sample mixed stocks



Genetically determined stock afiliation as either 'Eastern' (blue) or 'Western' (red) Baltic cod



Hemmer-Hansen et al. unpublished

Samples of juveniles (<20cm) reveal recruiment dynamics





Highly dynamic distribution in time and space, likely also in response to climatic and biological drivers





Hemmer-Hansen et al. unpublished

Atlantic cod in the W Baltic Sea – genetic marker based split stock assessment



The next 'perfect' example



Atlantic herring stocks and their mixing is now assessable with genetic markers throughout species' distibution



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Rento Lab Stories

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State of the art: Taking advantage of genomics to do genetics

ATGCCCTAAAC TACTGA CCCTGCTTTTA GAAATGCTC TCGATTGCATATGC CTAAA GGGTACTGACCCTGCTTGT AGAA GCTCTCGATGCATA TGC CTAAAGGGTACTGAC CCTGCTA TAGAAATGCTCT CATGCATATGCCCTAAATG GGTACTGACCCTGCTAGTA ATGCTCTCGATGCATAT CCCC TAAAGGGTACTGACC CTCTTTAGAAATGCTCTC AGCATATC

Sequencing genomes (~800*10⁶ bases) to obtain information about stock specific differences



From 9 mio. DNA sites to 48

Pelagic AC 06/06/2018

जिल्लामा तडात



BioMark Fluidigm platform



Gene frequency differences and how they are used to determine the origin of a fish – an example

North Sea autumn spawning herring Gene 1 AA = 25% of fish AC = 22%CC = 53%Gene 2 TT = 99% of fish TG = 1%GG = 0%

Norwegian spring spawning herring

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AA = 81% of fish
AC = 18%
CC = 1%
TT = 1% of fish
TG = 1%
GG = 98%
```



Probability of genotype (Gene 1 & Gene 2) in: North Sea = 0

Norwegian = 0.8

Population splits addressable with new markers

Area	ICES management area/sub area	Stocks/populations assumed to contribute to mixed aggregation fisheries	
Norwegian Sea	Div 2.a	ISSH, NSSH, NASH, NSAS, and potentially WBIH	
North Sea and English Channel	Sub Area 4 + Div 7.d	WBSS, NSAS, NSSH, and potentially ISSH, NASH, WBIH, Downs	In testing
Skagerrak	Div 3.a Skagerrak	WBSS, NSAS, Downs, NSSH, CBH	In testing
Western Baltic Sea	Div 3.a Kattegat and SD 22, 23, 24	WBSS, NSAS, NSSH, CBH	In testing
Ireland and West of the British Isles	Div 6.a, Div 7.b-c	WBIH, and potentially NSSH and NSAS	In testing



When can we trust genetic data on stock units?



Robustness of result depends on the depth of the analysis



And the list of species and stocks with genetic data keeps getting longer: Atlantic mackerel, sprat, mesopelagics,

Novel approaches to the application of molecular methods in ecosystem and fisheries assessment













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Silage

- Fish was dissolved in \sim 10L of formic acid
- 250 µl samples taken out and analyzed on qPCR.





Hansen et al. (unpublished data)

Costings of SNP analyses for input to assessment

Platform	Cost per individual (@ 2000 fish)		sh)	Bench & bio- informatics time 2000 fish
# markers typed	50 SNP	100 SNP	200 SNP	100 SNPs
SNP array based (Fluidigm)	14€	20€	34€	31 days
GT sequencing based*	4€	5€	6€	~17 days

(* Estimate from Aykanat *et al.*, J Fish Biol., 2016)

- Genetic studies are expensive Not really and prices are constantly decreasing
- Genetic results are often 'oversold' We have several cases proving the opposite. Genetic analyses cannot answer all management relevant questions, but lack of basic biological information leads to risk of less efficient management strategies
- Inconsistent interpretation of genetic results Structural, historical sampling issues
- Importance of genetic information is far outweighed by other inputs to management decisions – Substantially better resolution with genetic markers. Time series are emerging. Potential for front user-based analyses onsite.
- Obstacle is implementation into assessment and management

