



The power of billions: data-driven innovation

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23andMe overview

- Founded in 2006
- Product helps individuals learn more about their health, traits, and ancestry through their DNA
- Customers can choose to participate in online research, fill out surveys
- 10M+ kits sold, ~80% of customers consent to participate in research
- **The largest research platform with both genetic and phenotypic information**



Unique interconnected business



Consumer

Providing direct access to learn about own genetics: health, traits, wellness, and ancestry

Empowering easy research participation



Research

Recruiting consenting participants from diverse ancestries and particular diseases of interest

Making discoveries that help Therapeutics and power new insights for customers

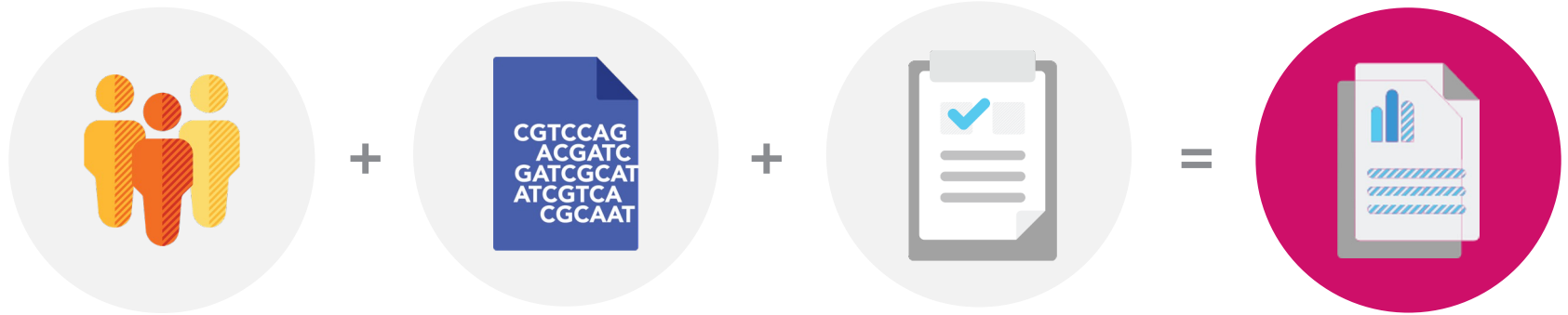


Therapeutics

Leveraging data to accelerate drug discovery

Addressing unmet medical needs and rare conditions

The 23andMe research platform is a new paradigm in biomedical research



Each research participant provides
500,000+ genotypic data points



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Participants contribute phenotypic data via online surveys

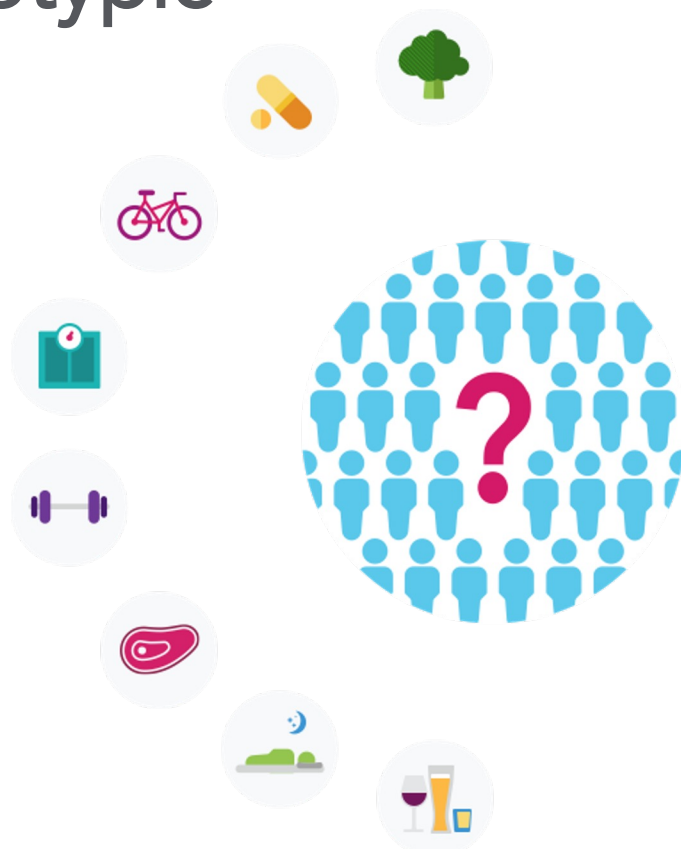
Disease diagnoses

Medication usage, side effects

Behavior and lifestyle

Environmental exposures

Family history



Engaged consumers create the foundation for crowdsourced science

7M+

customers
consented to
research

75%

take \geq 1
survey

~50%

log in
quarterly

2B+

survey
questions
answered

What's unique about 23andMe?

Sheer size of disease cohorts

19,111
with Parkinson's

2,860,246
confirmed controls

1,019,600
APOE e4 carriers

2,954,800
confirmed controls

107,005
with cancer (non skin)

2,603,630
confirmed controls

744,463
with depression

2,177,226
confirmed controls

293,853
with asthma

1,386,943
confirmed controls

115,622
with psoriasis

2,806,720
confirmed controls

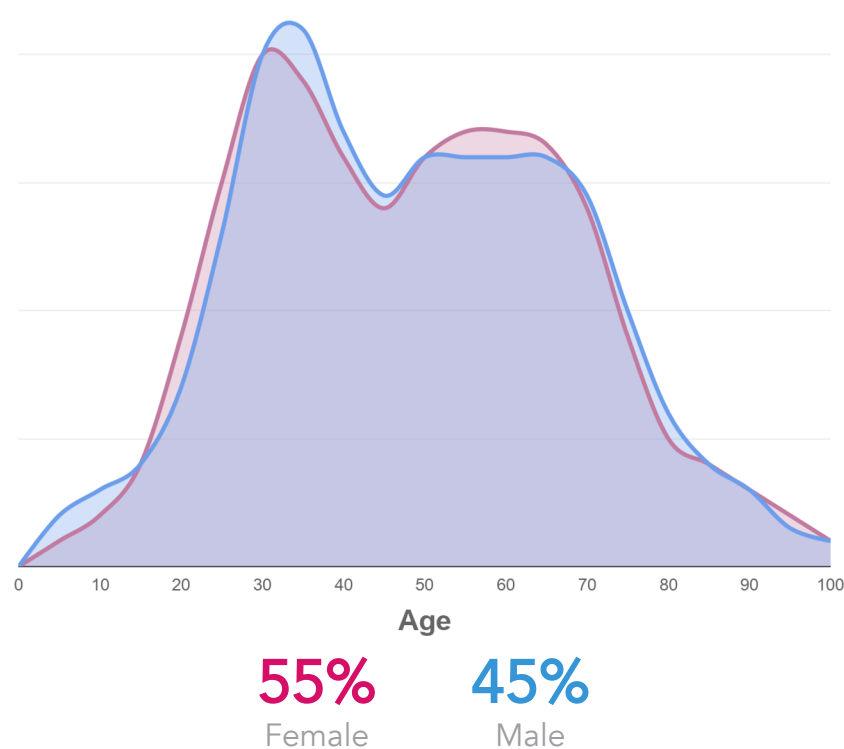
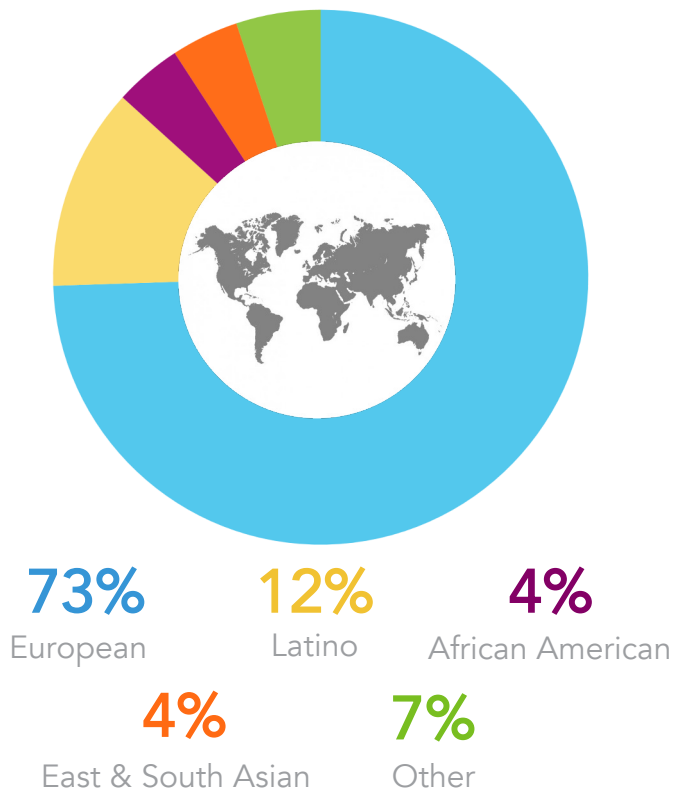
819,371
with CVD

2,258,687
confirmed controls

11,425 with
colorectal cancer

2,945,316
confirmed controls

Database diversity and efforts to improve



Phenotypic breadth

Data sources

Longitudinal, high priority surveys

- Health intake
- Wellness

Stand-alone surveys and one-off questions (targetable)

Emerging data sources

- Passive data
- Lab values

Topics

Addiction	Laterality
Autoimmune	Lifestyle, sleep, sports
Blood	Longevity
Cancer	Metabolic
Cardiovascular	Morphology
Cognitive	Musculoskeletal
Diet	Neurological
Drug usage	Personality
Drug efficacy	Pigmentation
Drug side effects	Pregnancy
Endocrine	Psychiatric
Environment	Renal
Eyes	Reproductive
Food preferences	Respiratory
Gastrointestinal	Sensation
Hair	Sexual orientation
Immune	Skin
Infection	Teeth

What can you do with
billions of data points?

Faster, better informed drug discovery

- Detect genetic associations with disease and identify potential drug targets
- Understand pathways, drug mechanism of action, and potential side effects to improve success rate
- (Future) Precisely stratify patient populations to make clinical trials more efficient and successful

GWAS

Fine-mapping

PheWAS

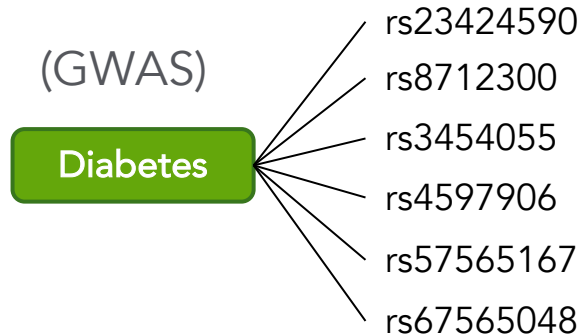
Environmental data

Application of public data

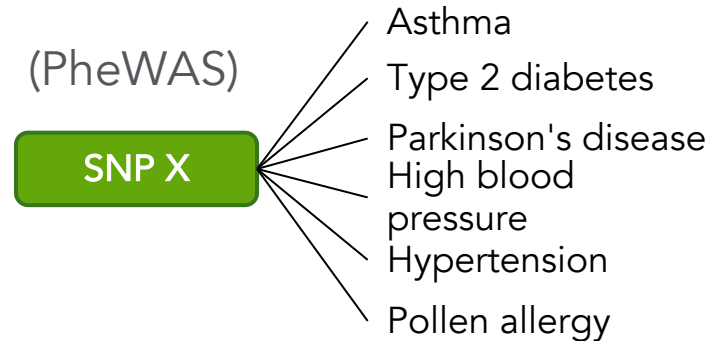
Polygenic risk scores

GWAS and PheWAS at scale

Identifies SNPs associating with phenotype of interest



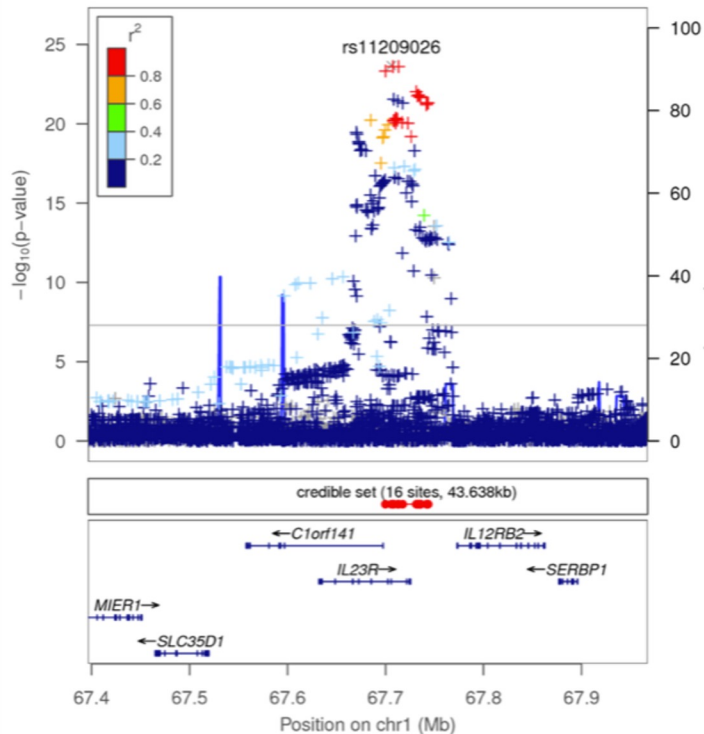
Identifies phenotypes that associate with SNP or gene of interest



We regularly perform GWAS with millions of SNPs for each of hundreds of phenotypes relevant to human disease, and perform PheWAS on all 20,000+ protein-coding genes in the human genome.

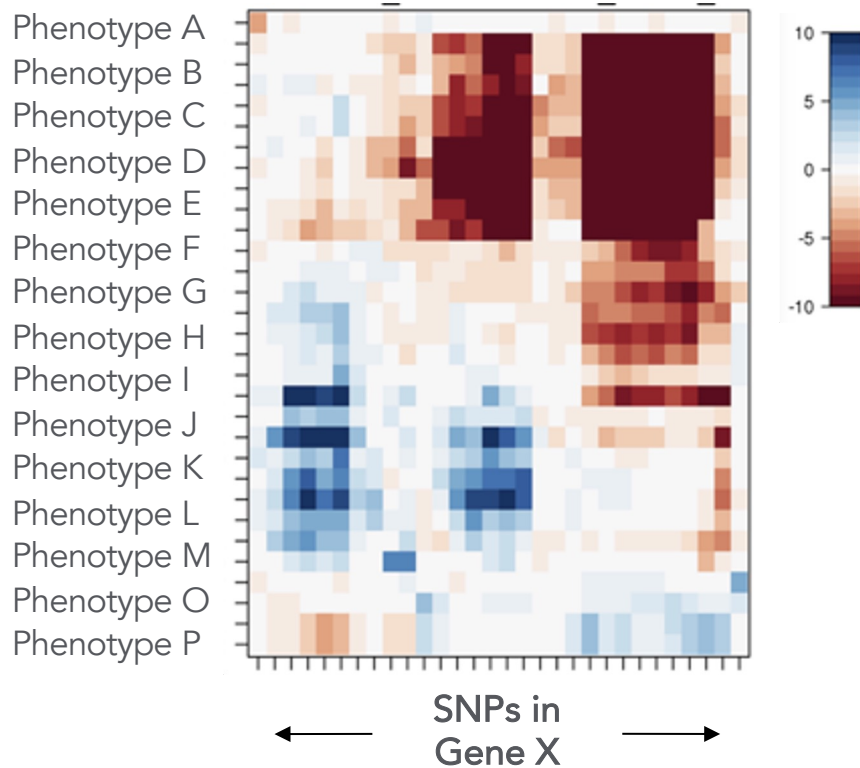
Use all the data!

Ulcerative colitis



- **Interrogate all available sources of data:** external and internal, genetics, expression, biochemical, *in vitro*, *in vivo* etc.
- **Huge sample sizes** result in narrow LD regions and sharp peaks, making fine mapping to associated eQTLs and identifying likely target much easier

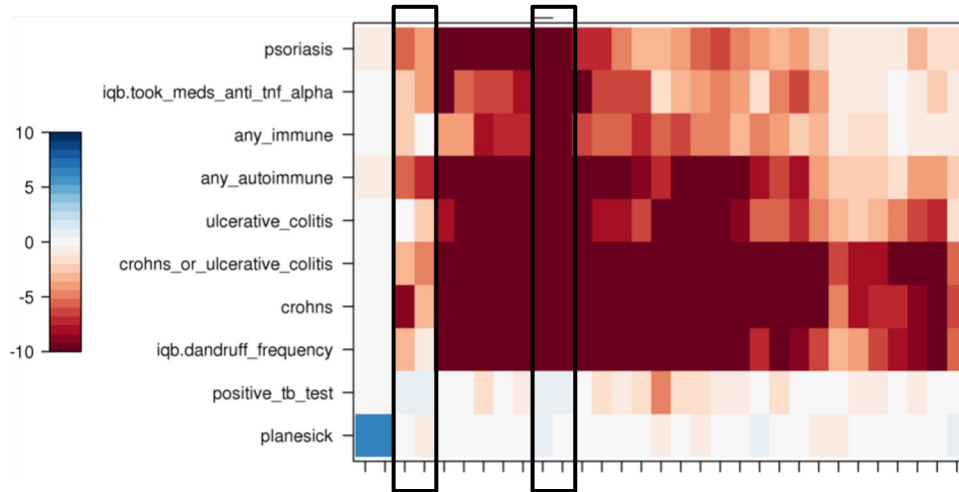
PheWAS for drug target analysis



- Clustering of related phenotypes can reinforce relevance of gene to disease biology and provide insight on potential therapeutic indications
- Opposing directionality of SNPs across different phenotypes can reinforce drug concept MOA and can reveal potential side effects of modulating gene function

IL23R PheWAS Reveals Potential Side Effects of IL-23 Blockade

SNPs identified with opposite directionality for autoimmune and infectious disease (tuberculosis) – reveals potential for increased risk of infectious disease with IL-23 antagonism.



ustekinumab drug safety information: <https://www.stelarainfo.com/>

ustekinumab, an anti-p40 antibody (IL-12/IL-23 antagonist) contains warnings for infectious disease risk, with specific warning for tuberculosis:

Serious Infections

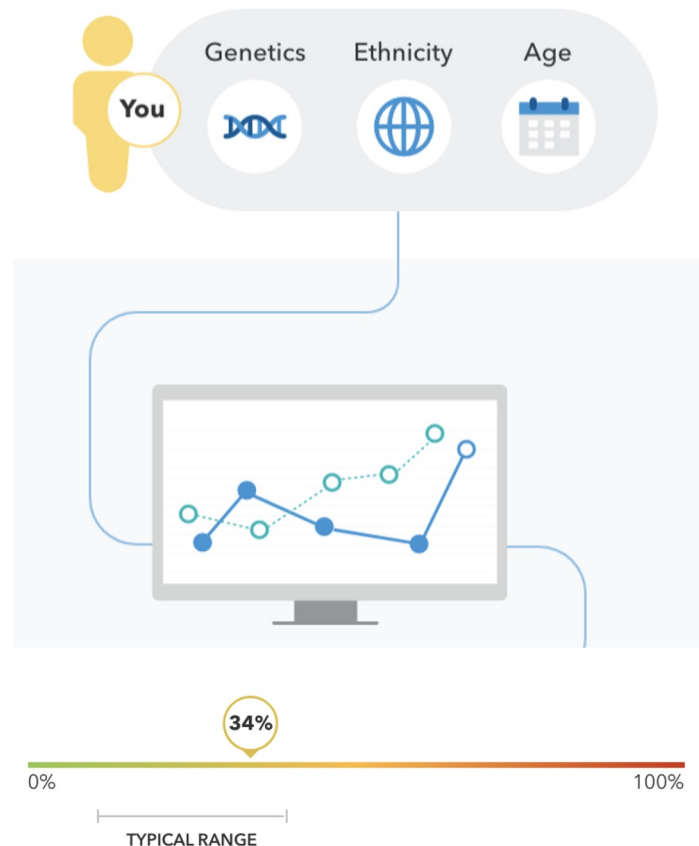
STELARA® may lower your ability to fight infections and may increase your risk of infections. While taking STELARA®, some people have serious infections, which may require hospitalization, including tuberculosis (TB), and infections caused by bacteria, fungi, or viruses.

- Your doctor should check you for TB before starting STELARA® and watch you closely for signs and symptoms of TB during treatment with STELARA®.
- If your doctor feels that you are at risk for TB, you may be treated for TB before and during treatment with STELARA®.

You should not start taking STELARA® if you have any kind of infection unless your doctor says it is okay.

Polygenic scores

A Polygenic Score (PGS) captures the cumulative effects of many genetic factors, optionally with other factors like age, sex, PCs, and ethnicity, to predict how likely a person is to have a trait or condition.



Delivering polygenic scores to consumers since 2015

2015

16 polygenic scores for common traits including cleft chin, sweet/salty taste preference, and male baldness

2017

Genetic Weight score, including gene-by-environment analysis against lifestyle factors

2018

12 additional polygenic scores for traits including wake-up time, mosquito attraction, and dandruff

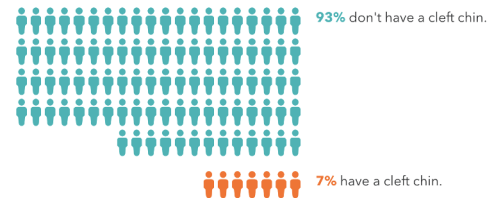
2019

Type 2 Diabetes polygenic score powered by data from more than **2.5 million research participants**



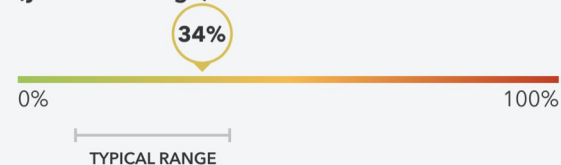
Jamie, the combination of your genetics and other factors makes you **unlikely to have a cleft chin.**

Of 23andMe research participants with results like yours:



Jamie, your genetics are associated with a **typical likelihood** of developing type 2 diabetes.

Based on data from 23andMe research participants, people of East Asian descent with genetics like yours have an estimated **34% chance** of developing type 2 diabetes at some point **between the ages of 37 (your current age) and 80.**



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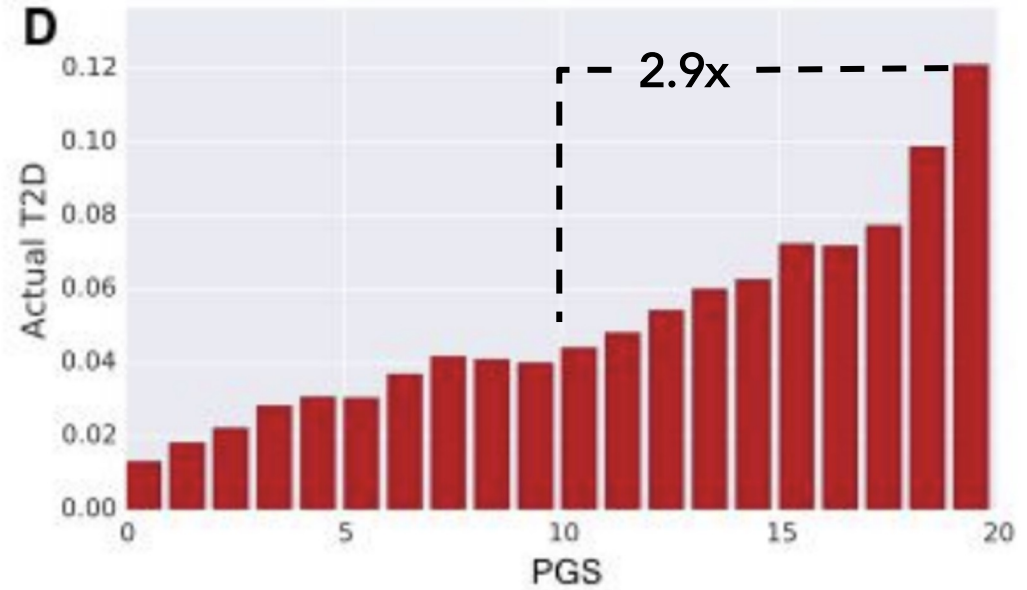
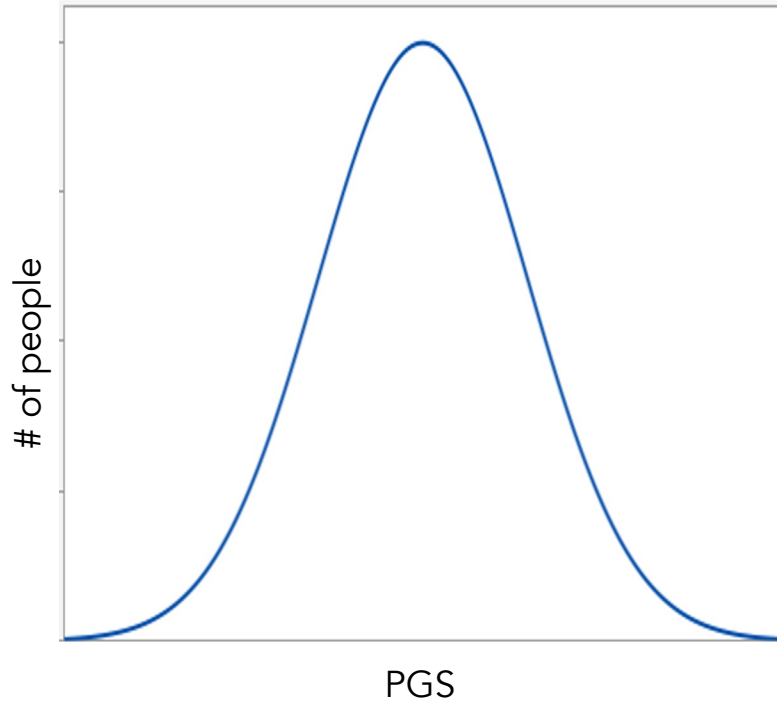
2019

Type 2 Diabetes polygenic score powered by data from more than **2.5 million research participants**

All derived from 23andMe's research platform

Our polygenic scores now routinely use data from over 1M+ consented individuals

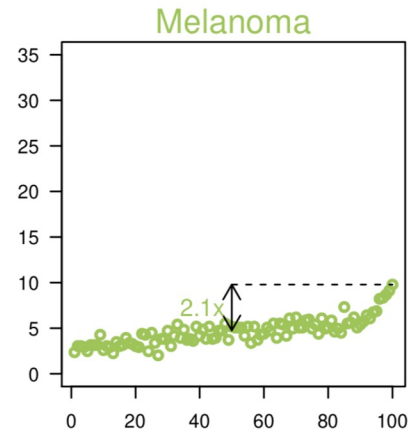
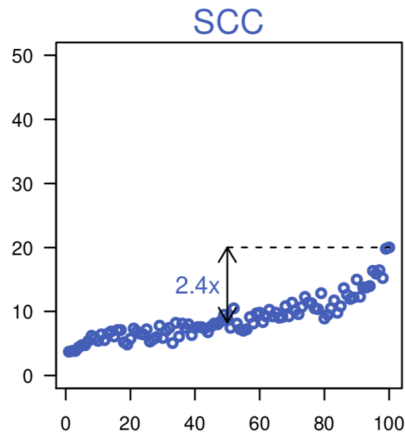
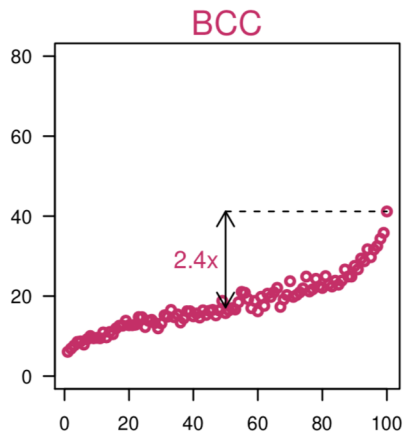
Polygenic scores for risk prediction



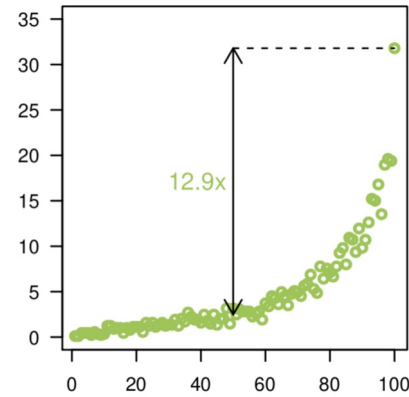
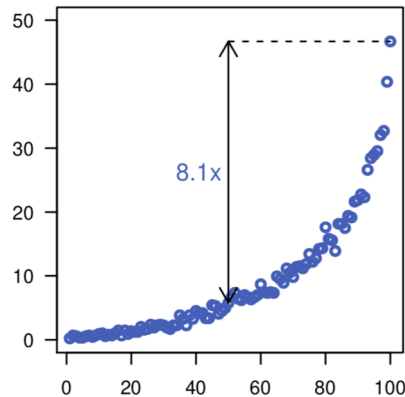
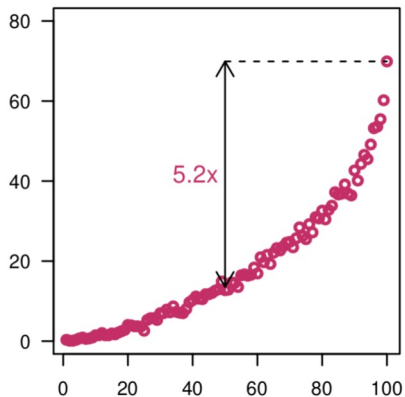
N=176,630

Including exposures improves stratification

Genetics
only



Genetics +
exposure



N=88,985

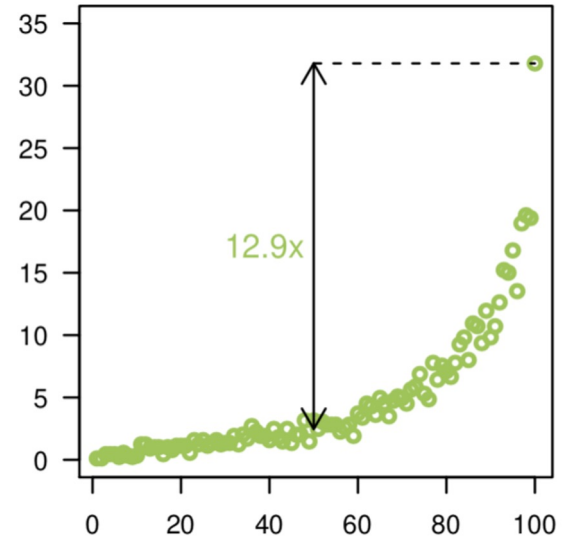
Risk stratification can identify interesting sub-cohorts

Why do some individuals > 60 yo in the top risk band with high exposure not have skin cancer?

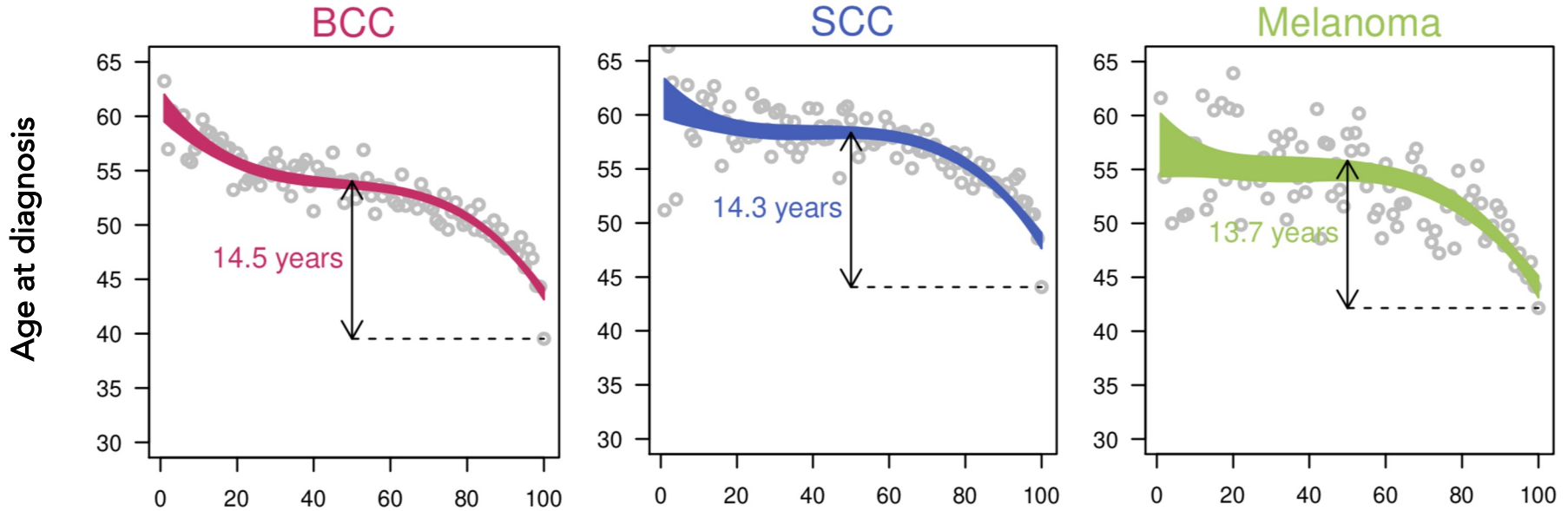
Why did some individuals in the lowest risk band with minimal exposure develop skin cancer in their 30s?

- Investigating either extreme can reveal novel biology
- Finding protective effects especially attractive for drug development

Melanoma prevalence by risk score percentile

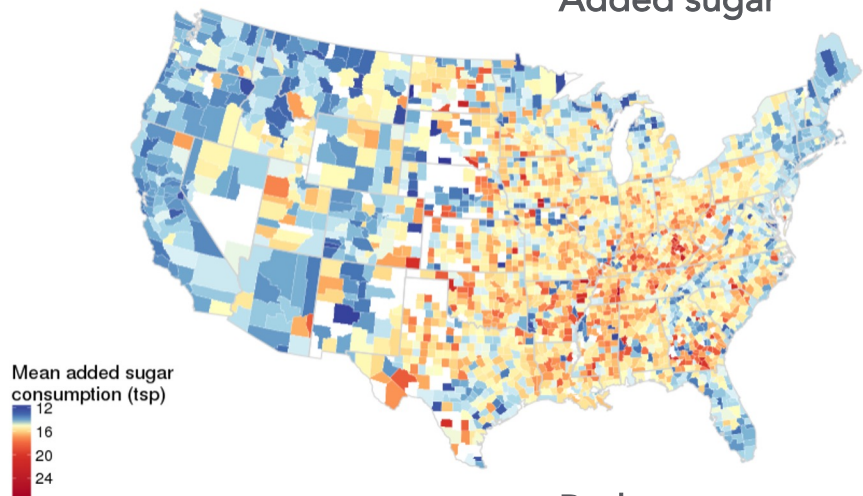


Risk stratification identifies interesting sub-cohorts

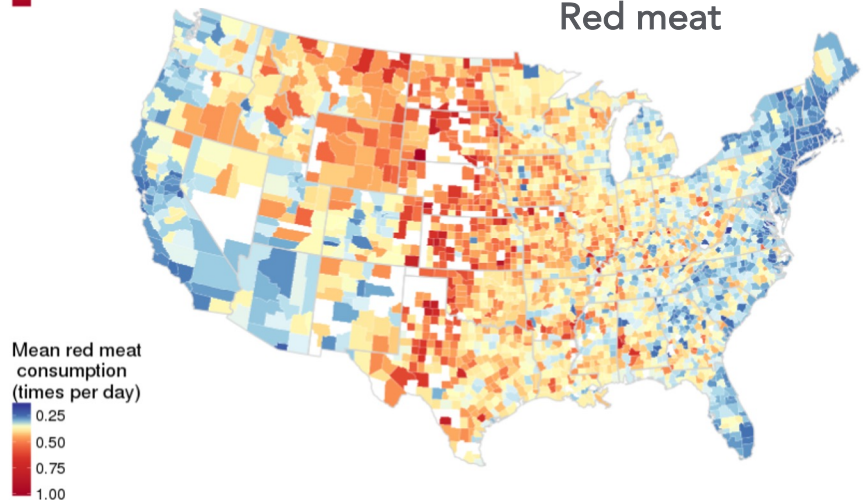


Identifying fast or early progressors, or individuals with extreme risk profiles, can enable targeted recruitment and more efficient clinical trials.

Added sugar



Red meat

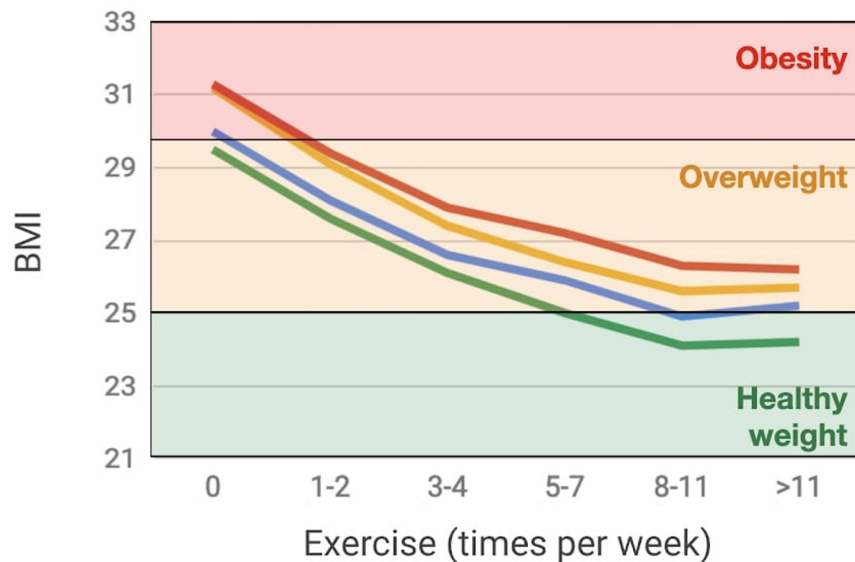
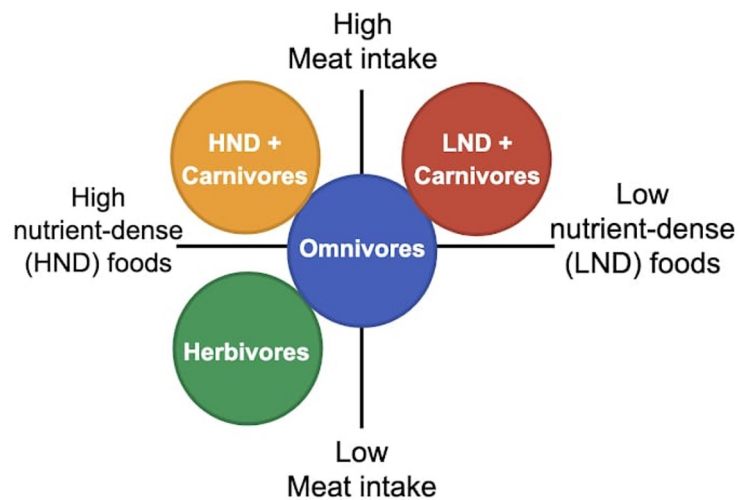


Detailed epidemiological data on behavioral phenotypes and health outcomes

Daily dietary habits in 23andMe customers by US county

Derived from a 23andMe administered NHANES-style survey, N ~ 700,000, only counties with >5 respondents shown

23andMe's data enable deeper analyses and insights into patterns of human behavior and health



Four diet types projected in the two main behavioral axes defining those types. Behaviors assessed via food frequency questionnaire in ~ 850,000 23andMe research participants.

<https://blog.23andme.com/23andme-research/23andme-explores-dietary-habit-and-health-outcomes/>

Future directions

Drug discovery:

- Leverage environment and other non-genetic data to further understand biology and drug mechanism of action and potentially develop drugs that are more effective for people with certain exposures
- Apply risk prediction to define precise cohorts that make clinical trials faster and more successful

Consumer products:

- Improve prediction through more sophisticated integration of non-genetic factors
- Provide actionable information about exposures to avoid, as long as we can reasonably establish causality
- Deliver deeper insights into relatable behaviors and their relationship to outcomes to drive individual change

Acknowledgments

23andMe research participants

23andMe Therapeutics team, including Computational Biology

23andMe Research team, including Data Collection, Statistical Genetics, and Research Ethics teams

23andMe Product, Product Science, and R&D teams

