



Data to Real World Impact: A Journey with Purpose

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Data to Real World Impact: Our End-to-End Approach

DATA	INTEGRATION	ANALYZE	REAL WORLD IMPACT								
 High quality Multi-dimensional Adequate Sample Size Secure interphase for storage and analysis 	 Discovery database and hypothesis playground Field/Disease specific use cases Cross- Population/cohort 	 Develop problem statements Use deep domain knowledge to derive meaningful insights 	 Disease Risk Prediction and Surveillance Clinical application Evidence based intervention 								
Trusted partner in the ecosystem with multiple collaborations with key stakeholders											
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DATA

Purpose: Addressing National health challenges





Use case 1 Clinical & Genomic Database Query = Clinical Impact



Genetic Counsellor App



Use case 2 Multi-platform Engagement and Data Integration for a Unified National Goal



Use case 3 Clinical and Multi-omics Data Integration to Study Biological Aging in Asians



Nature Genetics 2023 Feb;55(2):178-186 Nat Commun. 2022 Nov 5;13(1):6694 Cell 2019 Oct 17;179(3):736-749.e15 iScience. 2023 Mar 31;26(4):106546



Transcriptional signature of Frailty: Nat Commun. 2019 Dec 20;10(1):5808

Nestle

Abnormal Mitochondrial activity in Muscle wasting. J Cachexia Sarcopenia Muscle. 2023 May 4

Aging in Womb: BMC Med 2022 Jan 25;20(1)

Use case 4 Integrated Multi-Omics Database (iMOM_db) of Asian Pregnant Women

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	= * ۱	lome Statistics	FAQ	eQTL	meQTL	SNP	Transcript	CpG	
						iMOM DB			

What is included in iMOMdb?

iMOMdb consists of ethnicity based genome-wide association (GWAS), epigenome-wide association (EWAS) and transcriptome-wide association (TWAS) results. To identify potential molecular interactions between genetic and epigenetic mechanisms, Quantitative Trait Loci (QTL) information and their association with ethnicity were also made available. Most importantly, iMOMdb is open accessed with



Hong, Tan, Lim, Huang and Karnani et al. Human Mol Genet 2022



Use case 5 Diabetes Risk Prediction Modeling

50% GDM cases routinely missed in SG



> Diabetes Res Clin Pract. 2022 Feb 4;185:109237. doi: 10.1016/j.diabres.2022.109237. Online ahead of print.

Population-centric risk prediction modeling for gestational diabetes mellitus: A machine learning approach

Mukkesh Kumar ¹, Li Chen ², Karen Tan ², Li Ting Ang ³, Cindy Ho ³, Gerard Wong ², Shu E Soh ⁴, Kok Hian Tan ⁵, Jerry Kok Yen Chan ⁶, Keith M Godfrey ⁷, Shiao-Yng Chan ⁸ Mary Foong Fong Chong ⁹, John E Connolly ¹⁰, Yap Seng Chong ⁸, Johan G Eriksson ¹¹, Menaling Feng ¹². Neeria Karnani ¹³



UK NICE guidelines showed poor predictability in Singaporean women [AUC:0.60 (95% CI 0.51, 0.70)]. The noninvasive predictive model comprising of 4 non-invasive factors: mean arterial blood pressure in first trimester, age, ethnicity and previous history of GDM, greatly outperformed [AUC:0.82 (95% CI 0.71, 0.93)] the UK NICE guidelines.

Preconception Predictive Risk Modelling for GDM



MDPI

Article

Automated Machine Learning (AutoML)-Derived Preconception Predictive Risk Model to Guide Early Intervention for Gestational Diabetes Mellitus

Mukkesh Kumar ^{1,2,3}, Li Ting Ang ^{1,2}, Hang Png ^{1,2}, Maisie Ng ^{1,2}, Karen Tan ¹, See Ling Loy ^{4,5}, Kok Hian Tan ^{4,6}, Jerry Kok Yen Chan ^{4,5,7,8}, Keith M. Godfrey ⁹, Shiao-yng Chan ^{1,8}, Yap Seng Chong ^{1,8}, Johan G. Eriksson ^{1,8,10,11,†}, Mengling Feng ^{3,12,*,†} and Neerja Karnani ^{1,2,13,*,†}

GDM to T2D Risk analysis

Machine Learning Derived Prenatal Predictive Risk Model to Guide Intervention and Prevent the Progression of Gestational Diabetes Mellitus to Type 2 Diabetes. [Accepted for publication in JMIR Diabetes on 22 March 2022]

Kumar M, Ang LT, Ho C, Soh SE, Tan KH, Chan JK, Godfrey KM, Chan SY, Chong YS, Eriksson JG, Feng M, Karnani N





Federated Data Learning - Overcoming Data Sharing Obstacles

Homomorphic Encryption (HE)

Enables computation of encrypted information without need for decryption



Benefits:

Increased data security owing to HE
 Reduced regulatory hurdle for local and international cohort data integration

Example – The Cardiometabolic Health Initiative *Polygenic Risk Score calculation* Two local cohorts (PRISM & ATTRaCT) and >3000 subjects



The Science Mavericks





Kang Qi

Developing an LSTM Model for Predicting Nutritional Deficiency Diseases with Medication Administration Records

@ SICS



Jia Xu



Gong Min



Kelly Ong











Felicia Tin





THANK YOU

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