

## This Week in Genomics September 25, 2023

### AlphaFold tool pinpoints protein mutations that cause disease

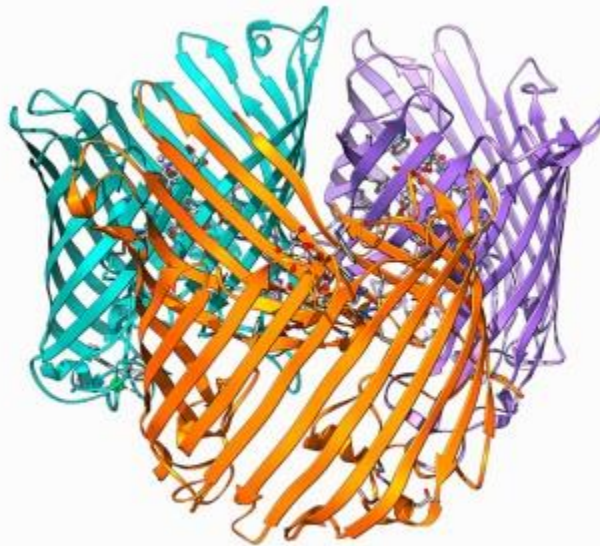
Researchers have adapted the AI network to search for genetic changes linked to ill health.

<https://www.nature.com/articles/d41586-023-02943-5>

**Google DeepMind** has wielded its revolutionary protein-structure-prediction AI in the hunt for genetic mutations that cause disease.

A new tool based on the AlphaFold network can accurately predict which mutations in proteins are likely to cause health conditions — a challenge that limits the use of genomics in healthcare.

The AI network — called AlphaMissense — is a step forward, say researchers who are developing similar tools, but not necessarily a sea change. It is one of many techniques in development that aim to help researchers, and ultimately physicians, to 'interpret' people's genomes to find the cause of a disease. But tools such as AlphaMissense — which is described in a 19 September paper in *Science*<sup>1</sup> — will need to undergo thorough testing before they are used in the clinic.



'A Pandora's box': map of protein-structure families delights scientists

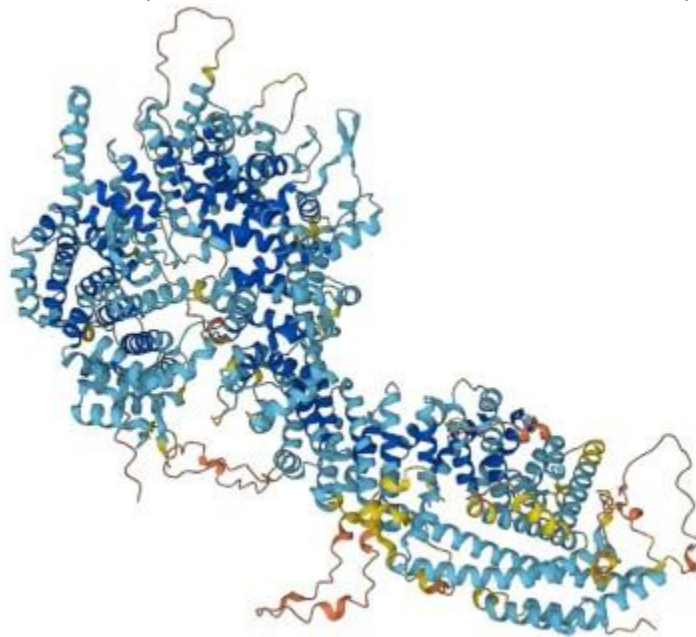
Many of the genetic mutations that directly cause a condition, such as those responsible for cystic fibrosis and sickle-cell disease, tend to change the amino acid sequence of the protein they encode. But researchers have observed only a few million of these single-letter 'missense mutations'. Of the more than 70 million possible in the human genome, only a sliver have been conclusively linked to disease, and most seem to have no ill effect on health.

So when researchers and **doctors find a missense mutation** they've never seen before, it can be **difficult to know what to make of it**. To help **interpret** such **'variants of unknown significance'**, researchers have developed dozens of different computational tools that can **predict whether a variant is likely to cause disease**. **AlphaMissense incorporates existing approaches** to the problem, which **are increasingly being addressed with machine learning**.

### Locating mutations

The **network is based on AlphaFold**, which **predicts a protein structure from an amino-acid sequence**. But instead of determining the structural effects of a mutation — an open challenge in biology — **AlphaMissense uses AlphaFold's 'intuition' about structure to identify where disease-causing mutations are likely to occur within a protein**, Pushmeet Kohli, DeepMind's vice-president of Research and a study author, said at a press briefing.

**AlphaMissense** also **incorporates a type of neural network** inspired by **large language models like ChatGPT** that has been **trained on millions of protein sequences** instead of words, called a **protein language model**. These have proven **adept at predicting protein structures and designing new proteins**. They are **useful for variant prediction** because they have **learned which sequences are plausible and which are not**, Žiga Avsec, the DeepMind research scientist who co-led the study, told journalists.



### **Foldseek gives AlphaFold protein database a rapid search tool**

**DeepMind's network** seems to **outperform other computational tools at discerning variants known to cause disease** from those that don't. It also does well at **spotting problem variants identified in laboratory experiments** that measure the **effects of thousands of mutations at once**. The researchers also **used AlphaMissense** to **create a catalogue of every possible missense mutation in the human genome**, determining that **57% are likely to be benign** and that **32% may cause disease**.

## Clinical support

***AlphaMissense is an advance*** over existing tools for predicting the effects of mutations, “***but not a gigantic leap forward***”, says Arne Elofsson, a computational biologist at the University of Stockholm.

***Its impact won't be as significant as AlphaFold***, which ***ushered in a new era in computational biology***, agrees Joseph Marsh, a computational biologist at the MRC Human Genetics Unit in Edinburgh, UK. “It’s exciting. ***It’s probably the best predictor we have right now***. But ***will it be the best predictor in two or three years?*** There’s a good chance it won’t be.”

***Computational predictions currently have a minimal role in diagnosing genetic diseases***, says Marsh, and recommendations from ***physicians’ groups say that these tools should provide only supporting evidence in linking a mutation to a disease***. ***AlphaMissense confidently classified a much larger proportion of missense mutations than have previous methods***, says Avsec. “As these ***models get better*** than I think ***people will be more inclined to trust them***.”

Yana Bromberg, a bioinformatician at Emory University in Atlanta, Georgia, emphasizes that ***tools such as AlphaMissense must be rigorously evaluated*** — using good performance metrics — ***before ever being applied in the real-world***.

For example, an exercise called the ***Critical Assessment of Genome Interpretation (CAGI)*** has ***benchmarked the performance*** of such prediction methods for years ***against experimental data that*** has not yet been released. “***It’s my worst nightmare to think of a doctor taking a prediction and running with it, as if it’s a real thing, without evaluation by entities such as CAGI***,” Bromberg adds.