# deepinfer - Bayesian approach for pre-calculating disease probabilities of genetic variants

#### 2024-05-27

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To pre-calculate the probability of every possible variant causing any set of diseases using a Bayesian approach. The goal is to integrate known priors and estimated unknowns into a cohesive Bayesian framework.

## Strategy Overview

- 1. Variant Data Collection and Annotation:
  - Variant Identification: Collect all known nucleotide variants and complex variants.
  - Variant Annotation: Use tools like VariantAnnotation to annotate each variant with functional consequences.

#### 2. Probability Estimation:

- Frequency Estimation: Estimate the probability of each variant based on true frequency in the population using databases such as gnomAD.
- Random Variant Estimation: Estimate the probability of random, novel variants.
- 3. Prior Information Incorporation:
  - **Biological Data Integration**: Integrate biological data (clinical pathogenicity, protein structure, etc.) as priors.
  - Bayesian Priors: Utilize priors from databases like ClinVar for pathogenicity and structural data from UniProt.

# 4. Bayesian Inference:

- Disease Probability Calculation: Apply Bayesian methods to update the probability of each variant causing specific diseases.
- Posterior Probability Calculation: Use the Bayes theorem to combine prior knowledge and observed data.

## 5. Tools and Packages in R:

- VariantAnnotation: For annotating genetic variants.
- gnomAD: To get population frequency of variants.
- ClinVar: To get clinical significance of variants.
- UniProt: To get protein structural information.
- brms: For Bayesian regression modeling.
- rstan: For Bayesian inference using Stan.

```
library(VariantAnnotation)
library(UniProt.ws)
library(brms)
library(rstan)
# gnomAD data
# ClinVar data
```

#### Step-by-Step Methodology

# 1. Data Collection and Annotation:

```
# Load variants
vcf <- readVcf("variants.vcf")

# Annotate variants
annotateVariants <- function(vcf) {
   return(vcf)
}
ann <- annotateVariants(vcf)</pre>
```

# 2. Probability Estimation:

```
# get frequencies
get_gnomAD_frequencies <- function(ann) {
    # Simulate frequencies for illustration
    freq <- runif(length(ann), min = 0, max = 0.01)
    return(freq)
}
estimate_random_variants <- function(ann) {
    # Simulate random variant probabilities
    random_prob <- runif(length(ann), min = 0, max = 0.001)
    return(random_prob)
}
# Frequency estimation from gnomAD
freq <- get_gnomAD_frequencies(ann)
# Random variant estimation
random_prob <- estimate_random_variants(ann)</pre>
```

### 3. Incorporate Prior Information:

```
get_clinvar_significance <- function(ann) {
    # Simulate clinical significance for illustration
    clin_significance <- sample(c("Pathogenic", "Benign", "VUS"), length(ann), replace = TRUE)
    return(clin_significance)
}
get_uniprot_data <- function(ann) {
    # Simulate UniProt data for illustration
    uniprot_data <- data.frame(
        protein_id = sample(letters, length(ann), replace = TRUE),
        structure_info = sample(c("Helix", "Sheet", "Loop"), length(ann), replace = TRUE)
    )
    return(uniprot_data)
}</pre>
```

```
# Clinical significance from ClinVar
clin_significance <- get_clinvar_significance(ann)

# Protein structure from UniProt
uniprot_data <- get_uniprot_data(ann)</pre>
```

## 4. Bayesian Inference:

```
library(brms)
library(rstan)
annotated_data <- data.frame(</pre>
 variant = sample(1:100, 100, replace = TRUE),
 gene = sample(letters, 100, replace = TRUE),
 disease = sample(c(0, 1), 100, replace = TRUE)
# Define the prior distributions
prior <- c(</pre>
  set_prior("normal(0, 1)", class = "b"),
  set_prior("cauchy(0, 2)", class = "sd")
# Fit a Bayesian model
fit <- brm(</pre>
 formula = disease ~ variant + (1|gene),
 data = annotated_data,
 family = bernoulli(),
 prior = prior,
 chains = 4, iter = 2000
)
# Extract posterior probabilities
posterior <- posterior_samples(fit)</pre>
```

# 5. Posterior Probability Calculation:

```
calculate_posterior_probabilities <- function(posterior) {
   post_prob <- apply(posterior, 2, mean)
   return(post_prob)
}

# Calculate the posterior probability for each variant-disease pair
post_prob <- calculate_posterior_probabilities(posterior)

# Output the results
write.csv(post_prob, "posterior_probabilities.csv")</pre>
```

## **Summary**

This methodology outlines a comprehensive Bayesian approach to estimate the probability of genetic variants causing specific diseases. By integrating population frequencies, biological data, and clinical annotations into a Bayesian framework, we can pre-calculate disease probabilities for known and novel variants.

We integrate of v genetic variants.	various data sourc	es to provide rol	bust estimates of d	lisease probabilities	associated with