## Bioinformatics and computational biology: a statistician's perspective

#### Shofi Andari

Departemen Statistika Institut Teknologi Sepuluh Nopember, Surabaya

Statistics Online Seminar Series #02 July 10, 2020







#### **Outline**

- A brief background
- What is bioinformatics, really?
- The central dogma in molecular biology
- The rise of omics fields and data
- A sneak peak on some databases
- Statistical challenges: data structure and analysis
- More examples on how statistics contributes in solving biological problem
  - 1 Phylogenetic tree for investigating SARS-CoV-2
  - 2 Hidden Markov model for sequence merging

Shofi Andari

## A (not so) brief background



**Shofi Andari** joined the Dept. of Statistics FSAD-ITS in 2013 as a teaching staff and is a member of *Lab. Statistika Lingkungan dan Kesehatan*.

She started to pursue her Ph.D. at Bioinformatics and Computational Biology (BCB) Program, under home department in Dept. of Statistics, Iowa State U of Science and Technology, Ames, IA, USA since Fall 2017 with a DIKTI Funded-Fulbright Foreign Program scholarship.

## A (not so) brief background cont'd



# **About Bioinformatics and Computational Biology** (BCB) Program



Figure 1: Molecular Biology Building (MBB) at Iowa State University

## BCB Program at Iowa State U: $\sim$ 20 departments

- an interdepartmental program (interdisciplinary)
- participated by  $\sim$ 20 departments in the university
  - Agronomy
  - Animal Science
  - Astronomy Physics
  - Biochemistry
  - Ecology, Evolution, and Organismal Biology
  - Chemistry
  - Biophysics and Molecular Biology

- Veterinary Pathology
- Biomedical Sciences
- Statistics
- Mathematics
- Computer Science
- Electrical and Computer Engineering
- more on https://www. bcb.iastate.edu/

## BCB Program at Iowa State U: the people (some of them)



Dr. Friedberg (DOGE) Dept. of VMPM



Dr. Dan Nettleton Dept. of STAT



Dr. Karin Dorman Dept of STAT / GDCR



Dr. Pena Liu Dept. of STAT



Dr. Heike Hoffman Dept. of STAT



Dr. Phillip Dixon Dept. of STAT



Dr. Yandeau-Nelson Dept. of GDCB





Dept. of CS



Dept. of GDCB



Dept. of EEOB







Dept. of BBMB



















Dr. Steven Rodermel Dept. of GDCB



Dr. Travesset Dept. of PHYS



Dr. Mary Grenlee Dept. of BIOMED

Dr. Guang Song Dept. of CS













Dr. Jack Dekkers Dept. of AN S

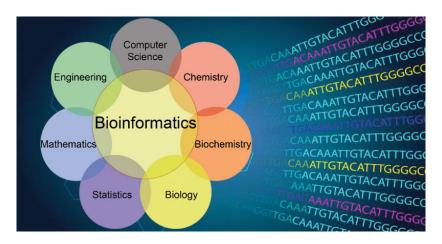
Dr. Crystal Lu Dr. Surai Khotari Dept. of EEOB Dept. of ECpE

Dept. of CS Dept. of CHEM

Dr. Jack Lutz Dept. of MATH / CS

Dept. of GDCB

#### What is bioinformatics, really?



**Figure 2:** Disciplines building bioinformatics (but not limited to these fields!)

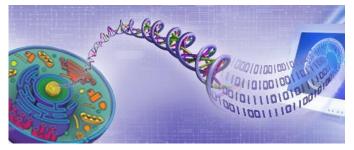
Shofi Andari

#### What is bioinformatics, really? cont'd

 ".. an interdisciplinary field involving computational biologists, computer scientists, mathematical modelers, systems biologists, and statisticians exploring different facets of the data ranging from storing, retrieving, organizing and subsequent analysis of biological data." (Morris, 2017)

#### What is bioinformatics, really? cont'd

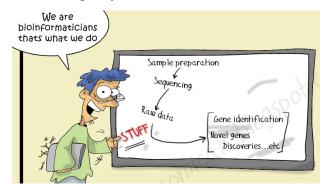
- "... an interdisciplinary field involving computational biologists, computer scientists, mathematical modelers, systems biologists, and statisticians exploring different facets of the data ranging from storing, retrieving, organizing and subsequent analysis of biological data." (Morris, 2017)
- ".. is the data science of biology"



## What is bioinformatics, really? cont'd

#### Computational biology

- ".. is translating and framing biological problems into computational problems (i.e., algorithms, math model derivations)"
- Often times, bionformatics and computational biology are used interchangeably.



#### The central dogma in molecular biology

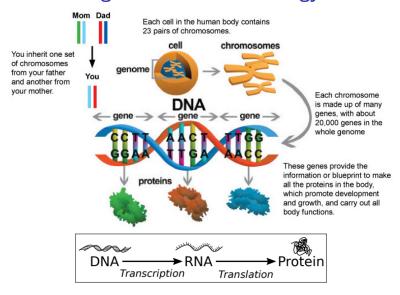


Figure 3: The central dogma of genes to protein (Ritz2018)

#### The central dogma in molecular biology cont'd

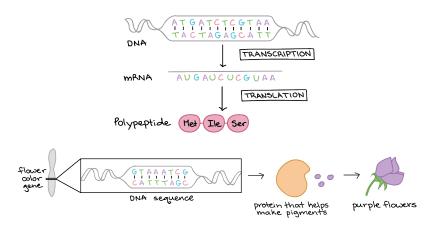
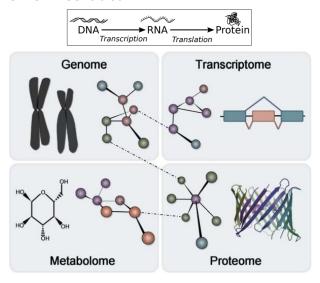


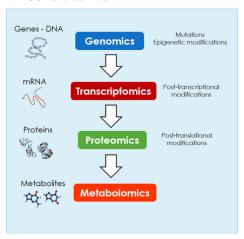
Figure 4: Mendel's flower color gene provides instructions for a protein that helps make colored molecules (pigments) in flower petals (Hellens (2010), Reece (2011), www.khanacademy.org)

#### The rise of omics data



**Figure 5:** Different networks emerging from the central dogma (Franklin2011)

#### The rise of omics data cont'd



**Figure 6:** Four main omics fields can be distinguished: genomics (DNA), transcriptomics (mRNA), proteomics (proteins) and metabolomics (metabolites)

(http://ch4eo.info/research/omics/)

#### The rise of omics data: the history

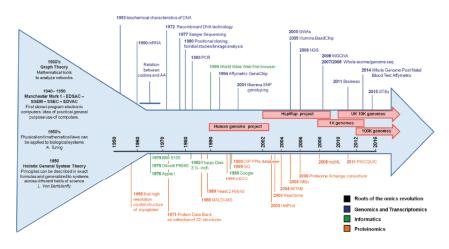
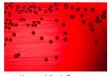


Figure 7: Progressive advance in omic-sciences (Manzoni, 2018)

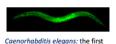
#### The rise of omics data: walk of fame



Haemophilus influenzae



Elaeis guineensis (oil palm): 10 years and 50 scientists to sequence



worm and animal to have its whole genome sequenced



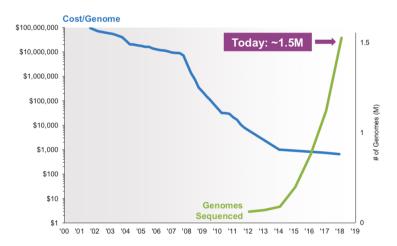


Arabidopsis thaliana: the first plant genome sequenced.

Human Genome Project (HGP)
Planned since 1984, launched in 1990,
completed in 2003.
Remains the biggest collaborative
biological project on earth.
Done in various research centers in the
US, UK, Japan, France, Germany, China,

Figure 8: Whole-genome sequencing (WGS) historic lane, from bacteria to human (en.wikipedia.org)

#### The rise of omics data: the falling cost



**Figure 9:** Advances in the field of genomics have led to substantial reductions in the cost of genome sequencing

(https://www.forbes.com/sites/kenberman/)

#### The rise of omics data: the trend

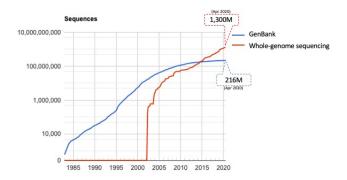


Figure 10: Volume of DNA information in GenBank (https://www.ncbi.nlm.nih.gov/genbank/statistics/)

## **Statistical challenges**



#### Statistical challenges: data structure and analysis

- Where to access the (benchmark) data sets (if there is any)?
- Let's talk about genomics data: DNA and/or RNA sequences
- Statistical bioinformatics

## A sneak peak on some databases

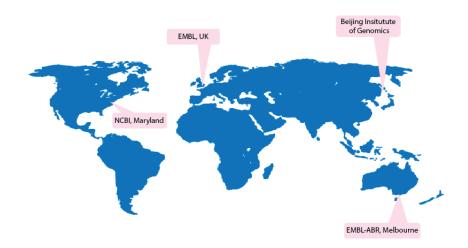


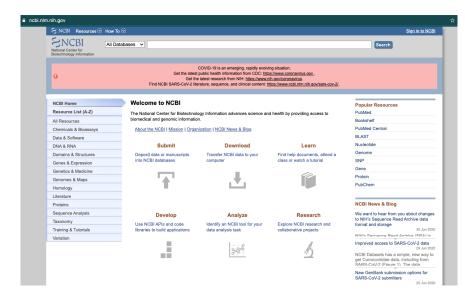
Figure 11: A few of bioinformatics labs on the globe

## A sneak peak on NCBI

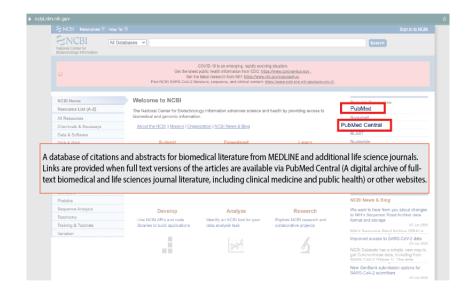
- National Center for Biotechnology Information (NCBI) website: https://www.ncbi.nlm.nih.gov/
- formed in 1988 as a complement to the activities of the National Institutes of Health (NIH) and the National Library of Medicine (NLM)
- paramount of bioinformatics data bank and tools



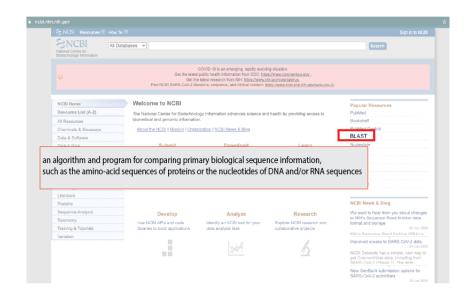
#### NCBI website cont'd



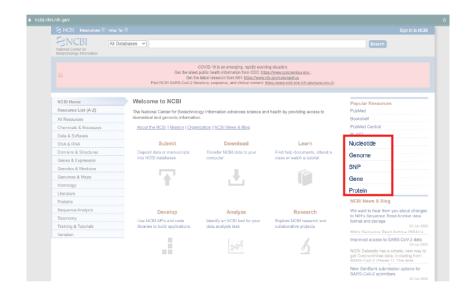
#### NCBI website: journal papers



#### NCBI website: BLAST



#### NCBI website: databases



## A sneak peak on EMBL - EBI



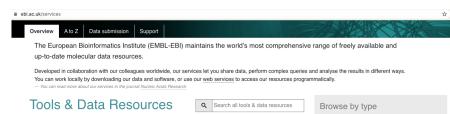
#### A sneak peak on EMBL - EBI cont'd

M InterPro

**PDBe** 

The European resource for the collection. organisation and dissemination of 3D structural data (from PDB

and EMDB) on biological macromolecules and their complexes.





#### Programmatic access

EMBL-EBI web services allow you to query our large biological data resources programmatically, so that you can develop data analysis pipelines or integrate public data with your own applications. The Web Services

protein signatures.

InterProScan searches sequences against InterPro's predictive

Protein feature detection Sequence motif recognition

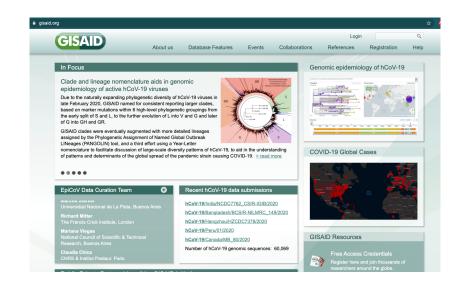
#### A sneak peak on GISAID

- .. a global science initiative and primary source for genomic data of influenza viruses and the novel coronavirus 2019.
- HQ: Munich, Germany
- Website: https://www.gisaid.org/

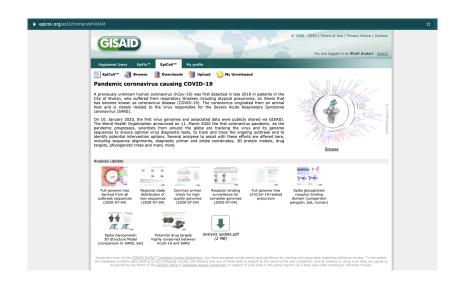


(Elbe, 2016, doi:10.1002/gch2.1018)

#### A sneak peak on GISAID cont'd



#### A sneak peak on GISAID cont'd



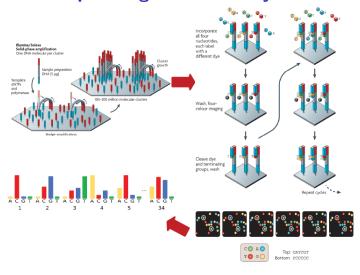
## DNA/RNA sequencing: high-throughput sequencing



Figure 12: Illumina MiSeq

(https://scientificservices.eu/item/illumina-miseq-next-generation-sequencer/5538)

#### DNA/RNA sequencing: how do they do it?



**Figure 13:** DNA sequencing: amplification, cycle, and base-calling (Whiteford, 2009; Metzker, 2009)

## DNA sequences data format: FASTQ format

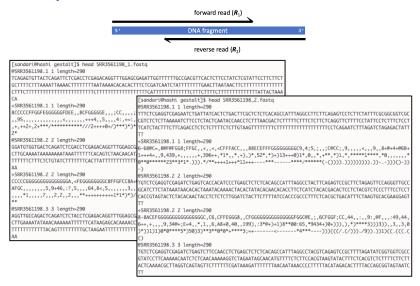


Figure 14: Paired-end sequencing results: FASTQ files

#### DNA sequences data format cont'd

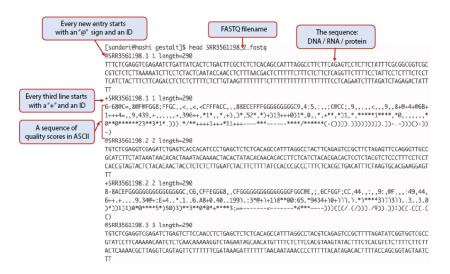


Figure 15: An example of FASTQ files

Shofi Andari

## DNA sequences data format: FASTA format

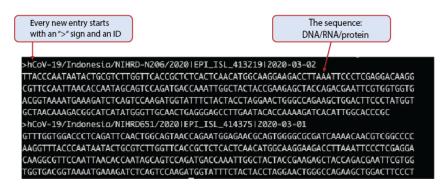


Figure 16: An example of FASTA files

 Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration
- High volume data

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration
- High volume data
- Uncertainty (i.e., due to genetic variants)

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration
- High volume data
- Uncertainty (i.e., due to genetic variants)
- Data curation

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration
- High volume data
- Uncertainty (i.e., due to genetic variants)
- Data curation
- Data sharing

- Data heterogeneity (various types of data is hard to create a comprehensive view of studies ⇒ integrative methods are necessary (Ren, 2015))
- Large scale data integration
- High volume data
- Uncertainty (i.e., due to genetic variants)
- Data curation
- Data sharing
- Dynamic and subject to change

## Statistical challenges: statisticians' perspective

- The advent of high-throughput multi-platform genomics technologies ⇒ highly structured big data
- Bioinformatics is necessarily interdisciplinary in nature: clinical, biological, computational, data management, mathematical modeling, and statistical knowledge and skills
- One of the key attributes that sets statisticians apart from other quantitative scientists is their understanding of variability and uncertainty quantification

(Morris, 2017)

## Statistical challenges: statisticians' perspective cont'd

### With the basis of statisticians as data scientists:

- sampling design decisions
- multi-step processing algorithms
- reductionistic feature extraction
- inferential reasoning
- design algorithms to search high-dimensional spaces
- build predictive models while obtaining accurate measures of their predictive accuracy

(Morris, 2017)

# What does it take to be a statistical bioinformatician?

### The basic!

- collecting data and experiment design
- descriptive statistics and data visualization
- randomness and probability concept
- estimation: point and interval, MoM, MLE, exponential family, RBT, UMVUE, FI, CRLB
- hypothesis testing: type I and II errors, test statistics, power and sample calculations, NPL, LRT
- simulations

# What does it take to be a statistical bioinformatician? cont'd

 Gene mapping and association studies: clustering, maximum likelihood estimations, QTL/eQTL, linear models



#### RESEARCH ARTICLE

Fast and flexible linear mixed models for genome-wide genetics

Daniel E. Runcieo1\*, Lorin Crawfordo2

1 Department of Plant Sciences, University of California Davis, Davis, California, United States of America, 2 Department of Biostatistics, Brown University, Providence, Rhode Island, United States of America



### COPEN ACCESS

Citation: Runcle DE, Crawford L (2019) Fast and flexible linear mixed models for genome-wide genetics. PLoS Genet 15(2): e1007978. https://doi. org/10.1371/journal.pgen.1007978

Editor: Michael P. Epstein, Emory University, UNITED STATES

Received: September 7, 2018
Accepted: January 21, 2019

#### Abstract

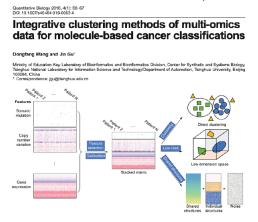
Linear mixed effect models are powerful tools used to account for population structure in genome-wide association studies (WMSAs) and estimate the genetic architecture of complex traits. However, fully-specified models are computationally demanding and common simplifications often lead to reduced power of biased inference. We describe Gild—DMM (Interplatible) and ordering the described of the properties linear models that account for multiple sources of heterogeneity, such as additive and non-additive genetic variance, spatial heterogeneity, and genotype-environment interactions. Gild—DMM can outline to a genome wide scele in a fraction of the time compared to existing general-purpose methods. We apply Gild—DMM to two types of quantitative genetic variance while scanning for CTIL, and the second aims to identify gene expression traits affected by non-additive genetic variance while scanning for CTIL, and the second aims to identify gene expression traits affected by non-additive genetic variance values to new blood of the training training that the other opening the scanning for CTIL, and the second aims to identify gene expression traits affected by non-additive genetic variance values to new discoveries.

<sup>\*</sup> deruncie@ucdavis.edu

### What else?

 Gene regulatory network & multi-omics data integration: gene expression 

heatmap, feature extraction (PCA, PLS)



### What else? cont'd

 High-throughput data processing: genome assembly, alignment

### **BMC Bioinformatics**



Methodology article

Open Access

## A statistical score for assessing the quality of multiple sequence alignments

Virpi Ahola\*1,2, Tero Aittokallio3,6, Mauno Vihinen4,5 and Esa Uusipaikka2

Address: Biotechnology and Food Research, MTT Agrifood Research Finland, Jokioinen, Finland, <sup>2</sup>Department of Statistics, University of Turku, Turku, Finland, <sup>3</sup>Department of Mathematics, University of Turku, Turku, Finland, <sup>4</sup>Institute of Medical Technology, University of Tampere, Tampere, Finland, <sup>3</sup>Research Unit, Tampere University Hospital, Tampere, Finland and <sup>4</sup>Systems Biology Unit, Institut Pasteur, Paris, France

Email: Virpi Ahola\* - virpi.ahola@mtt.fi; Tero Aittokallio - tero.aittokallio@utu.fi; Mauno Vihinen - mauno.vihinen@uta.fi; Esa Uusipaikka - esa.uusipaikka@utu.fi

Published: 03 November 2006

Received: 11 April 2006 Accepted: 03 November 2006

BMC Bioinformatics 2006, 7:484 doi:10.1186/1471-2105-7-484

This article is available from: http://www.biomedcentral.com/1471-2105/7/484

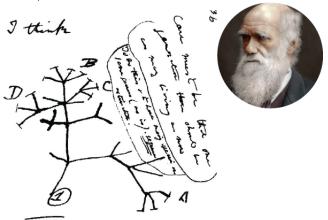
© 2006 Ahola et al; licensee BioMed Central Ltd.

This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/2.0), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

<sup>\*</sup> Corresponding author

### What else? cont'd

Evolutionary genomics: clustering concept, Bayesian framework, CTMC



Charles Darwin's sketches is considered to be the first metaphor of a tree to represent evolutionary relationships. (Image source: Wikimedia Commons.)

# What does it take to be a statistical bioinformatician? cont'd

### **Softwares:**

- R, Python
- Web resources

### **Further readings:**

- Handbook of Statistical Genomics Balding, 2019
- A Guide to QTL Mapping with R/qtl Broman, 2009
- Statistical Contributions to Bioinformatics: Design, Modeling, Structure Learning, and Integration - Morris, 2017
- Orchestrating high-throughput genomic analysis with Bioconductor - Huber, 2014

# More examples on how statistics contributes in solving biological problem

- 1 Phylogenetics tree for tracing the origin of SARS-CoV-2
- 2 Paired-end sequences alignment via pair-hidden Markov model

## Phylogenetics tree for tracing the origin of SARS-CoV-2



**Figure 17:** SARS-CoV-2 has a shape like a dandelion, but ... (*Image source: CDC/Alissa Eckert & Dan Higgins*)

Shofi Andari

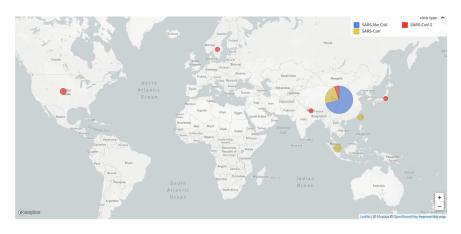


Figure 18: Covid-19's spread around the globe (https://nextstrain.org/)

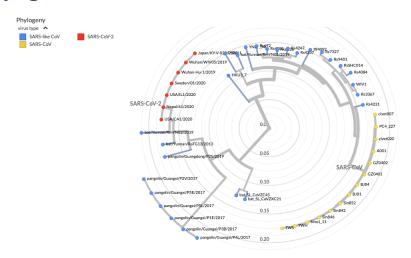


Figure 19: Covid-19's spread around the globe (https://nextstrain.org/)

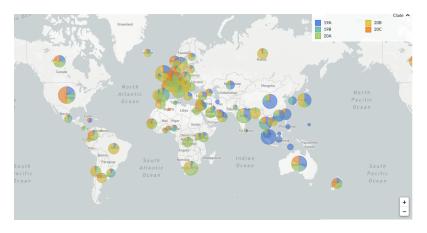


Figure 20: Covid-19's spread around the globe (https://nextstrain.org/)

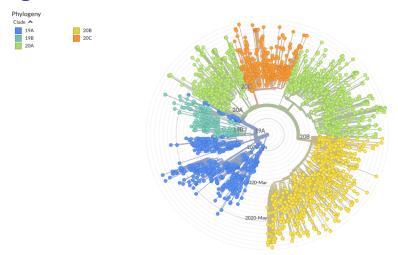
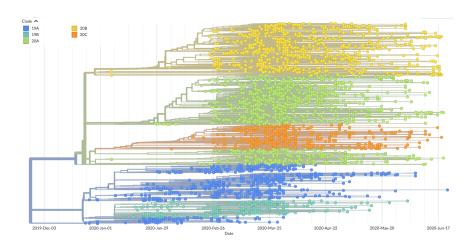
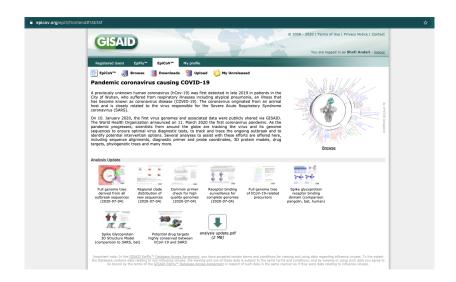


Figure 21: Covid-19's spread around the globe : radial phylogenetic tree (https://nextstrain.org/)

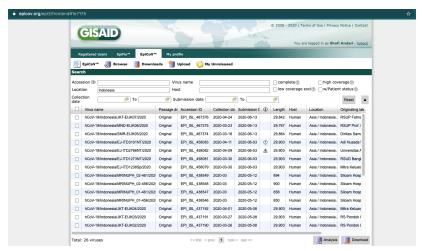


**Figure 22:** Covid-19's spread around the globe: rectangular phylogenetic tree (https://nextstrain.org/)

Revisiting GISAID home page:



Samples from Indonesia (26 submissions)

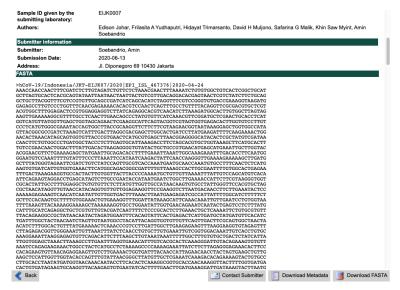


4 labs submitting SARS-CoV-2 from Indonesia: Eijkman Institute, Litbangkes RI, Mochtar Riady Institute (UPH), and ITD UNAIR.

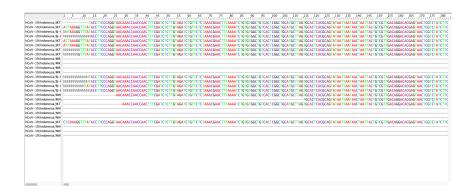
### Details of the first row:

Virus detail	
Virus name:	hCoV-19/Indonesia/JKT-EIJK07/2020
Accession ID:	EPI_ISL_467376
Type:	betacoronavirus
Lineage (GISAID Clade):	B (L)
Passage details/history:	Original
Sample information	
Collection date:	2020-04-24
Location:	Asia / Indonesia / Jakarta
Host:	Human
Additional location information:	
Gender:	Male
Patient age:	74
Patient status:	Hospitalized
Specimen source:	Nasopharyngeal and Oro-pharyngeal swab
Additional host information:	
Outbreak:	
Last vaccinated:	
Treatment:	
Sequencing technology:	Illumina MiSeq
Assembly method:	Bowtie2 + SPAdes
Coverage:	860x
Comment:	
Institute information	
Originating lab:	RSUP Fatmawati
Address:	Jl. RS. Fatmawati Raya No.4,Kota Jakarta Selatan, Daerah Khusus Ibukota Jakarta 12430, Indonesia
Sample ID given by the sample provider:	
Submitting lab:	Eijkman Institute for Molecular Biology, Ministry of Research and Technology/National Agency for Research and
Submitting lab.	Innovation

### Continuing the details of the first row:



### Multiple sequence alignment:



Choosing model and distance matrix:

- Jukes-Cantor (JC69): equal base frequencies, all substitutions equally likely
- Felsentein (F81): variable base frequencies, all substitutions equally likely
- Hasegawa-Kishino-Yano (HKY): variable base frequencies, one transition rate and one transversion rate
- Kimura-2params (K80): equal base frequencies, one transition rate and one transversion rate
- General time reversible (GTR): variable base frequencies, symmetrical substitution matrix

More on substitution models: http://evomics.org/resources/substitution-models/nucleotide-substitution-models/

### Choosing model and distance matrix:

- Jukes-Cantor (JC69): equal base frequencies, all substitutions equally likely
- Felsentein (F81): variable base frequencies, all substitutions equally likely
- Hasegawa-Kishino-Yano (HKY): variable base frequencies, one transition rate and one transversion rate
- Kimura-2params (K80): equal base frequencies, one transition rate and one transversion rate
- General time reversible (GTR): variable base frequencies, symmetrical substitution matrix
- + Gamma dist (G) / proportion of invariable sites (I): describing rate variation among sites in a sequence.

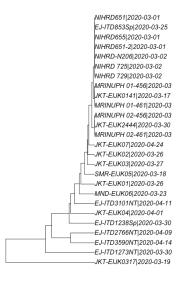
Choosing model and distance matrix on R (packages: phylogram, phangorn, seqinr)

```
> modelTest(cov_phyDat)
negative edges length changed to 0!
[1] "JC+I"
Γ17 "JC+G"
Γ17 "JC+G+I"
Γ17 "F81+I"
Γ17 "F81+G"
Γ17 "F81+G+I"
Γ17 "K80+I"
Γ17 "K80+G"
Γ17 "K80+G+I"
[1] "HKY+I"
[1] "HKY+G"
[1] "HKY+G+I"
[1] "SYM+I"
Γ17 "SYM+G"
Γ17 "SYM+G+I"
Γ17 "GTR+I"
Γ17 "GTR+G"
Γ17 "GTR+G+I"
```

### Choosing model and distance matrix on R

```
Model df
                 loaLik
                             AIC
                                         AICw
                                                  AICc
                                                              AICcw
1
        JC 49 -42308.54 84715.08 0.000000e+00 84715.25 0.000000e+00 85122.10
      JC+I 50 -42303.53 84707.06 0.000000e+00 84707.23 0.000000e+00 85122.38
     JC+G 50 -42306.04 84712.09 0.000000e+00 84712.26 0.000000e+00 85127.41
    JC+G+I 51 -42303.54 84709.08 0.000000e+00 84709.26 0.000000e+00 85132.70
       F81 52 -41406.31 82916.63 4.408795e-05 82916.81 4.504991e-05 83348.56
    F81+I 53 -41401.48 82908.95 2.049250e-03 82909.14 2.086538e-03 83349.19
    F81+G 53 -41403.83 82913.66 1.940636e-04 82913.86 1.975948e-04 83353.90
  F81+G+I 54 -41401.49 82910.97 7.460532e-04 82911.17 7.568843e-04 83359.52
       K80 50 -42304.75 84709.51 0.000000e+00 84709.68 0.000000e+00 85124.83
    K80+I 51 -42299.67 84701.34 0.000000e+00 84701.52 0.000000e+00 85124.97
     K80+G 51 -42302.25 84706.50 0.000000e+00 84706.68 0.000000e+00 85130.12
12 K80+G+I 52 -42299.68 84703.37 0.000000e+00 84703.55 0.000000e+00 85135.30
13
       HKY 53 -41401.76 82909.53 1.536213e-03 82909.72 1.564166e-03 83349.77
    HKY+I 54 -41396.92 82901.85 7.137944e-02 82902.05 7.241571e-02 83350.39
    HKY+G 54 -41399.28 82906.56 6.762151e-03 82906.76 6.860322e-03 83355.11
16 HKY+G+I 55 -41396.94 82903.87 2.592739e-02 82904.08 2.620702e-02 83360.73
       SYM 54 -42296.71 84701.42 0.000000e+00 84701.61 0.000000e+00 85149.96
    SYM+I 55 -42290.76 84691.52 0.000000e+00 84691.73 0.000000e+00 85148.37
    SYM+G 55 -42294.14 84698.29 0.000000e+00 84698.50 0.000000e+00 85155.14
20 SYM+G+T 56 -42290.79 84693.58 0.000000e+00 84693.79 0.000000e+00 85158.74
       GTR 57 -41395.94 82905.89 9.466819e-03 82906.11 9.496713e-03 83379.35
    GTR+I 58 -41390.77 82897.54 6.167872e-01 82897.77 6.163339e-01 83379.31
    GTR+G 58 -41393.43 82902.87 4.292927e-02 82903.10 4.289772e-02 83384.64
24 GTR+G+T 59 -41390 79 82899 58 2 221780e-01 82899 82 2 211384e-01 83389 66
```

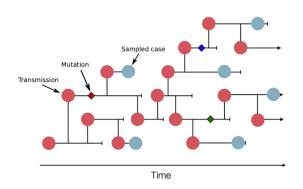
The tree:



# Phylogenetics tree for SARS-CoV-2: the tree, the meaning

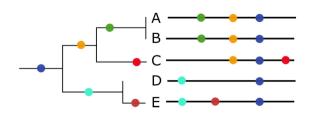


A rooted tree with 3 tips and 2 nodes (Hall & Colijin, 2019: doi:10.1093/molbev/msz058)



https://nextstrain.org/help/general/how-to-read-a-tree

# Phylogenetics tree for SARS-CoV-2: the tree, the meaning cont'd



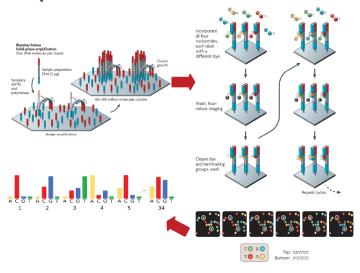
https://nextstrain.org/help/general/how-to-read-a-tree

Studying the variability among the sequences (e.g., due to random mutations)  $\Rightarrow$ 

- tracking the spread of the pathogen,
- learning its transmission routes and dynamics.

# Paired-end sequences merging via pair-Hidden markov model

#### Illumina's paired-end reads



**Figure 23:** DNA sequencing: amplification, cycle, and base-calling (Whiteford, 2009; Metzker, 2009)

#### Illumina's paired-end reads cont'd

All Illumina NGS systems are capable of paired-end sequencing

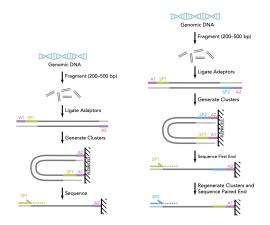


Figure 24: Illumina's single-end vs. paired-end sequencing, ©Illumina

# Merging PE reads: why it matters?

 Merging PE reads can substantially improve various subsequent bioinformatics processes, including genome assembly, binning, mapping, annotation, and clustering for taxonomic analysis (Bushnell, 2017).

# Merging PE reads: why it matters?

- Merging PE reads can substantially improve various subsequent bioinformatics processes, including genome assembly, binning, mapping, annotation, and clustering for taxonomic analysis (Bushnell, 2017).
- Existing tools:
  - SHERA (Rodrigue et al. 2010)
  - FLASH (Magoc and Salzberg 2011)
  - COPE (Liu et al. 2012)
  - PANDAseq (Masella et al. 2012)
  - PEAR (Zhang et al. 2014)
  - AdapterRemoval v2 (Schubert, Lindgreen, and Orlando 2016)
  - MeFiT (Parikh et al. 2016)
  - NGmerge (Gaspar 2018)

All of these methods either ignore the quality scores or assume all nucleotides are equally likely.

# Merging PE reads: why it matters?

- Merging PE reads can substantially improve various subsequent bioinformatics processes, including genome assembly, binning, mapping, annotation, and clustering for taxonomic analysis (Bushnell, 2017).
- Existing tools:
  - SHERA (Rodrigue et al. 2010)
  - FLASH (Magoc and Salzberg 2011)
  - COPE (Liu et al. 2012)
  - PANDAseq (Masella et al. 2012)
  - PEAR (Zhang et al. 2014)
  - AdapterRemoval v2 (Schubert, Lindgreen, and Orlando 2016)
  - MeFiT (Parikh et al. 2016)
  - NGmerge (Gaspar 2018)

All of these methods either ignore the quality scores or assume all nucleotides are equally likely.

 What we are seeking: a merging tool that could provide accurate merged sequences (and fast!)

#### Illumina's paired-end reads: finding overlap

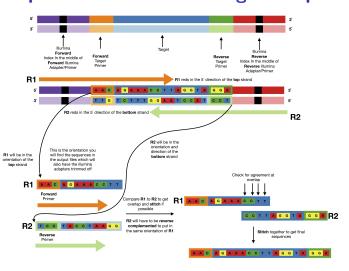


Figure 25: Illumina paired-end sequencing (Source: https://seekdeep.brown.edu/illumina\_paired\_info.html)

#### Obtaining a final (merged) sequence via alignment

forward read (R <sub>1</sub> )	С	Α	Т	Т	G	Α	С	Α
Q scores ( <b>q</b> <sub>1</sub> )	32	34	20	20	28	16	14	10
reverse read (R <sub>2</sub> )	Α	Α	Т	G	Т	С	Т	Α
Q scores ( <b>q</b> <sub>2</sub> )	40	38	20	12	8	4	5	2

Quantifying Q-scores: https:

//support.illumina.com/help/BaseSpace\_OLH\_009008/Content/
Source/Informatics/BS/QualityScoreEncoding\_swBS.htm

Obtaining a final (merged) sequence via alignment

<b>R</b> <sub>1</sub>	С	Α	Т	Т	G	Α	С	Α		
$\boldsymbol{q}_1$	32	34	20	20	28	16	14	10		
<b>W</b> <sub>2</sub>			Т	Α	G	Α	С	Α	Т	Т
$\boldsymbol{q}_2$			2	5	4	8	12	20	38	40

Final (consensus)	С	Α	Т	Т	
Posterior Q score	32	34	22	16	

G 35 28 30 34 38 40

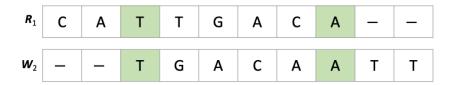
Should consider gaps in the alignment to represent insertions/deletions (indels).

The thing about sequence alignment...

forward read ( $R_1$ )	С	Α	Т	Т	G	Α	С	Α	
Q scores ( $q_1$ )	32	34	20	20	28	16	14	10	
reverse read (R <sub>2</sub> )	Α	Α	Т	Т	G	Т	С	Α	
Q scores ( <b>q</b> <sub>2</sub> )	40	38	24	10	8	5	6	2	

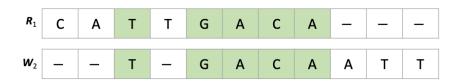
Suppose the reverse read does not match as good as our previous illustration.

The thing about sequence alignment...



Is it a good alignment?

The thing about sequence alignment...

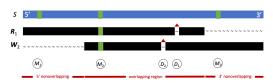


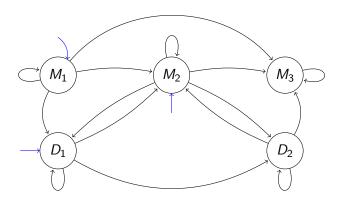
How about this one?

We introduce a gap after the first match.

It can be "costly", but now we have more matches.

# PE merging via pair-HMM cont'd





# Merging PE reads: pair-HMM approach

The blue fragment here is the reference.

Sadly, we do not always have the reference (genome). In fact, these FASTQ files are going to be used to rebuild the genome (genome assembly).

Nucleotide pairing in  $\mathbf{R}_{1j}$  and  $\mathbf{W}_{2j} =$  observable state Nucleotide pairing in the alignment = hidden state  $\Rightarrow$  pair-hidden Markov model (Rabiner, 1989; Durbin, 1998).

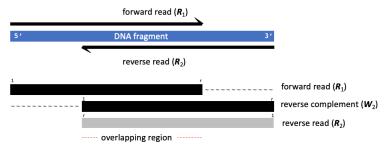


Figure 26: Alignment between forward and reverse read

Courtesy to Prof. Mardia (Dept. of Statistics, Univeristy of Leeds, UK) at the 2006 LASR Conference



"Statisticians need to be more open, more ready to learn *molecular biology*, more computationally aware, more ready to understand databanks. ...

But above all, we always need great scientist friends!! This all is a part of solving great questions in life sciences of taming the nature and immortality, etc!"

iscb.org/iscbacademy-webinars



#### ISCBacademy Webinar Series

Upcoming Webinars

Upcoming Webinars

ISCBacademy Webinar Series

Archived Webinars

Welcome to the ISCBacademy Webinar Series. In conjunction with the communities of special

Other ISCB Webinars

interest (COSIs), select presentations are invited to give a live-streamed talk about their research. Access to the webinar series is complimentary for all. Non ISCB Members must register.

Be sure to check back regularly for information about upcoming webinars or to watch recordings of previous presentations.

Please use the links below to find more information or to register for an upcoming webinar:

- July 21, 2020 at 11:00AM EDT, Pooled CRISPR screens with imaging on microRaft arrays reveals stress granule-regulatory factors by Emily Wheeler, hosted by IRNA COSI and the RNA Society
- July 30, 2020 at 9:00AM EDT, Southern African Human Population Structure an Opportunity to Expand Genomics Research Worldwide by Caitlin Uren, hosted by ASBCB
- August 24, 2020 at 11:00AM EDT, Unravelling the mystery of orphan genes to understand the origins of genetic novelty by Nikos Vakirlis, hosted by EvolCompGen and SMRE
- September 30, 2020 at 11:00AM EDT, RAXML-NG: a fast, scalable and user-friendly tool for maximum likelihood phylogenetic inference by Alexey Kozlov, hosted by EvolComoGen COSI

https://www.iscb.org/iscbacademy-webinars

🕯 iscb.org/ismb2020



https://www.iscb.org/ismb2020-registration



https://eccb2020.info/



RSG with DREAM 2020 | November 16-18, 2020 | DREAM Submissions

#### https://www.iscb.org/recomb-regsysgen2020

iscb.org/rockv2020



ROCKY 2020 | Dec 3 - 5, 2020 | Aspen/Snowmass, CO | HOME - ROCKY 2020

https://www.iscb.org/rocky2020



https://giw2020.ncku.edu.tw/

HOME

. . . . .

DCANICATION

PAF

s su

WEWREKZHIN NOR

NOINTORMATICIANS DECISTOV

#### International Conference on Bioinformatics 2020 (InCoB2020)

⑤ FEBRUARY 10, 2020

#### The 19th International Conference on Bioinformatics (InCoB 2020) will be held virtually from Nov 25-29, 2020 across Asia-Pacific and beyond.

We are delighted to invite you to the 19th InCoB. The conference will be virtually hosted for the first time. InCoB 2020 took on the theme of "Bioinformatics and the translation of data-driven discoveries", and will include presentations of original research results, discussions in plenary sessions, poster sessions, workshops, software demos and panel discussions related to the field of bioinformatics. This is a great opportunity for you to showcase your research!

It was originally planned to be held in Kunming, Yunnan, China and hosted by Kunming University of Science and Technology (KMUST), Kunming, China. However, due to the global Covid-19 crisis, it was decided that the conference will be held virtually. More details will be made available as soon as possible.



https://www.apbionet.org/ international-conference-on-bioinformatics-2020-incob2020/ Terima kasih.

shofi.andari@statistika.its.ac.id shfandari@gmail.com