

# KSBI-BIML 2026

Bioinformatics & Machine Learning(BIML)  
Workshop for Life Scientists

생명정보학 & 머신러닝 워크샵(온라인)



## Proteogenomic analysis of human cancer

빈진혁 \_ 연세대학교



**KSBI**  
KOREAN SOCIETY FOR  
BIOINFORMATICS

| 한국생명정보학회



본 강의 자료는 한국생명정보학회가 주관하는 BIML 2026 워크샵을 목적으로  
제작된 것으로 해당 목적 이외의 다른 용도로 사용할 수 없음을 분명하게 알립니다.

이를 다른 사람과 공유하거나 복제, 배포, 전송할 수 없으며 만약 이러한 사항을 위반할 경우  
발생하는 **모든 법적 책임은 행위자 본인에게 있음**을 알립니다.

# KSBI-BIML 2026

## Bioinformatics & Machine Learning (BIML) Workshop for Life Scientists

한국생명정보학회가 주최하는 BIML-2026 동계 Bioinformatics & Machine Learning 교육 워크숍에 여러분을 초대합니다.

BIML 워크숍은 생명정보학 연구자들이 최신 AI바이오 분야의 인공지능 기반 분석 기술과 바이오 데이터 분석 기법을 이론과 실습을 통해 체계적으로 배울 수 있는 전문 교육 프로그램입니다. 2015년에 시작된 BIML 워크숍은 올해로 12년 차를 맞이하며, 국내 생명정보학 분야의 최초이자 최고 수준의 교육 프로그램으로 자리 잡았습니다. 이번 워크숍은 크게 인공지능바이오(AI바이오) 분야와 디지털바이오 분야, 두 분야로 구성됩니다.

AI바이오 분야에서는 생명정보 분석에 폭넓게 응용되고 있는 다양한 인공지능 기반 자료 모델링 기법을 다룰 예정입니다. 특히, 인공지능 심층학습을 활용한 단백질 구조 예측, 유전체 분석, 신약 개발에 대한 이론 및 실습 강의를 진행됩니다.

또한 디지털바이오 분야에서는 단일세포오믹스, 공간오믹스, 멀티오믹스, 메타오믹스에 대한 강의도 마련되어 있어, 연구자들의 분석 역량 강화에 실질적인 도움을 줄 것으로 기대됩니다.

또한 2024년부터 추가된 의료정보 자료 분석을 다루는 강의를 올해도 지속해서 운영하고자 합니다. 이는 최근 의료정보 자료 분석에 관한 연구 수요 증가를 반영한 것으로, 관련 연구를 수행하는 의과학자 및 의료정보 연구자들에게 유용한 지침을 제공할 것입니다.

또한, 올해도 생명정보학 기술의 다양화에 발맞춰 온라인 강좌를 대폭 확대했습니다. 올해는 무료 강좌 10개를 포함한 총 40개 이상의 강좌가 개설되며, 연구 주제에 맞는 강좌 추천과 강연료 할인 혜택도 제공합니다.

BIML-2026는 국내 주요 연구 중심 대학의 전임 교수 및 각 분야 최고 전문가들의 강의로 구성되어 있으며, 기초 이론부터 최신 연구 동향까지 아우르는 심도 있는 교육의 장이 될 것으로 확신합니다.

여러분의 많은 관심과 참여를 기대합니다!

2026년 2월

한국생명정보학회장 류 성 호

# Proteogenomic analysis of human cancer

최근에 급속하게 발전된 바이오테크놀로지 기술의 발달로 인해 생명 시스템에 대한 다차원적 이해가 가능하게 되었으며, 특히 Next generation sequencing (NGS)을 활용한 유전체 연구와 Mass-spectrometry (MS)를 활용한 단백질체 연구가 활발해지고 있다. 암과 같이 복잡한 분자 기전으로 일어나는 질병은 단일 차원의 데이터만으로는 진단 및 치료 타겟 발굴을 위한 생물학적 이해가 어렵다고 알려지면서 유전자 레벨에서 단백질 레벨까지 통합적인 해석을 하기 위한 연구들이 활발해지고 있다. 이러한 접근법 중 대표적인 유전단백체 (Proteogenomics: Proteomics + Genomics) 연구 방법은 동일한 생물학적 샘플에서 유전체, 전사체, 단백질체 데이터를 동시에 생산하고 이들 간의 통합분석을 통해 연구대상에 대한 새로운 생물학적 의미를 도출하는 데 유용하게 활용되고 있다. 본 강의를 통해 현대 오믹스 연구의 두 축인 유전체, 단백질체 데이터에 대해 이해하고 이 두 데이터가 어떻게 통합되어 암을 진단하고 치료 타겟을 발굴하는데 활용될 수 있는지 배울 수 있다.

강의는 다음의 내용을 포함한다:

- NGS 및 MS 개요
- 유전단백체 연구 개요
- 유전체-단백체 데이터 통합 방법론 소개
- 암 연구를 위한 유전단백체 연구동향

\* 참고강의교재: 강의 자료 제공

\* 교육생준비물: 노트북 (메모리 8GB 이상, 디스크 여유공간 30GB 이상)

\* 강의 난이도: 초급

\* 강의: 빈진혁 교수 (연세대학교 의과대학 의생명시스템정보학교실)

# Curriculum Vitae

**Speaker Name: Jinhyuk Bhin, Ph.D.**



## ► Personal Info

Name Jinhyuk Bhin  
Title Assistant Professor  
Affiliation Yonsei University College of Medicine

## ► Contact Information

Address 20, Eonju-ro 63-gil, Gangnam-gu, Seoul, Republic of Korea  
Email jbhin@yuhs.ac

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## Research Interest

Oncogenomics, Proteogenomics, single cell and spatial transcriptomics

## Educational Experience

2010 B.S. in Chemical Engineering, Sungkyunkwan University, Korea  
2016 Ph.D. in Bioinformatics & Systems Biology, POSTECH, Korea

## Professional Experience

2016-2016 Postdoc, Institute for Basic Science, DGIST, Korea  
2017-2022 Postdoc, Netherlands Cancer Institute, The Netherlands  
2023- Assistant Professor, Yonsei University College of Medicine

## Selected Publications (5 maximum)

1. Bhin J\*, Yemelyanenko J\* et al., MYC is a clinically significant driver of mTOR inhibitor resistance in breast cancer, *Journal of Experimental Medicine*. 2023 Nov;220(1):e20211743
2. Bhin J\*, Dias.M\* et al., Multi-omics analysis reveals distinct non-reversion mechanisms of PARPi resistance in BRCA1- versus BRCA2-deficient mammary tumors, *Cell Reports*. 2023 May; 42(5):112538.
3. Zingg D\*, Bhin J\* et al., Truncated FGFR2 is a clinically actionable oncogene in multiple cancers, *Nature*. 2022 Aug;608(7923):609-617.
4. Moon D\*, Bhin J\* et al., Proteogenomic characterization of human early-onset gastric cancer, *Cancer Cell*. 2019 Jan 14;35(1):111-124.e10.
5. Bin B\*, Bhin J\* et al, Requirement of zinc transporter ZIP10 for epidermal development: Implication of the ZIP10-p63 axis in epithelial homeostasis, *Proc Natl Acad Sci U S A*. 2017 Nov 14;114(46):12243-12248.

# KSBi-BIML 2024

## Proteogenomic analysis of human cancer

빈진혁

연세대학교 의과대학  
의생명시스템정보학교실

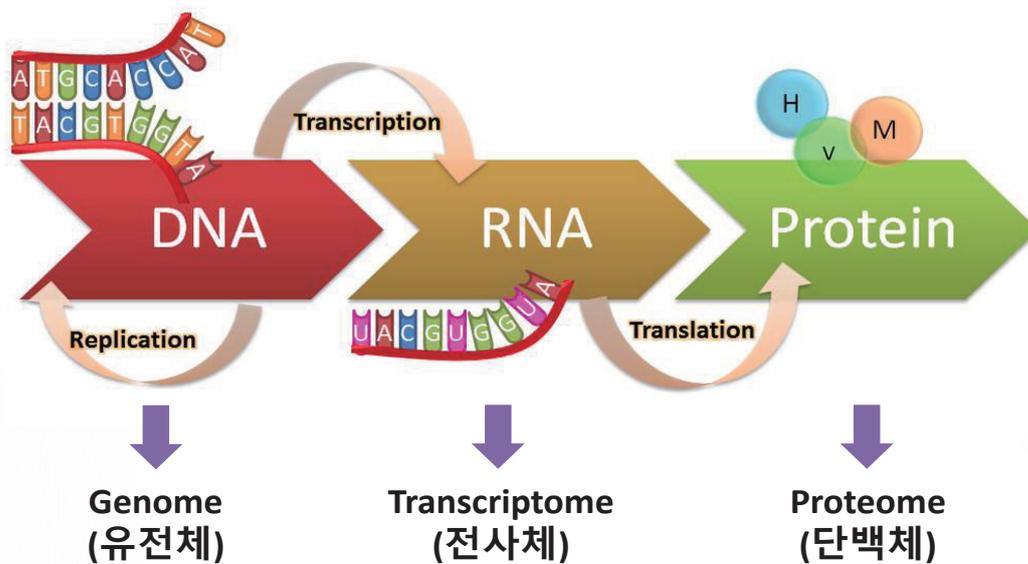
### 목차

- 유전단백체학 (Proteogenomics) 개념 소개
- 유전단백체학을 이용한 암 연구 동향
- 유전단백체학 분석 방법론/툴 소개

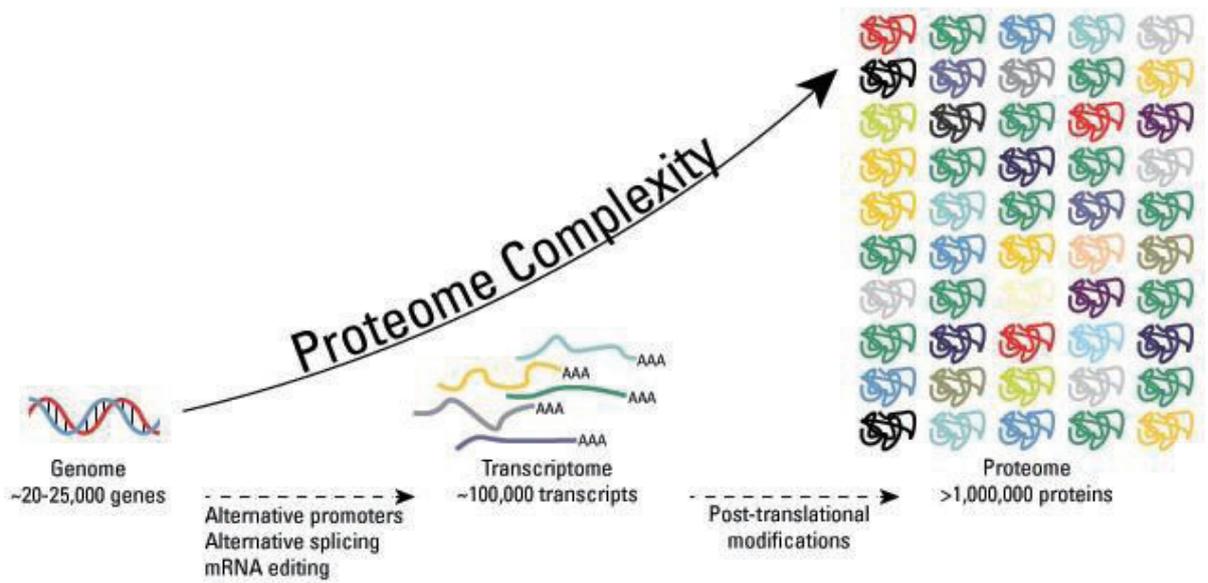
# 목차

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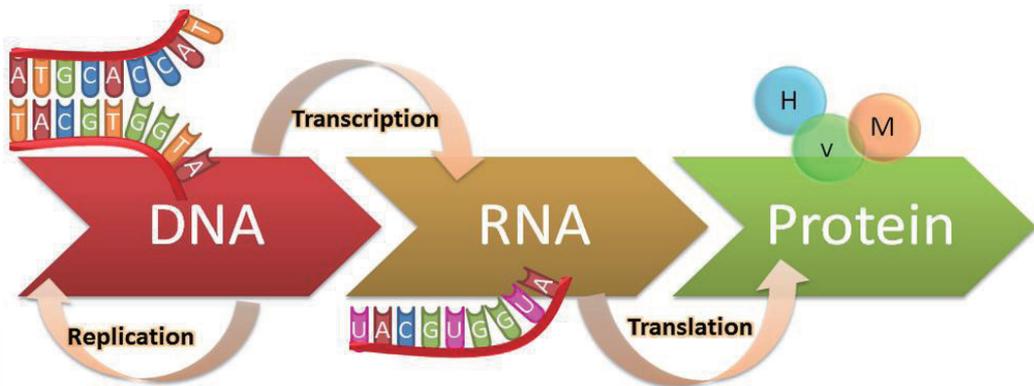
## Central Dogma



# Complexity of molecular entities



# Central Dogma



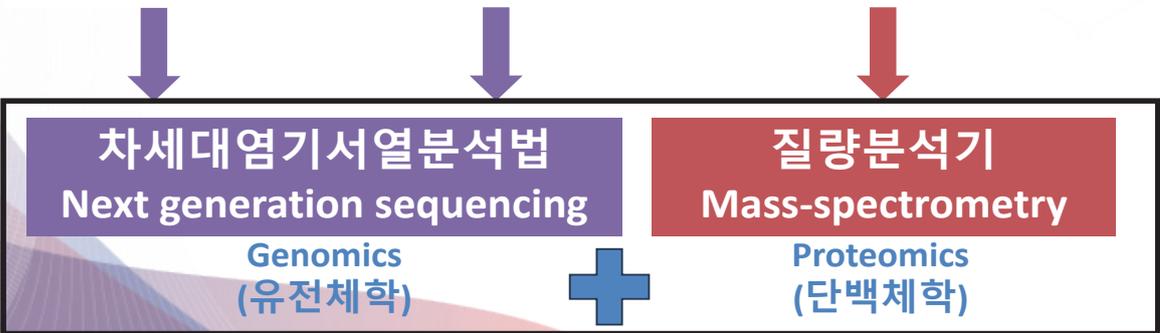
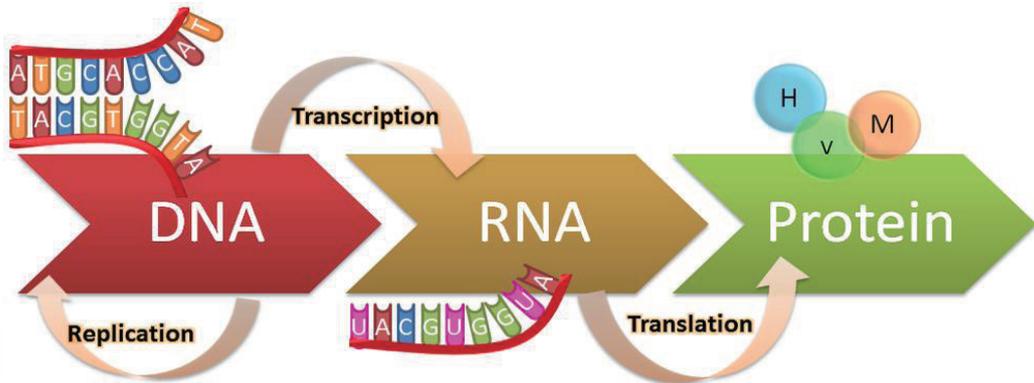
차세대염기서열분석법  
Next generation sequencing

Genomics  
(유전체학)

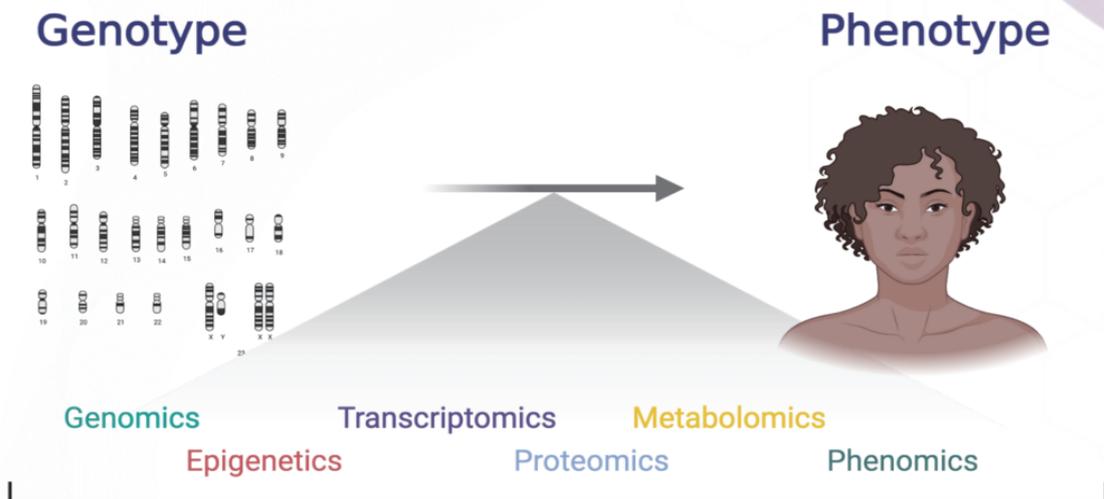
질량분석기  
Mass-spectrometry

Proteomics  
(단백체학)

# 유전단백체학 (Proteogenomics)



# 유전체-단백체 통합 연구의 필요성



A multi-omics approach is needed to understand complex biological systems

<https://arimagenomics.com>

# 유전체-단백체 통합 연구의 필요성



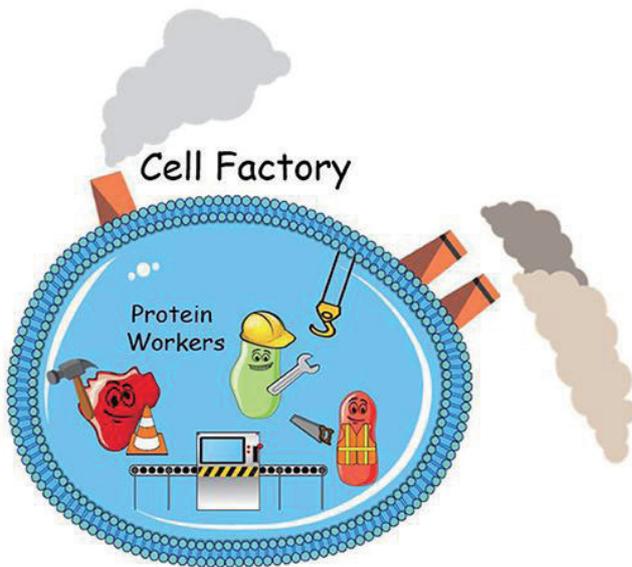
Genome remains the same but proteome changes

**DNA** - tells what possibly

**RNA** – tells what probably

**Proteins** – tells what actually happens

# 왜 단백체를 이해해야 하는가?



**To be functional in a cell**

- Folding
- Post-translational modification
- Transport
- Assembly into complexes
- Quality control

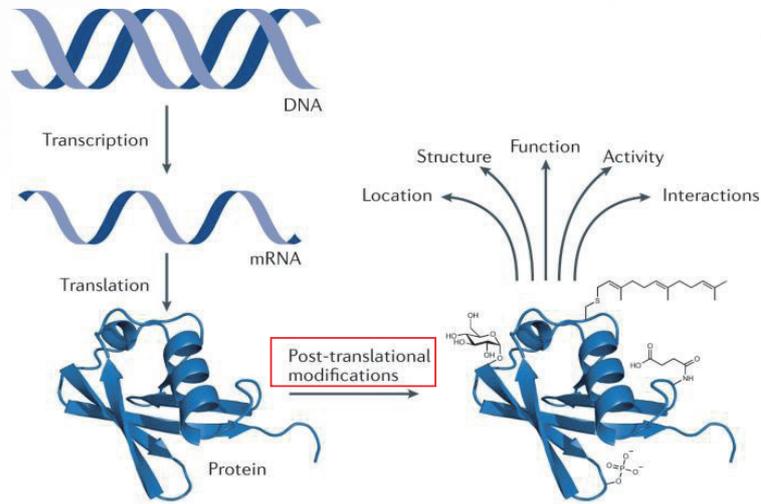
Proteins are the workers inside your cells

<https://kids.frontiersin.org>

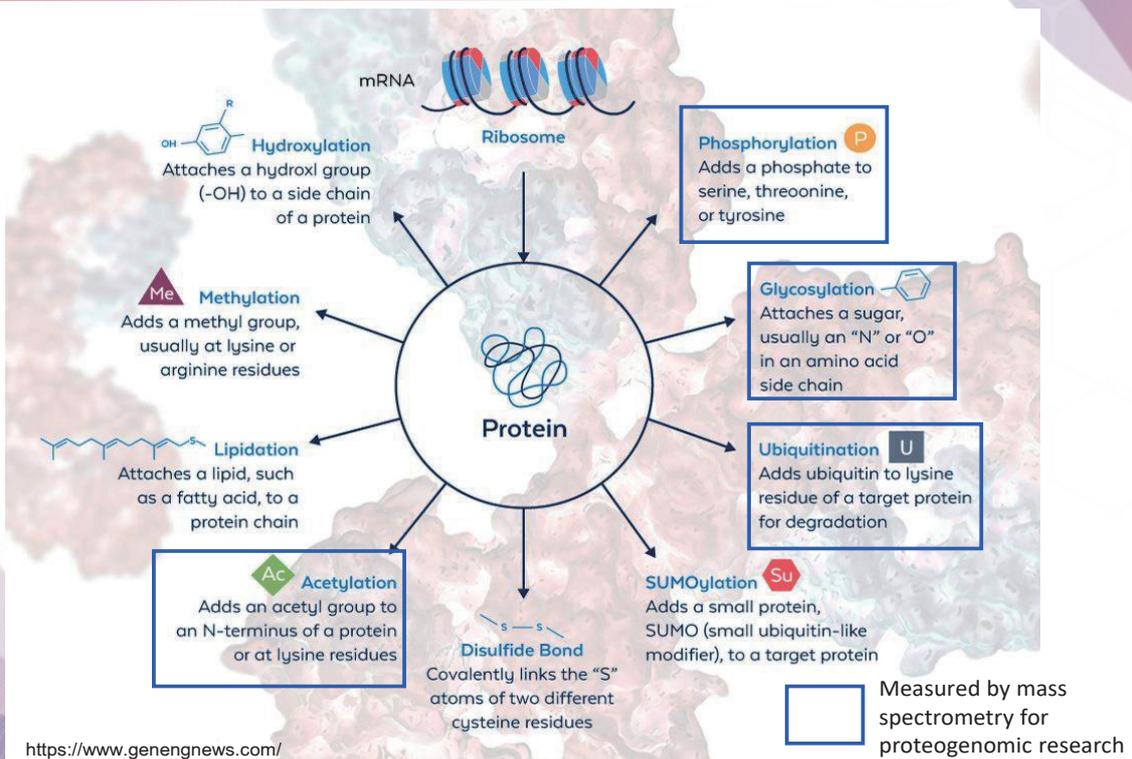
# 단백질 번역 후 변형 - 기능 조절

## Post-translational modification (PTM)

: chemical modification of a protein after translation is complete, essential for its functionality



# 단백질 번역 후 변형 - 기능 조절

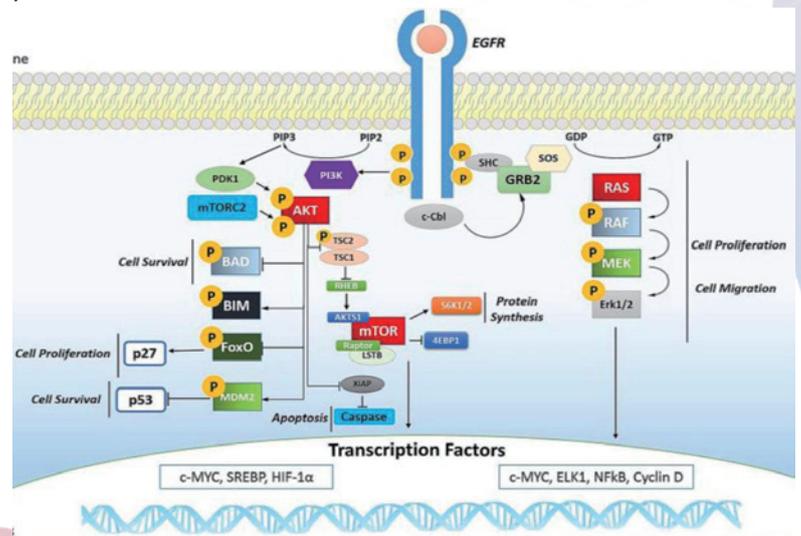


## 단백질 번역 후 변형 - 기능 조절

- **Phosphorylation**

- addition of a phosphate group to an amino acid, commonly serine (86.4%), threonine (11.8%), or tyrosin (1.8%)

- key regulatory mechanism in many cellular processes, including cell cycle, signal transduction pathway, and metabolic regulation



## 단백질 번역 후 변형 - 기능 조절

- **Glycosylation**

- attachment of sugar moieties to proteins, typically at asparagine (N-linked) or serine/threonine (O-linked) residue

- cell-cell recognition, immune response, signaling pathway

- **Acetylation**

- addition of an acetyl group to the lysine residue of proteins

- gene expression regulation by modifying histone, chromatin structure

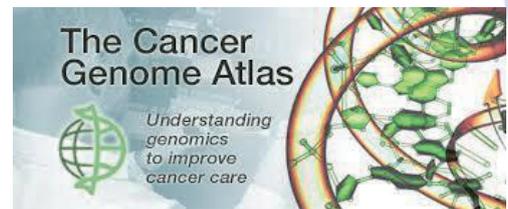
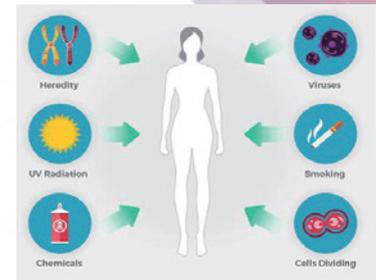
- **Ubiquitination**

- attachment of ubiquitin (a small regulatory protein) to lysine residue

- cell cycle control, DNA repair, immune response

# 유전단백체 연구의 필요성 - 암 연구

- Cancer is a genetic disease, caused by changes in genes that control the way cells grow and multiply
- **The Cancer Genome Atlas (TCGA)**
  - started in 2006 and ended in 2013, by NCI
  - to catalog genetic mutations responsible for cancer
  - to identify unique and reproducible genomic differences that exist among patients and determine whether these differences could lead to the development of individual treatment



# 유전단백체 연구의 필요성 - 암 연구

## NATIONAL CANCER INSTITUTE THE CANCER GENOME ATLAS

### TCGA BY THE NUMBERS

TCGA produced over

**2.5**  
PETABYTES  
of data

To put this into perspective, **1 petabyte** of data is equal to

**212,000**  
DVDs

TCGA data describes

**33**  
DIFFERENT  
TUMOR TYPES

...including

**10**  
RARE  
CANCERS

...based on paired tumor and normal tissue sets collected from

**11,000**  
PATIENTS

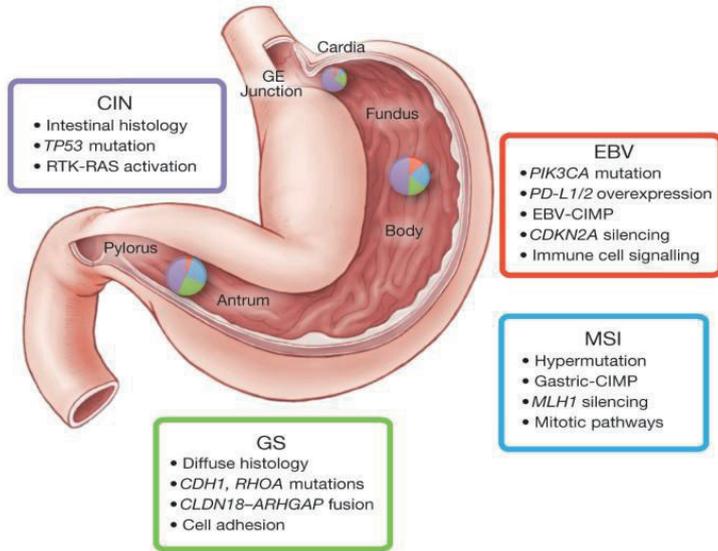
...using

**7**  
DIFFERENT  
DATA TYPES



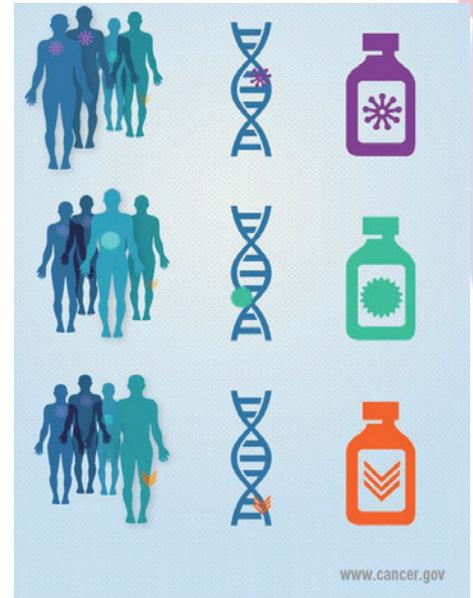
# 유전단백체 연구의 필요성 - 암 연구

## TCGA - Gastric Cancer



TCGA, 2014, Nature (Gastric cancer)

## Precision Medicine

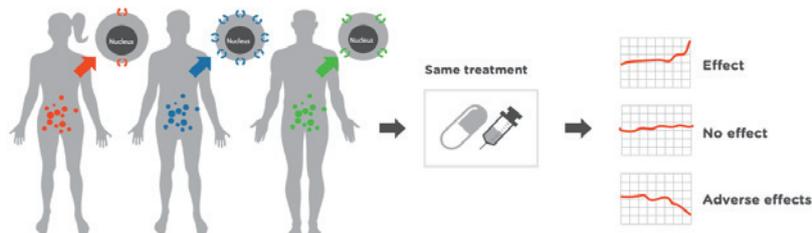


www.cancer.gov

# 유전단백체 연구의 필요성 - 암 연구

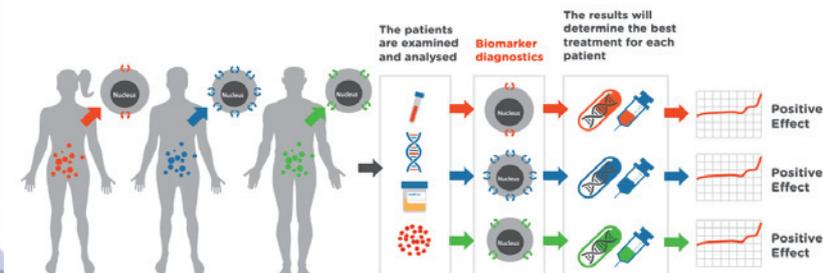
### TRADITIONAL MEDICINE: SAME TREATMENT FOR ALL

Cancer patients with e.g. colon cancer receive the same therapy even though they have different biomarkers



### INNOVATIVE MEDICINE: PERSONALISED MEDICINE

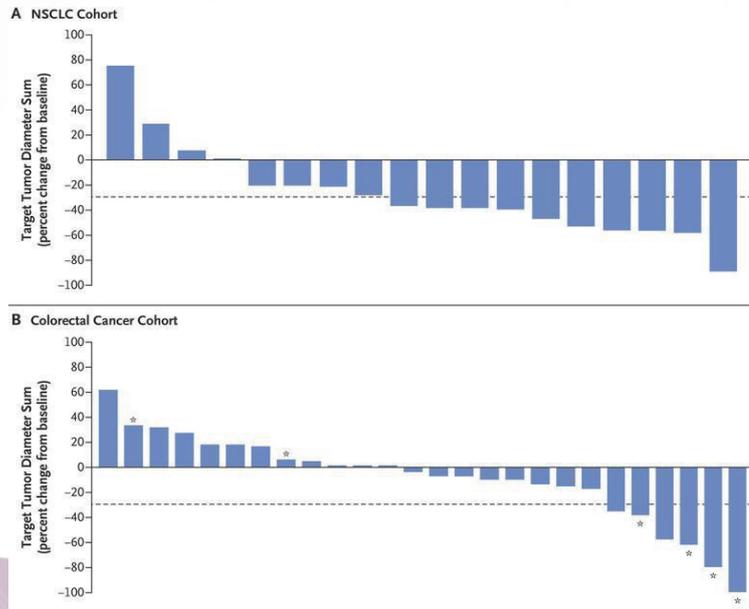
Cancer patients with e.g. colon cancer receive a personalised therapy based on their biomarkers



## 유전단백체 연구의 필요성 - 암 연구

- However, mutational profiling alone cannot guarantee accurate predictions about the quantity or activity of the proteins affected by driver mutations

Different responses to BRAF inhibitor Vemurafenib among patients with BRAF V600 mutations

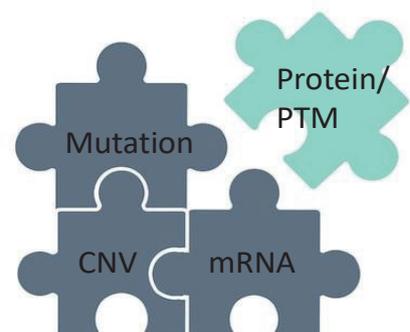


## 유전단백체 연구의 필요성 - 암 연구

- Need to meaningfully interpret the contribution of genomic features to a specific patient's tumor biology
- Need to have a more complete and precise picture of molecular pathology
- RNA expression levels are often poor predictors of actual protein levels
- Oncogenic/cancer-driving signalings are usually mediated by PTM



**Need to incorporate proteomic information**



# CPTAC – 유전단백체 콘소시움 탄생

## CPTAC CLINICAL PROTEOMIC TUMOR ANALYSIS CONSORTIUM

From 2011, by NCI

### Flagship papers from CPTAC

**nature**

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Article | Published: 20 July 2014

**Proteogenomic characterization of human colon and rectal cancer**

Bing Zhang, Jing Wang, Xiaoling Wang, Jing Zhu, Qi Liu, Zhaohui Shi, Matthew C. Chambers, Lisa J. Zimmerman, Kent F. Shaddox, Sangtae Kim, Sherri R. Davies, Sean Wang, Pei Wang, Christopher B. Kinsinger, Robert C. Rivers, Henry Rodriguez, R. Reid Townsend, Matthew J. C. Ellis, Steven A. Carr, David L. Tabb, Robert J. Coffey, Robbert J. C. Slebos, Daniel C. Liebler & the NCI CPTAC

Colorectal cancer  
(Nature, 2014)

**nature**

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Article | Published: 25 May 2016

**Proteogenomics connects somatic mutations to signalling in breast cancer**

Philipp Mertins, D. R. Mani, Kelly V. Ruggles, Michael A. Gillette, Karl R. Clauser, Pei Wang, Xiantong Wang, Jana W. Qiao, Song Cao, Francesca Petralia, Emily Kawaler, Filo Mundt, Karsten Knust, Zhidong Tu, Jonathan T. Lei, Michael L. Gatza, Matthew Wilkerson, Charles M. Perou, Yenkata Yellaosantula, Kuan-Sin Huang, Chenwei Lin, Michael D. McLellan, Ping Yan, Sherri R. Davies, NCI CPTAC

Breast cancer  
(Nature, 2016)

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**Integrated Proteogenomic Characterization of Human High-Grade Serous Ovarian Cancer**

Hui Zhang, Tao Liu, Zhen Zhang, Samuel H. Payne, Bai Zhang, Jason E. McDermott, Jian-Ying Zhou, Vladimir A. Petuk, Li Chen, Dejiyi Ray, Shaohong Sun, Feng Yang, Lijun Chen, Jing Wang, Parit Shah, Seung Won Cho, Paul Ayyan, Songhua Wu, Yuan Tian, Maria A. Gillette, Theresa R. Coak, Colton Choi, Matthew E. Monroe, Shihua Thomas, Bing He, Chaochao Wu, Ronald J. Moore, Kun-Heng Yu, David L. Tabb, David Femp, Wael Bahra, Yue Wang, Henry Rodriguez, Emily S. Buj, Tara Hiltz, Robert C. Rivers, Lori Sokol, Hang Zhu, Le-Ming Shih, Leslie Cope, Akhlesh Pandey, Bing Zhang, Michael P. Snyder, Douglas A. Levine, Richard D. Smith, Daniel W. Chan, Karin D. Rodland & the CPTAC Investigators | Show full text | Show footnotes

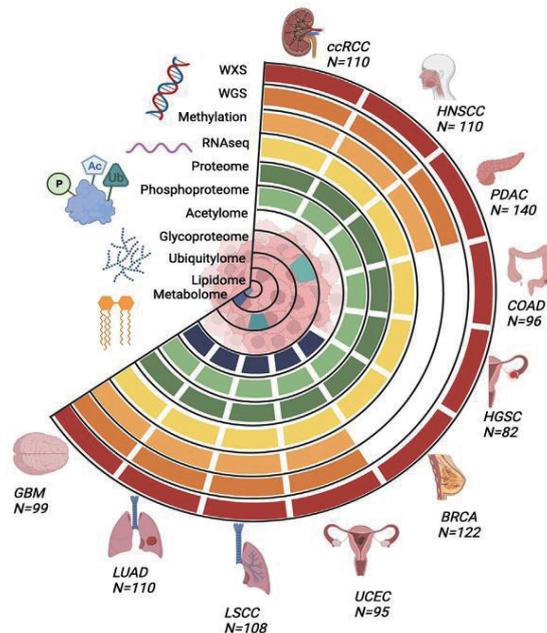
Ovarian cancer  
(Cell, 2016)

# CPTAC – 유전단백체 콘소시움 탄생

## CPTAC CLINICAL PROTEOMIC TUMOR ANALYSIS CONSORTIUM

From 2011, by NCI

~10 cancer types, ~ 1,000 tumors



# 국내유전단백체 연구 근황



Gastric cancer, 2019



Pancreatic cancer, 2022

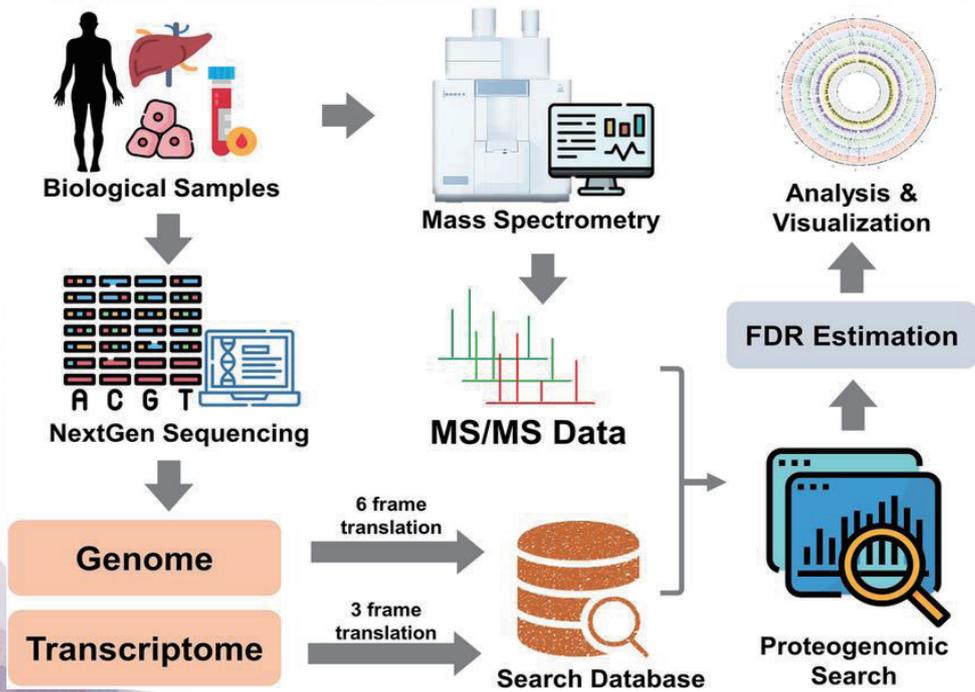


Glioblastoma, 2024

## 목차

- 유전단백체학 (Proteogenomics) 개념 소개
- 유전단백체학을 이용한 암 연구 동향
- 유전단백체학 분석 방법론/툴 소개

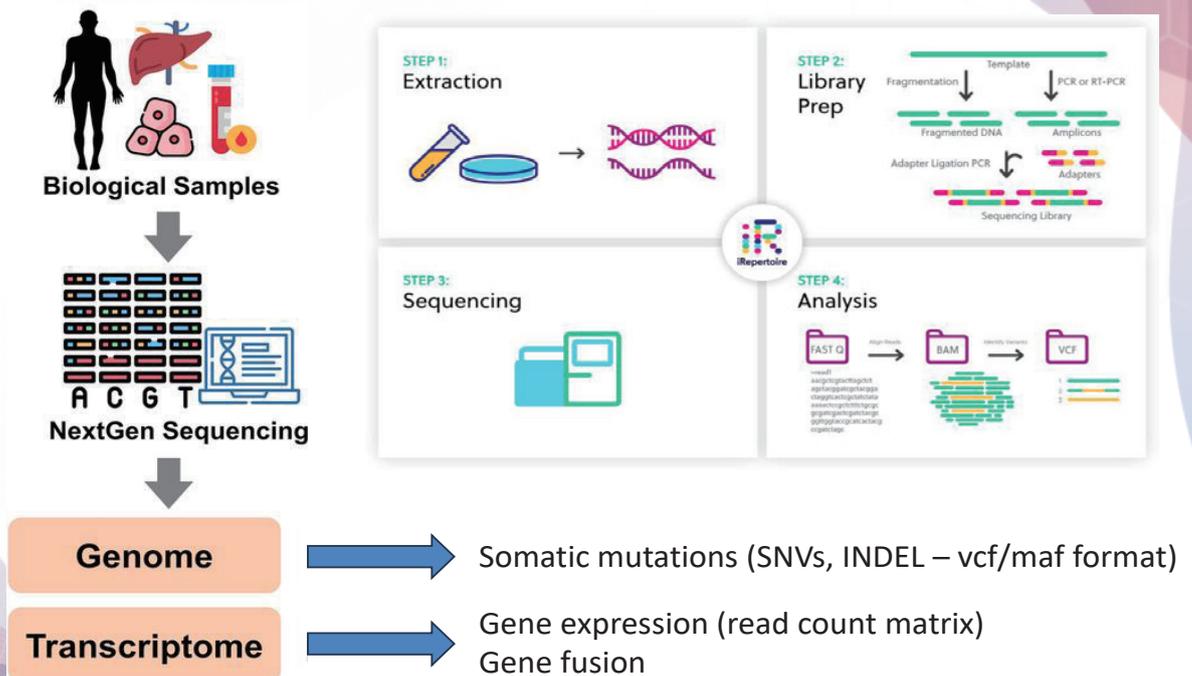
# 유전단백체 기반 암 연구 workflow



Raj et al., 2023, J. Proteins and Proteomics

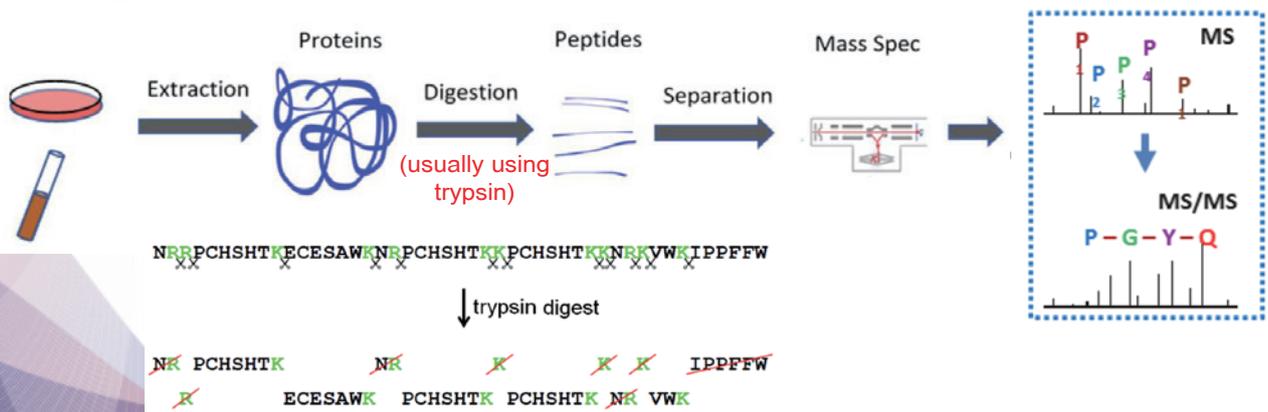
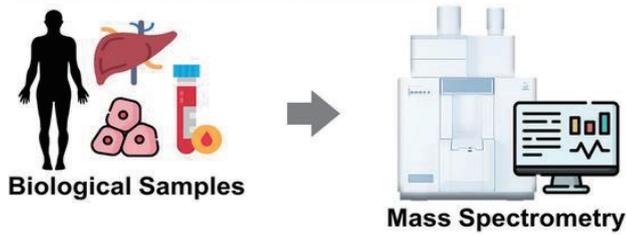
# 유전단백체 기반 암 연구 workflow

## 1. NGS for genomic and transcriptomic analyses



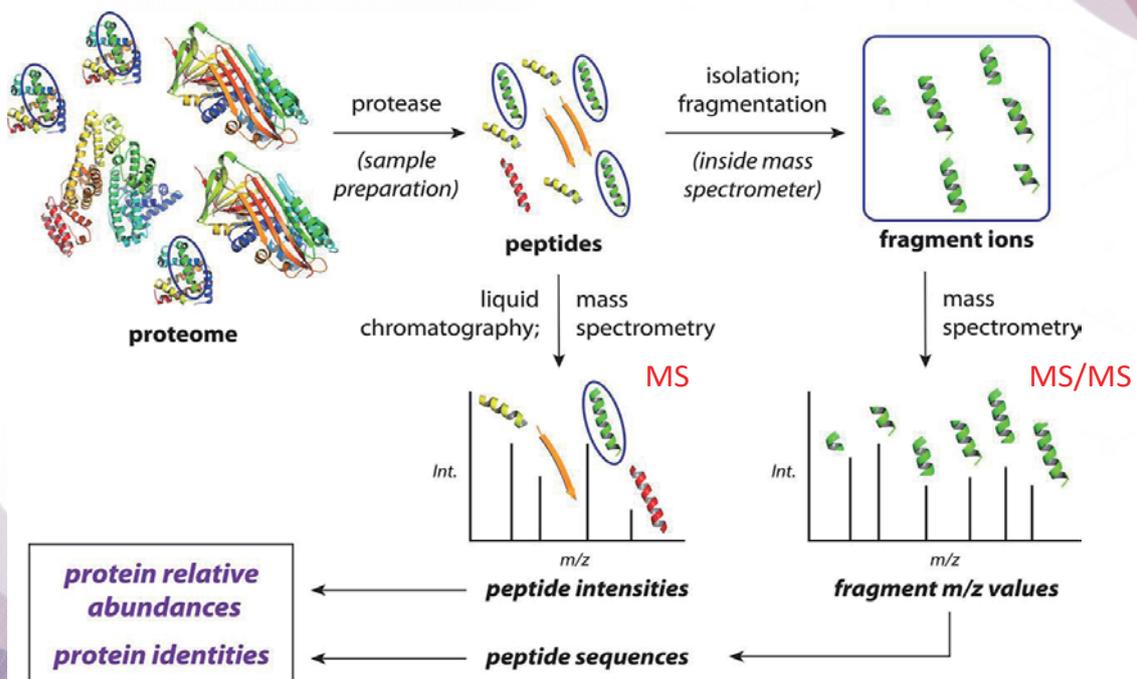
# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses



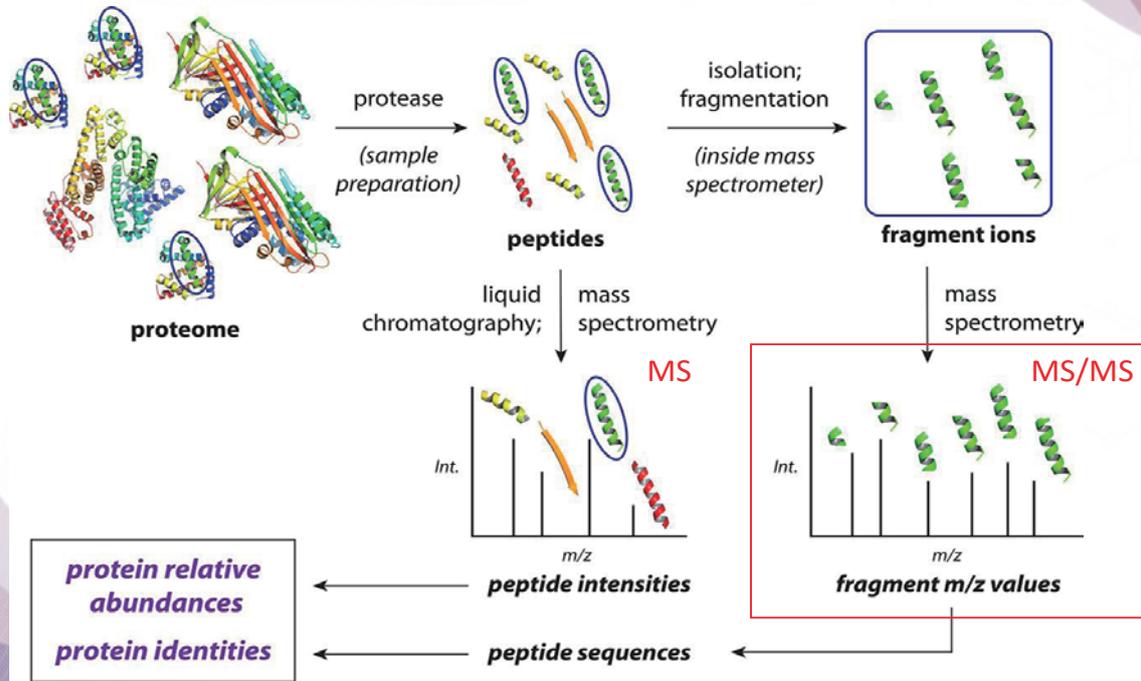
# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses



# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses



Shuken, 2023, J. Proteome Res

# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses

### Peptide sequence identification

Reference protein sequence database

```
>sw|Q9H8E1|F2B1_HUMAN Icaulin-like grow
protein 1 OS=Homo sapiens GN=IQ2B1 PE=
MFKLVIKGLNEPTVFALEKVFARERISVQGFVRSOVA
GEYEQWALELIDYVFPQKQKRIQ1NIIKPLRHWYLD
RELAVYVYVYVYVYVYVYVYVYVYVYVYVYVYVYV
QKPKQKQKQKQKQKQKQKQKQKQKQKQKQKQKQK
KREKNAKAKALISYVYVYVYVYVYVYVYVYVYVYVYV
LIEKREKREKREKREKREKREKREKREKREKREKREK
DAVYKDAVYKDAVYKDAVYKDAVYKDAVYKDAVYKDA
MVKYFIKDAVYKDAVYKDAVYKDAVYKDAVYKDAVYK
KAGQRIKDAVYKDAVYKDAVYKDAVYKDAVYKDAVYK
YKQKQKQKQKQKQKQKQKQKQKQKQKQKQKQKQK
>sw|Q9TF68|ZK384_HUMAN Zinc finger prote
GN=ZNF384 PE=1 DT=2
```

peptide seq.

theoretical spectra

peptide scores

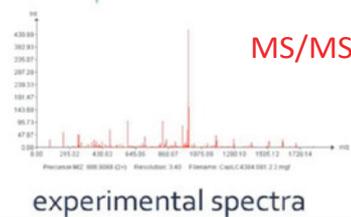
*in silico*  
digest

*in silico*  
MS/MS

scoring  
function

- 1) YSFVATAER 34
- 2) YSFVSAIR 12
- 3) FFLIGGGGK 12
- ...

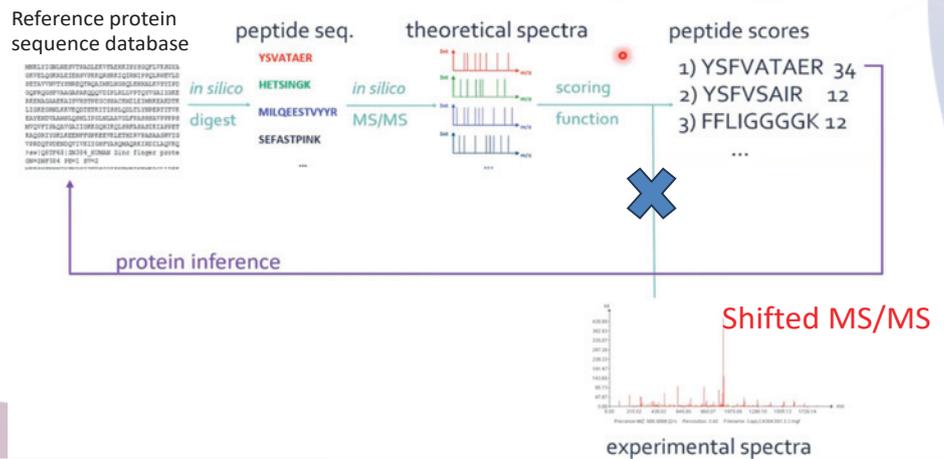
protein inference



# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses

- **Peptide identification in tumor samples is incomplete!**
  - A typical method to identify peptide sequences from MS/MS scans relies on known protein sequences (wild-type) in reference proteome database
  - tumor samples contain many mutations, novel splice junctions, and novel protein-coding regions, resulting in novel peptide sequences
  - experimental MS/MS scans cannot match theoretical spectra due to shifts in scan peaks



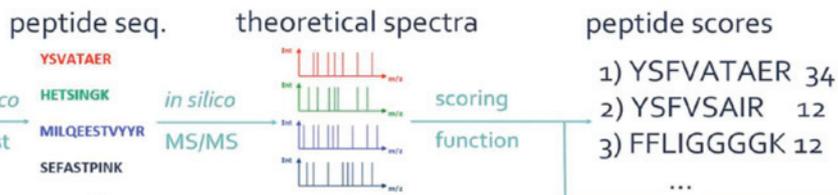
# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses

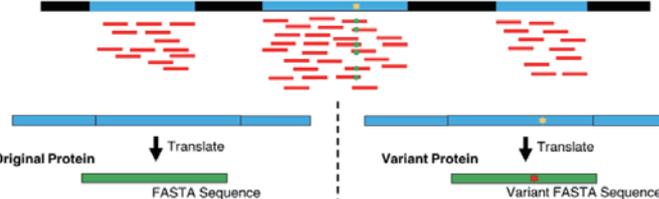
### Peptide sequence identification

**Customized (sample-specific) protein sequence database**

```
>=|QI0E1I1F2B1_M0M0 IaaLis-like grow
protein 1.000000 epiLis QN1G2E2P1 PE*
MRLK11DGLNEFTTALLETFAERLIVDQDFVRSOVA
GEYEGQALELIDYVFFQQRH1G1N1H1PGLRHWYLD
SESNVYVYVYVYVYVYVYVYVYVYVYVYVYVYVYV
QGRQDQFFAAGAFAGQVQVYVYVYVYVYVYVYVYV
RREDAIAAKAL1FVYVYVYVYVYVYVYVYVYVYVYV
L1EKEHMLLAVYVYVYVYVYVYVYVYVYVYVYVYV
DAVYDVAHMLQML1FVYVYVYVYVYVYVYVYVYVYV
MVFYVYVYVYVYVYVYVYVYVYVYVYVYVYVYVYV
FAQQR1VYVYVYVYVYVYVYVYVYVYVYVYVYVYV
YVYVYVYVYVYVYVYVYVYVYVYVYVYVYVYVYV
>=|QI0F48|2K384_M0M0 Zinc finger prote
DN=2K384 PE=1 0T=2
```



Reference gene/genome

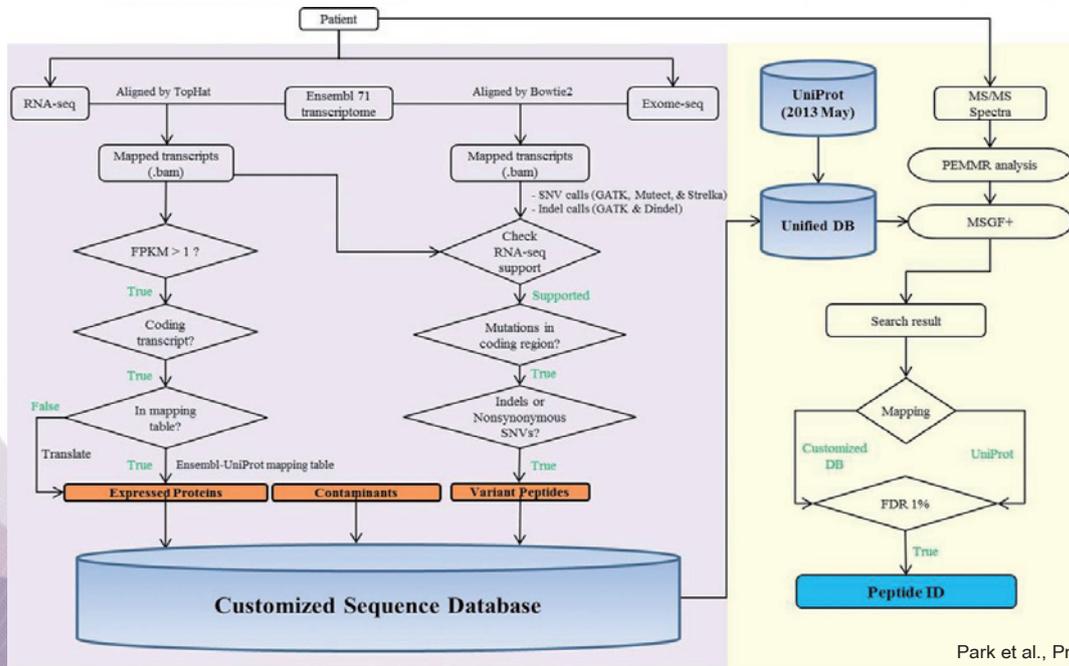


experimental spectra

# 유전단백체 기반 암 연구 workflow

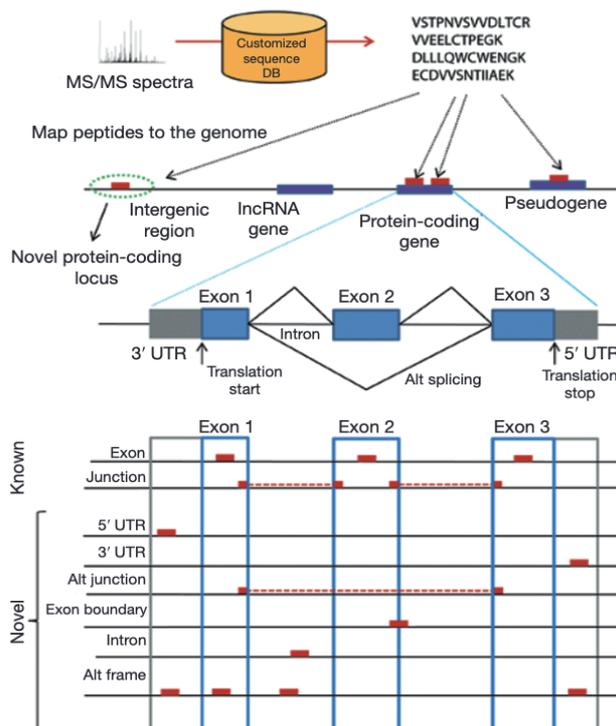
## 2. Mass-spectrometry for proteomic analyses

### Building customized protein sequence database



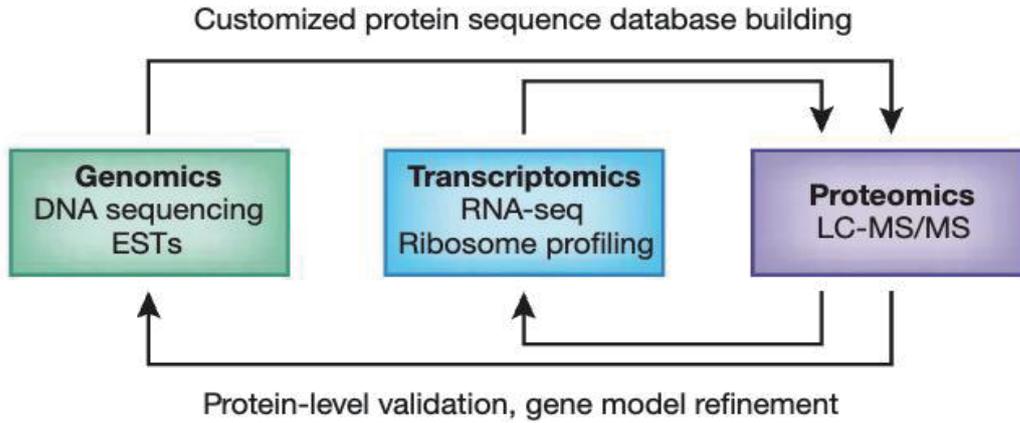
# 유전단백체 기반 암 연구 workflow

## 2. Mass-spectrometry for proteomic analyses



# 유전단백체 기반 암 연구 workflow

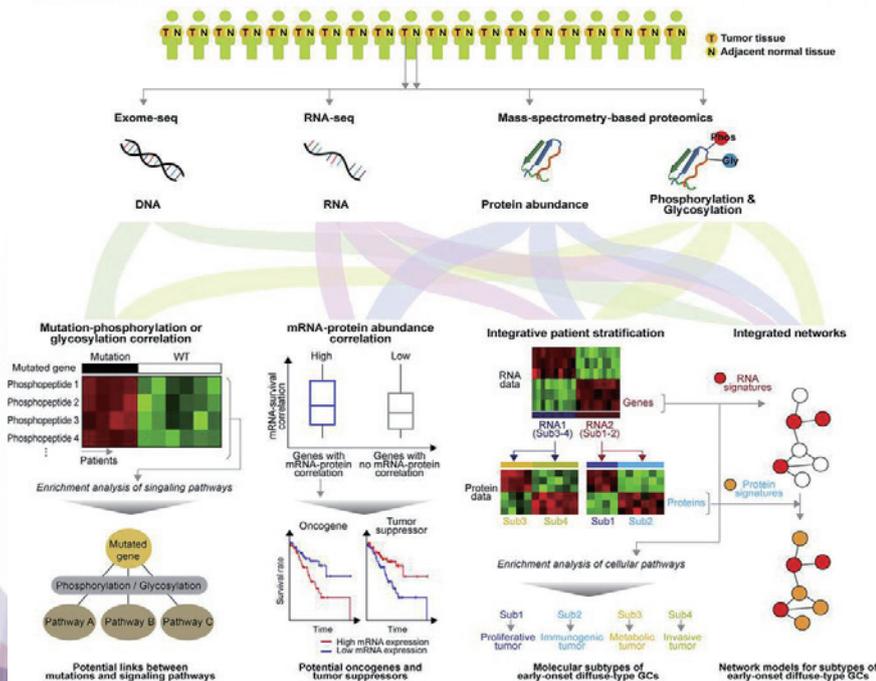
## 2. Mass-spectrometry for proteomic analyses



Nesvizhskii, Nature Methods, 2014

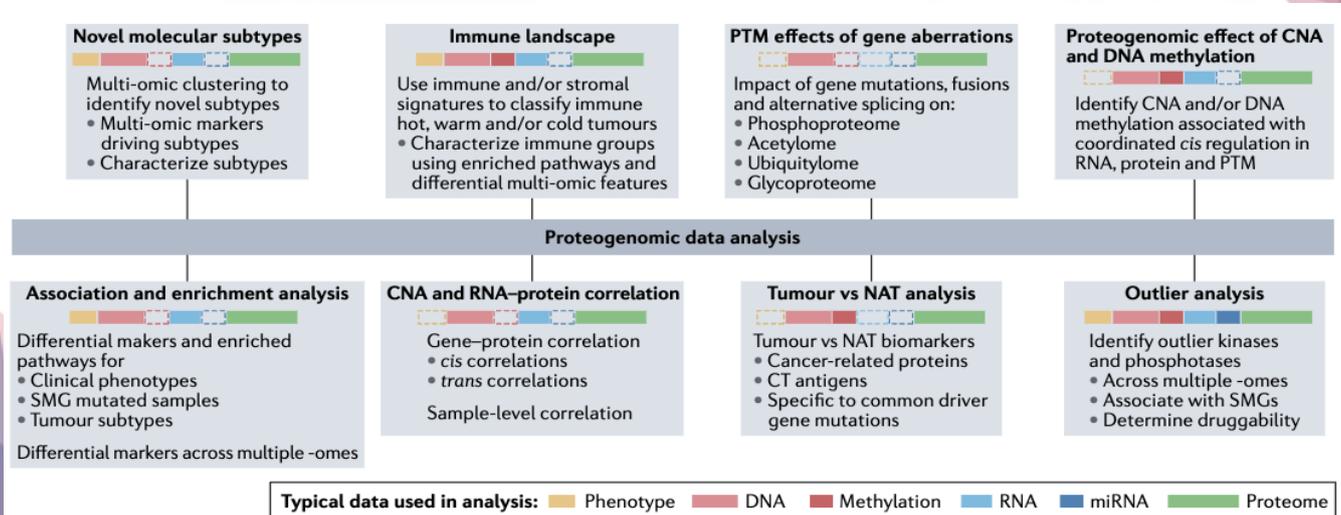
# 유전단백체 기반 암 연구 workflow

## 3. integrative analyses of genomic and proteomic data



Bhin and Mun et al., Cancer Cell, 2019

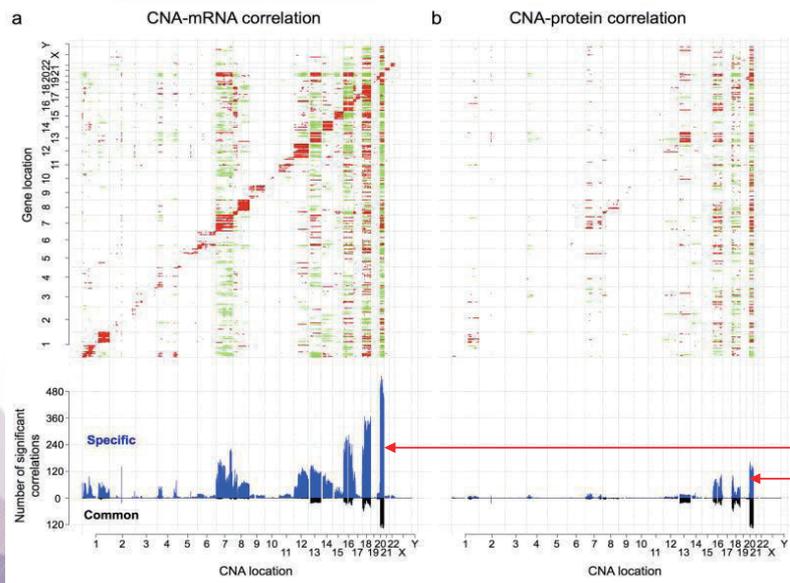
# 유전단백체 기반 암 연구 동향



Carr, Nature Review Cancer, 2022

# 유전단백체 기반 암 연구 동향

## 1. Deciphering proteomic effects of genomic alterations - measuring the correlation between different entities



**Proteogenomic effect of CNA and DNA methylation**  
Identify CNA and/or DNA methylation associated with coordinated cis regulation in RNA, protein and PTM

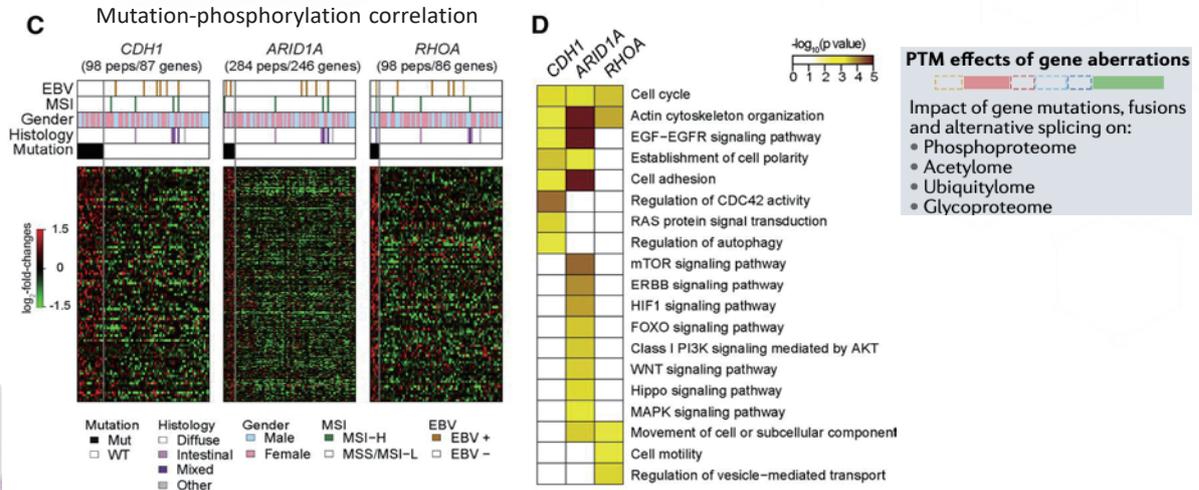
Impact of chr20q amplification on mRNA and protein (trans-acting CNN hotspot)  
→ potential cancer driver

Zhang, Nature, 2014 (Colorectal cancer)

# 유전단백체 기반 암 연구 동향

## 1. Deciphering proteomic effects of genomic alterations

- measuring the correlation between different entities

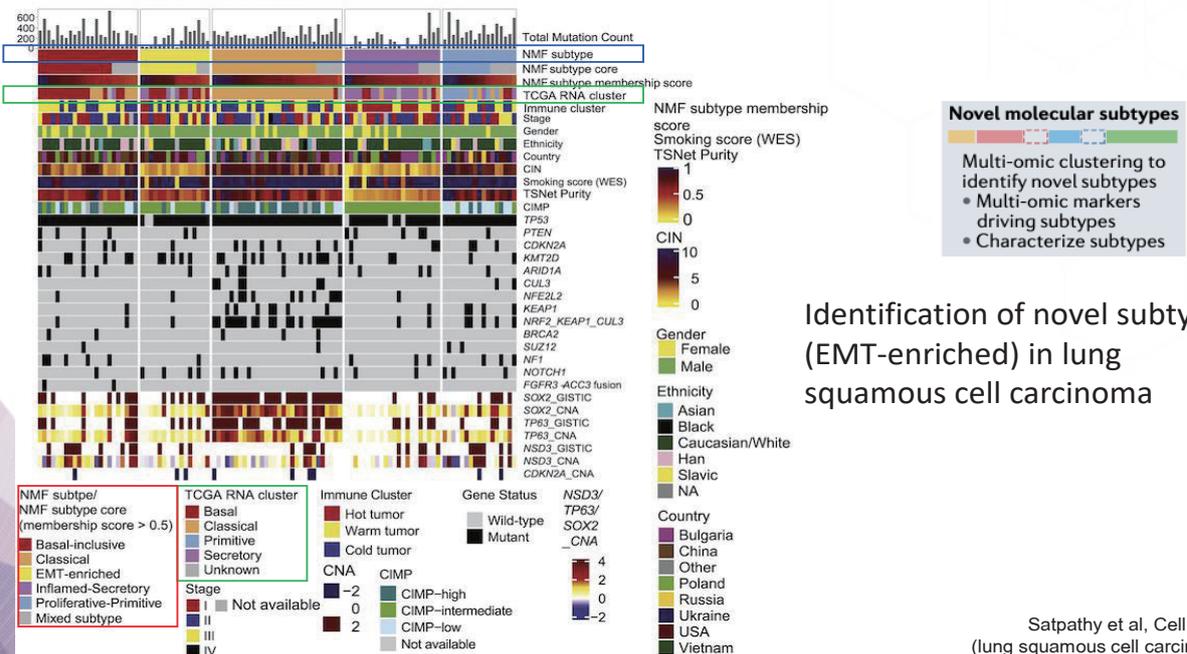


Bhin and Mun et al, Cancer Cell, 2019 (Gastric cancer)

# 유전단백체 기반 암 연구 동향

## 2. Redefining and characterizing cancer subtypes

- integrative clustering of transcriptomic and proteomic data

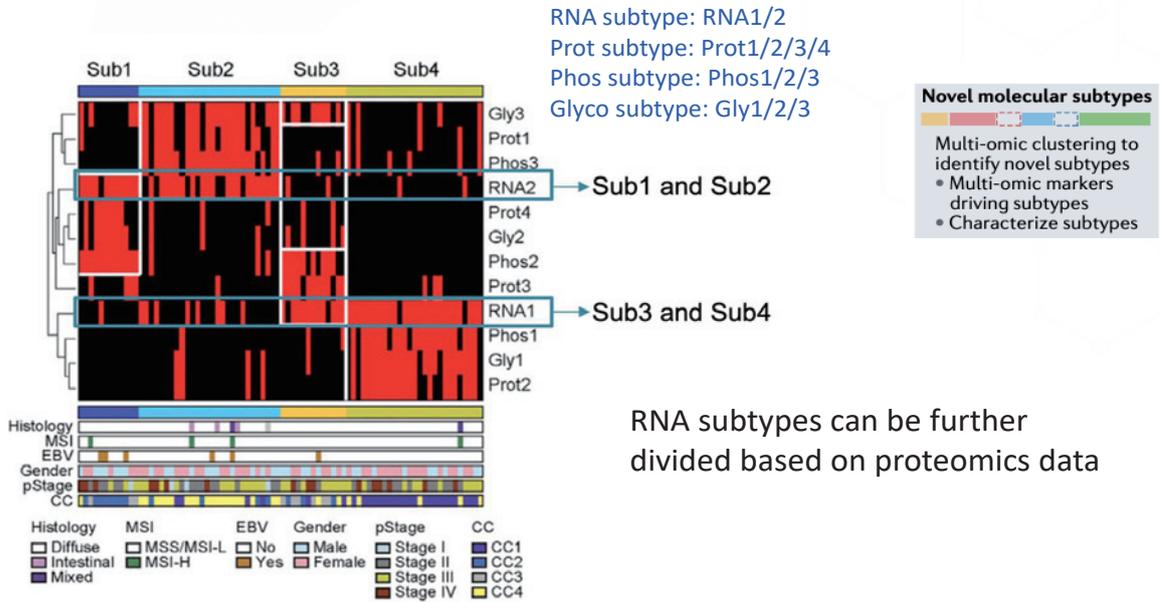


Satpathy et al, Cell, 2021 (lung squamous cell carcinoma)

# 유전단백체 기반 암 연구 동향

## 2. Redefining and characterizing cancer subtypes

- integrative clustering of transcriptomic and proteomic data

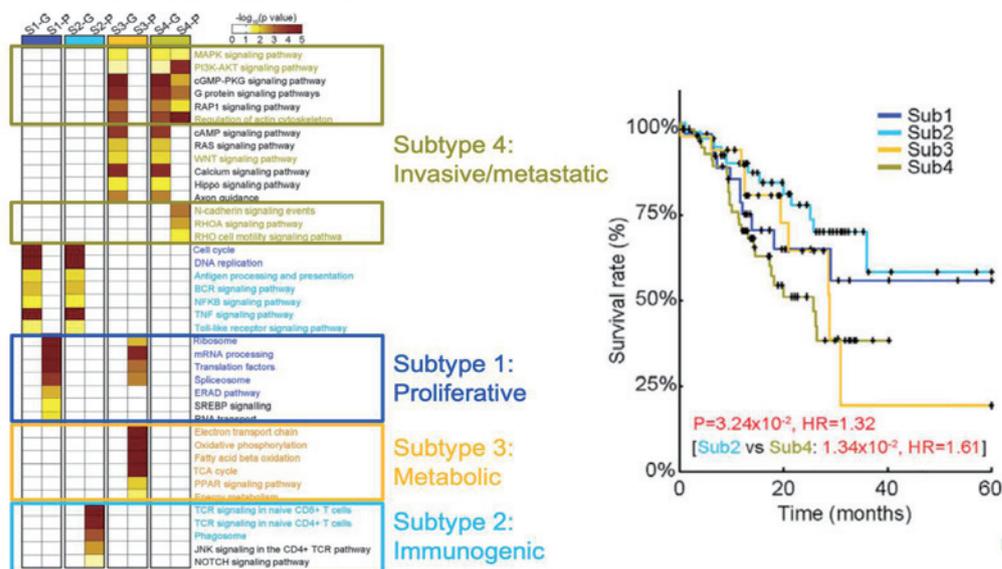


Bhin and Mun et al, Cancer Cell, 2019 (Gastric cancer)

# 유전단백체 기반 암 연구 동향

## 2. Redefining and characterizing cancer subtypes

- integrative clustering of transcriptomic and proteomic data

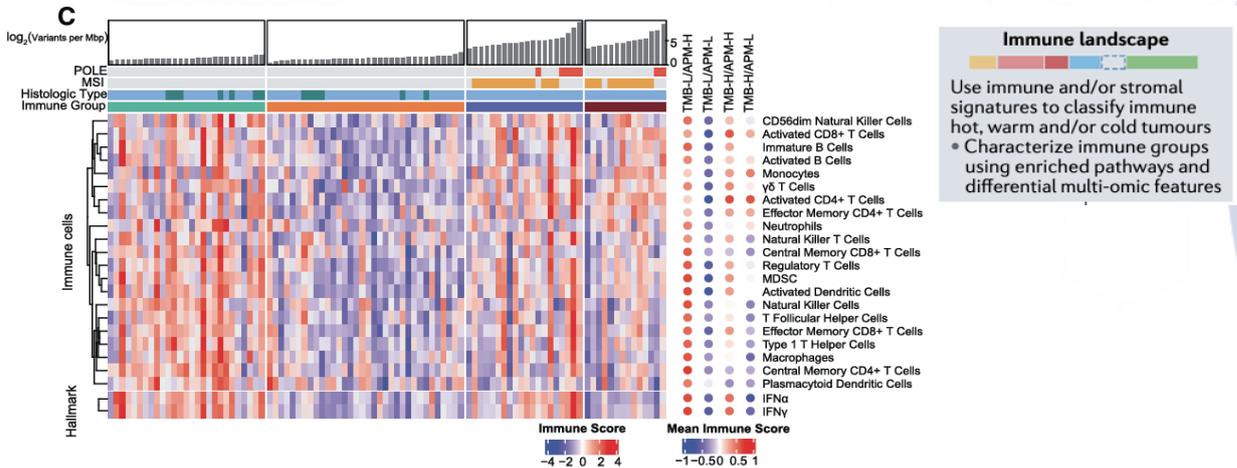


Bhin and Mun et al, Cancer Cell, 2019 (Gastric cancer)

# 유전단백체 기반 암 연구 동향

## 3. Assessing tumor immunogenicity

- proteomic evidence for immune status and neoantigen prediction



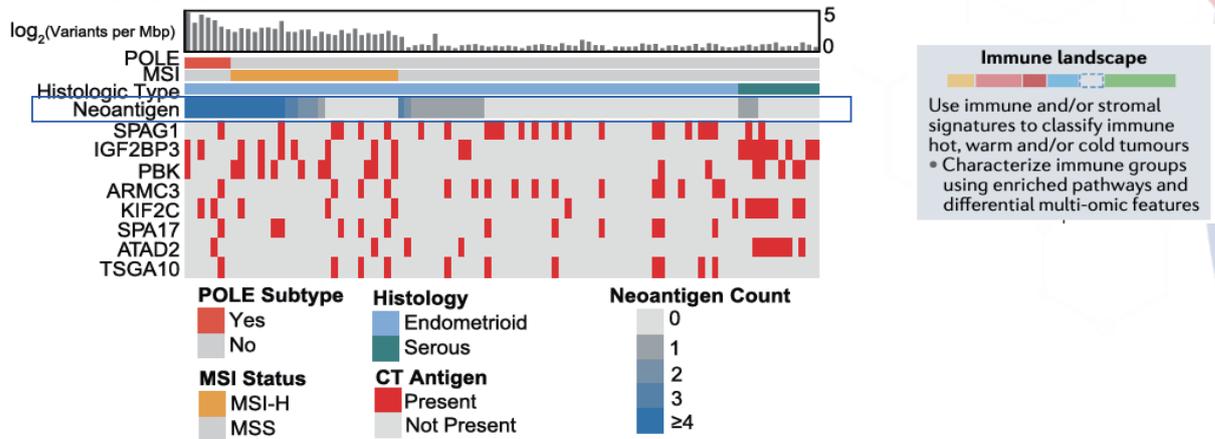
4 Immune subtypes in Endometrial carcinoma supported by proteomics data

Dou et al, Cell, 2020 (Endometrial carcinoma)

# 유전단백체 기반 암 연구 동향

## 3. Assessing tumor immunogenicity

- proteomic evidence for immune status and neoantigen prediction



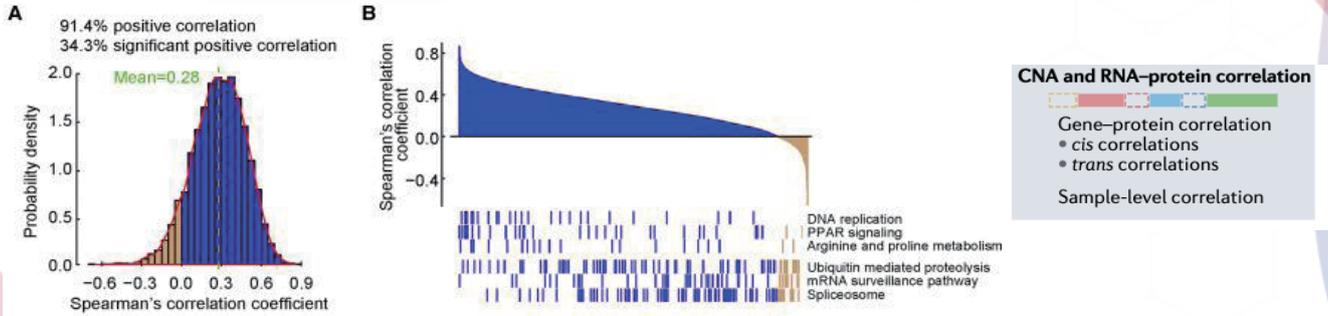
~50% of samples have evidence for putative neoantigens in endometrial carcinoma

Dou et al, Cell, 2020 (Endometrial carcinoma)

# 유전단백체 기반 암 연구 동향

## 4. mRNA-protein correlation

- identifying pathways showing high or low mRNA-protein correlation



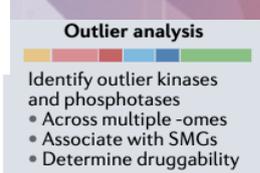
- In general, Pearson's correlation coefficient between mRNA and protein is 0.3-0.45 (~1/3 of genes show significant correlations)
- Genes in amino acid and fatty acid metabolism show good correlations while genes in house keeping pathways show poor correlations

Bhin and Mun et al, Cancer Cell, 2019 (Gastric cancer)

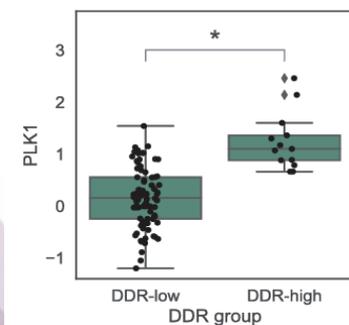
# 유전단백체 기반 암 연구 동향

## 5. Outlier analysis

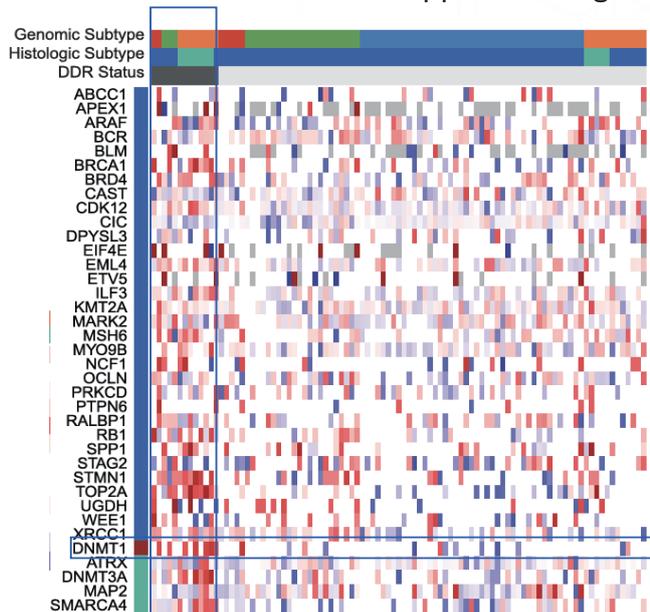
- identifying new potential therapeutic targets



Phosphorylation levels of known DNA damage response (DDR) markers  
→ DDR-low and DDR-high



DNMT1: FDA-approved drug target



Dou et al, Cell, 2020 (Endometrial carcinoma)

## 목차

- 유전단백체학 (Proteogenomics) 개념 소개
- 유전단백체학을 이용한 암 연구 동향
- 유전단백체학 분석 방법론/툴 소개

## 유전단백체 연구를 위한 분석툴

1. Data access
2. Generation of customized protein sequence database
3. Multi-omics-based patient subtyping
4. Correlation analysis between different molecular entities
5. Neoantigen prediction

# 유전단백체 연구를 위한 분석툴

## 1. Data access

Proteomics data from CPTAC and its partners: Proteomic Data Commons

<https://proteomic.datacommons.cancer.gov/pdc/>



# 유전단백체 연구를 위한 분석툴

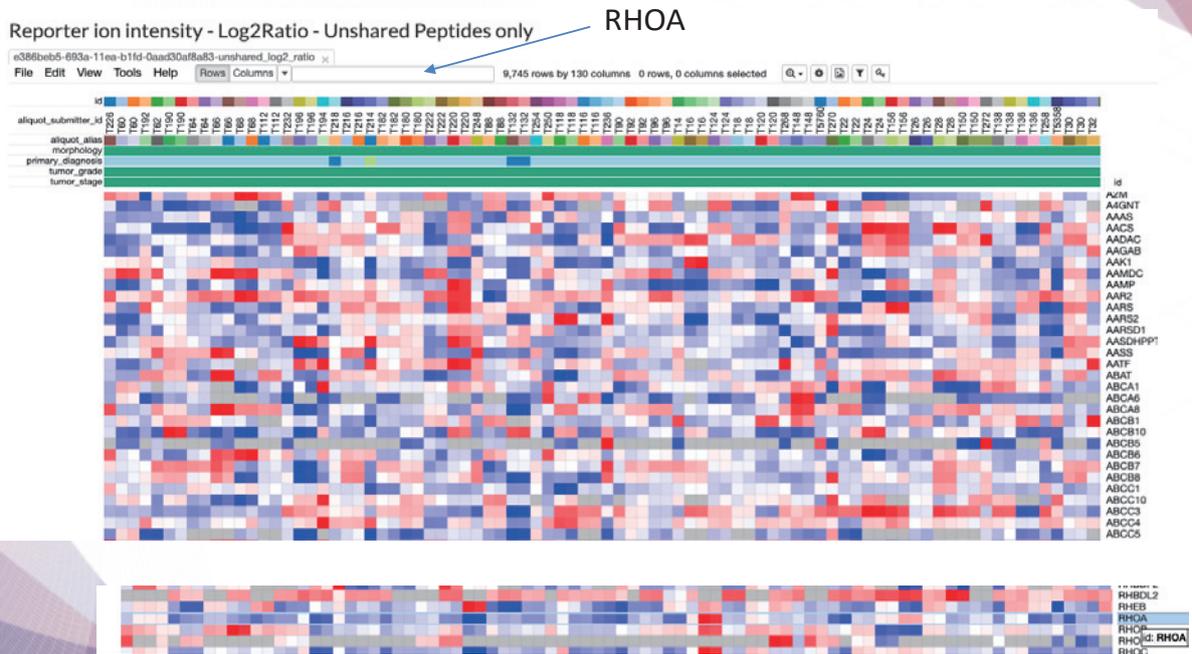
## 1. Data access

Proteomics data from CPTAC and its partners: **Proteomic Data Commons**

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# 유전단백체 연구를 위한 분석툴

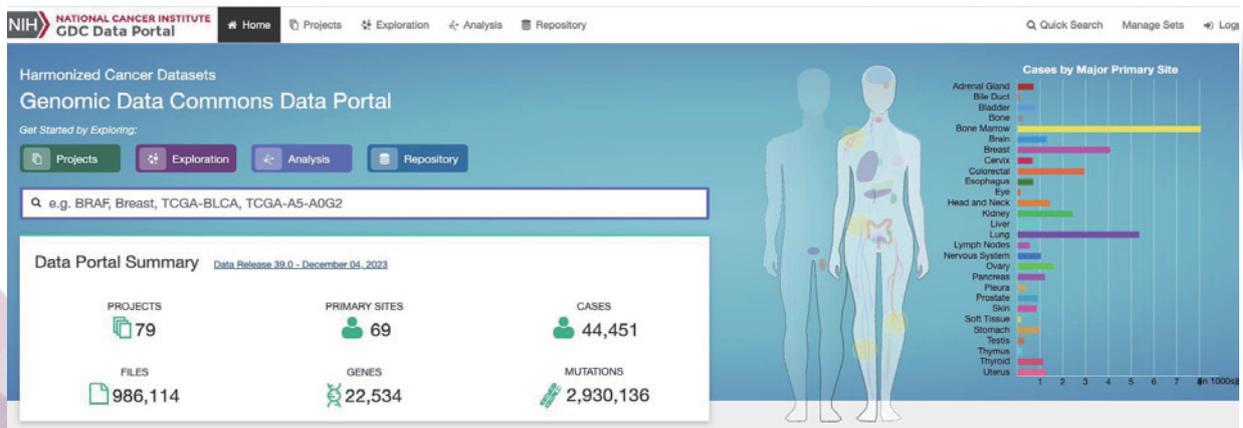


# 유전단백체 연구를 위한 분석툴

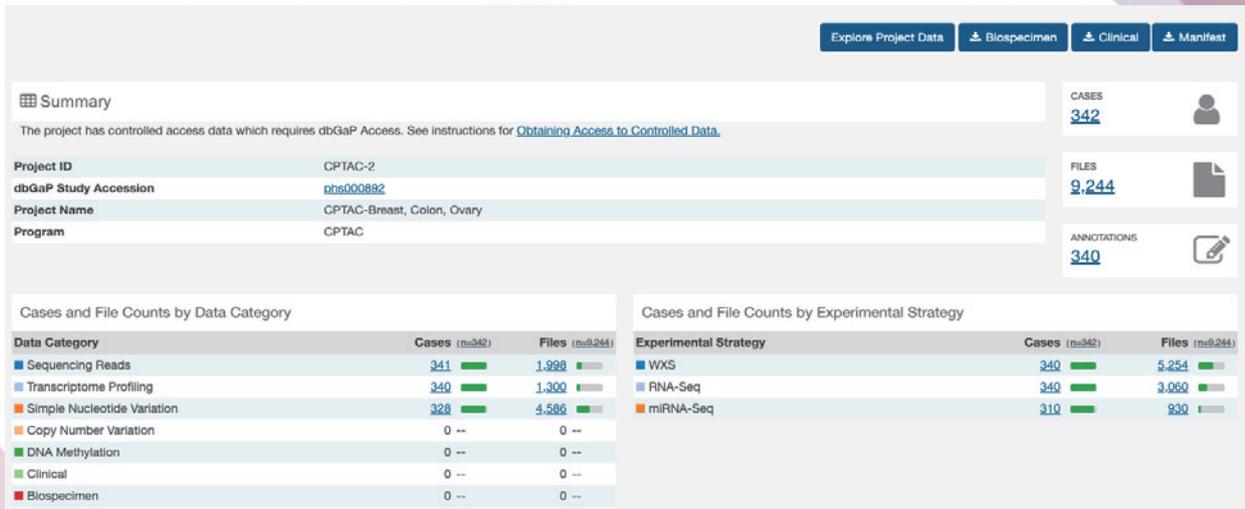
## 1. Data access

Genomics data from CPTAC: **Genomic data commons**

<https://portal.gdc.cancer.gov/>



# 유전단백체 연구를 위한 분석툴



# 유전단백체 연구를 위한 분석툴

## 1. Data access

Standardized multi-omic data matrices (TCGA, CPTAC)

<https://www.linkedomics.org/>

**LinkedOmics** Analyzing multi-omics data within and across 32 cancer types

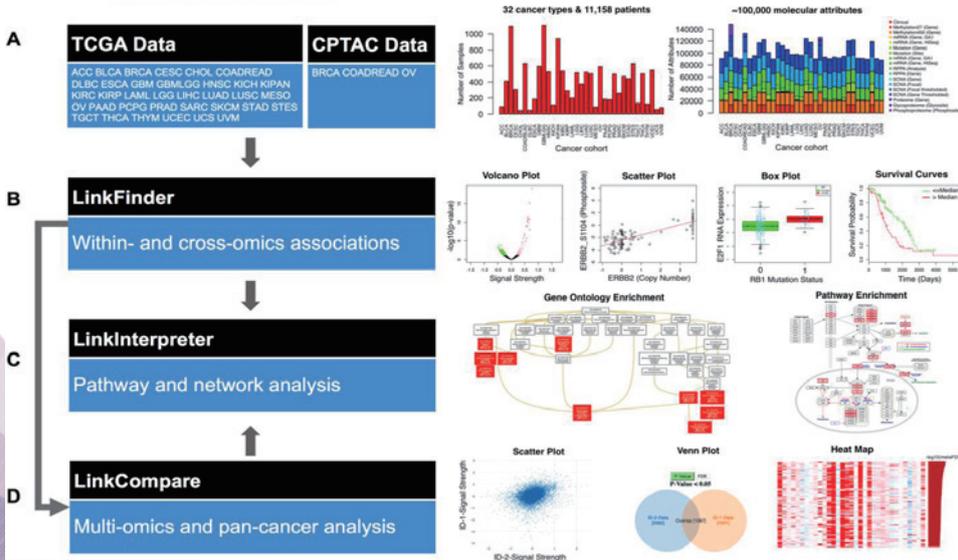
<ul style="list-style-type: none"> <li><a href="#">About</a></li> <li><a href="#">Login</a></li> <li><a href="#">Citation</a></li> <li><a href="#">OMICS Datatype</a></li> <li><a href="#">Data Source</a></li> <li><a href="#">Manual and tutorial</a></li> <li><a href="#">News/Update</a></li> <li><a href="#">Share Your Data</a></li> <li><a href="#">Browser Compatibility</a></li> </ul>	<p><b>ABOUT</b></p> <p>LinkedOmics is publicly available portal that includes multi-omics data from all 32 TCGA Cancer types and 10 Clinical Proteomics Tumor Analysis Consortium (CPTAC) cancer cohorts.</p> <p>The web application has three analytical modules: LinkFinder, LinkInterpreter and LinkCompare. LinkFinder allows users to search for attributes that are associated with a query attribute, such as mRNA or protein expression signatures of genomic alterations, candidate biomarkers of clinical attributes, and candidate target genes of transcriptional factors, microRNAs, or protein kinases. Analysis results can be visualized by scatter plots, box plots, or Kaplan-Meier plots. To derive biological insights from the association results, the LinkInterpreter module performs enrichment analysis based on Gene Ontology, biological pathways, network modules, among other functional categories. The LinkCompare module uses visualization functions (interactive venn diagram, scatter plot, and sortable heat map) and meta-analysis to compare and integrate association results generated by the LinkFinder module, which supports multi-omics analysis in a cancer type or pan-cancer analysis.</p> <p><i>LinkedOmics provides a unique platform for biologists and clinicians to access, analyze and compare cancer multi-omics data within and across tumor types.</i></p>
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# 유전단백체 연구를 위한 분석툴

## 1. Data access

Standardized multi-omic data matrices (TCGA, CPTAC)

<https://www.linkedomics.org/>



# 유전단백체 연구를 위한 분석툴

## 2. Customized protein sequence database

JOURNAL ARTICLE

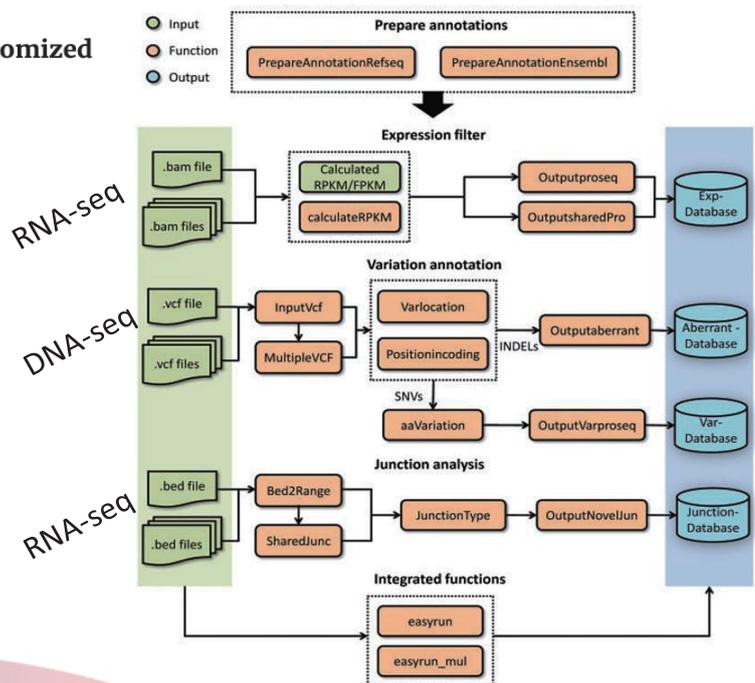
**customProDB: an R package to generate customized protein databases from RNA-Seq data for proteomics search**

Xiaojing Wang, Bing Zhang Author Notes

Bioinformatics, Volume 29, Issue 24, December 2013, Pages 3235–3237,

<https://doi.org/10.1093/bioinformatics/btt543>

Published: 20 September 2013 **Article history** ▾

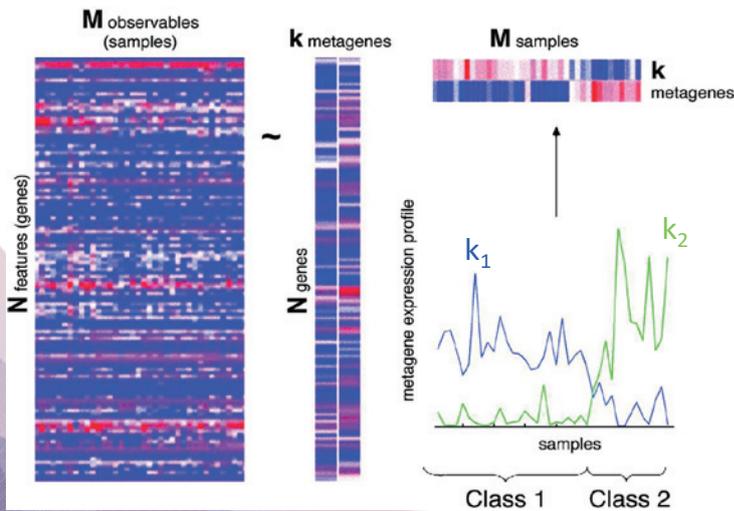


# 유전단백체 연구를 위한 분석툴

## 3. Multi-omics-based patient subtyping

- Non-negative matrix factorization (NMF) : for single omics data

$$A \text{ (rank } M) \sim W \quad H \text{ (rank } k=2)$$



A: non-negative values

$$A = W \times H$$

Objective function:

$$\sum (A - WH)^2;$$

Assign each sample  $m$  to class  $k$  that has the highest score

# 유전단백체 연구를 위한 분석툴

## 3. Multi-omics-based patient subtyping

- Non-negative matrix factorization (NMF) : for single omics data

- Consensus clustering - multiple iterations of the chosen clustering method on subsamples of the dataset
- By introducing sampling variability with sub-sampling, this provides metrics to assess the stability of the clusters

K = 2 (0: not sampled)

	iter1 ↕	iter2 ↕	iter3 ↕	iter4 ↕
Alice	2	0	2	1
Brian	1	1	1	2
Sally	1	2	0	2
James	0	2	1	0

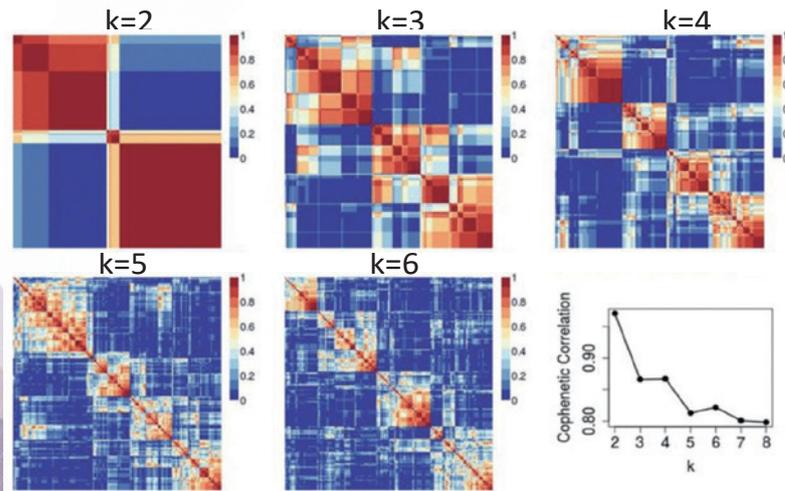
	Alice ↕	Brian ↕	Sally ↕	James ↕
Alice	1	0.00	0.00	0.0
Brian	0	1.00	0.67	0.5
Sally	0	0.67	1.00	1.0
James	0	0.50	1.00	1.0

# 유전단백체 연구를 위한 분석툴

## 3. Multi-omics-based patient subtyping

- Non-negative matrix factorization (NMF) : for single omics data

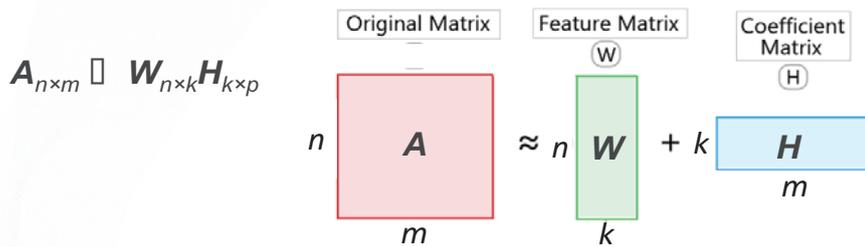
Consensus NMF clustering



# 유전단백체 연구를 위한 분석툴

## 3. Multi-omics-based patient subtyping

- Integrative Non-negative matrix factorization (NMF)



Modification for integrative clustering

$$A_{n \times m}^i \approx W_{n \times k}^i \times H \quad i : \text{data type } (i = 1, 2, \dots, p)$$

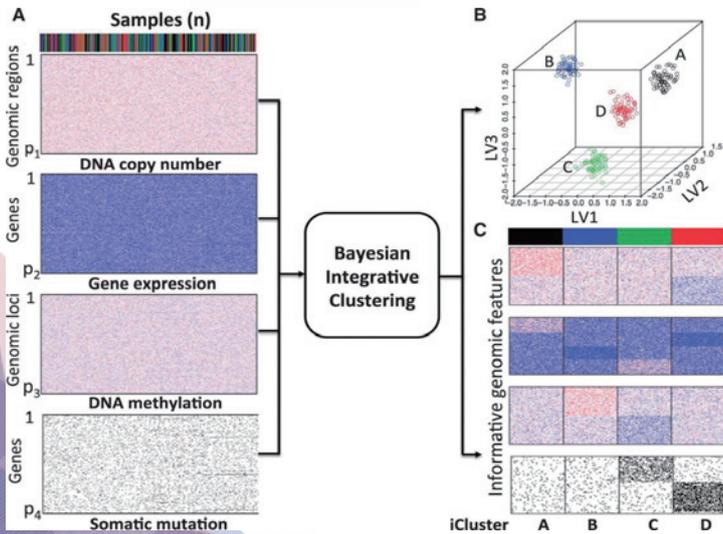
$k \times m$

$$Q = \min_{W, H} \sum_{i=1}^p \theta^i \|X^i - W^i H\|_2$$

# 유전단백체 연구를 위한 분석툴

## 3. Multi-omics-based patient subtyping

- iCluster → iClusterPlus → iClusterBayes



- Joint integrative clustering to model continuous and discrete omics data
- Reduce multi-high dimensional space to a low dimensional subspace that will collectively capture the major variations of the multiple datasets  
→ low dimensional subspace can be used to cluster the samples
- Simultaneous identification of features that contribute to sample clustering

Mo et al, Biostatistics, 2018

# 유전단백체 연구를 위한 분석툴

## iClusterPlus: integrative clustering of multiple genomic data sets

Qianxing Mo<sup>1</sup> and Ronglai Shen<sup>2</sup>

October 24, 2023

<sup>1</sup>Department of Biostatistics & Bioinformatics  
H. Lee Moffitt Cancer Center & Research Institute  
qianxing.mo@moffitt.org

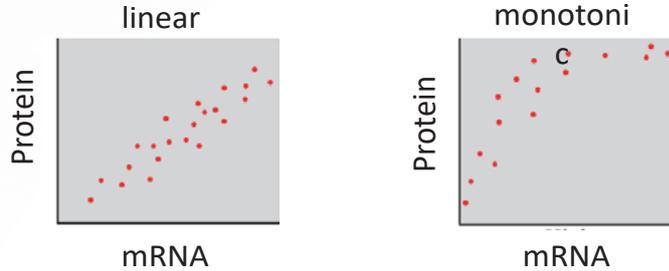
<sup>2</sup>Department of Epidemiology and Biostatistics  
Memorial Sloan-Kettering Cancer Center  
shenr@mskcc.org

R package for iClusterPlus and iClusterBayes

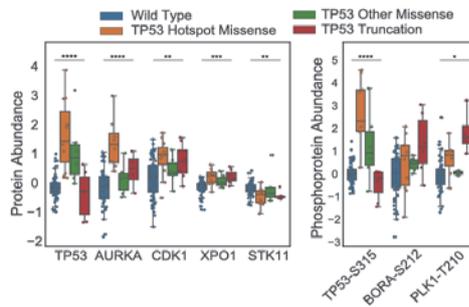
# 유전단백체 연구를 위한 분석툴

## 4. Correlation analysis between different molecular entities

- Use Pearson (linear relationship) or Spearman (monotonic relationship)



- t-test (parametric) or Mann-Whitney U test (Non-parametric)

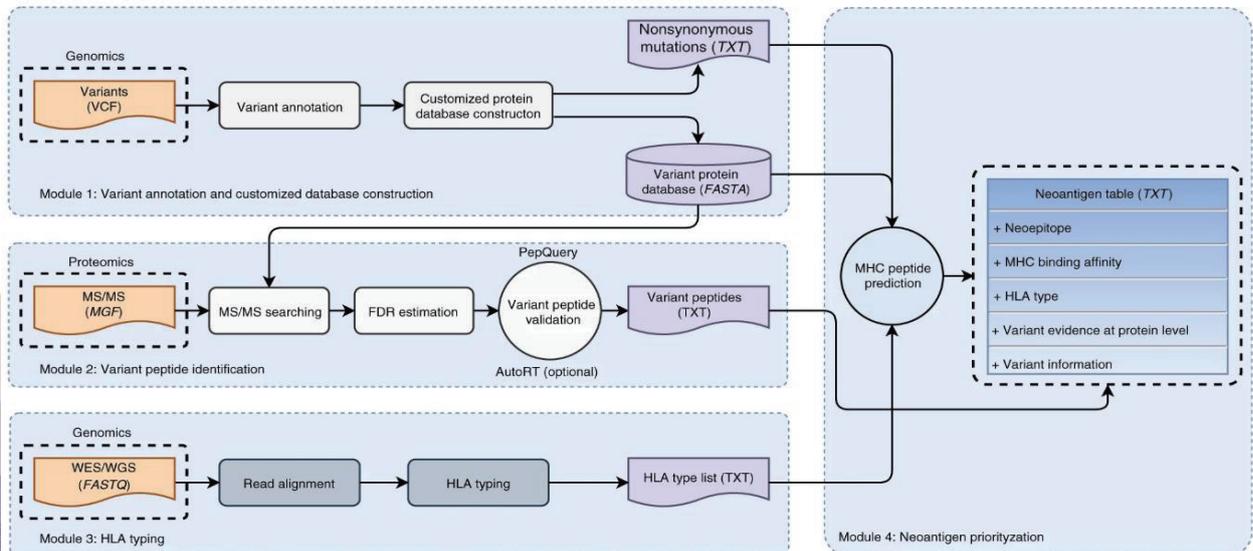


Dou et al, Cell, 2020 (Endometrial carcinoma)

# 유전단백체 연구를 위한 분석툴

## 5. Neoantigen prediction

### NeoFlow



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**Cancer neoantigen prioritization through sensitive and reliable proteogenomics analysis**

Bo Wen<sup>1,2</sup>, Kai Li<sup>1,2</sup>, Yun Zhang<sup>1,2</sup> & Bing Zhang<sup>1,2\*</sup>

# 유전단백체 연구를 위한 툴

## Data analysis — association discovery and visualization

- **LinkedOmics**: Interactive visualization and analysis of multi-omics data from TCGA and CPTAC within and across cancer types<sup>155</sup>
- **multiOmicsViz**: Visualizing the effect of one omics data type on other omics data along the chromosome<sup>27</sup>
- **iProFun**: Characterization of functional consequences of copy number alterations (CNAs) and methylation alterations in tumours<sup>99</sup>
- **ProNetView**: Network visualization inferred from proteogenomic data of the CPTAC-CCRCC project
- **NetSAM**: Network seriation and modularization<sup>164</sup>
- **NetGestalt**: Visualization and analysis of biological networks from multi-omics data<sup>164</sup>
- **Panoptes (1), Panoptes (2)**: Analysing pathological images using deep learning
- **TSNet**: A statistical model for cell type-specific inference based on bulk tumour profiling data<sup>165</sup>
- **Perseus**: Bioinformatics platform for integrative analysis of proteomic data<sup>166,167</sup>

## Data analysis — PTM and pathway analysis

- **PTMsigDB**: Collection of PTM site-specific signatures<sup>102</sup>
- **PTM-SEA**: Modified version of ssGSEA for PTMsigDB<sup>102</sup>
- **WebGestalt**: GSEA toolkit with support for phosphosite enrichment analysis<sup>170</sup>
- **PTMcosmos**: Database of PTMs and cancer mutations in humans
- **PHOTON**: Delineating signalling pathways from large-scale phosphoproteomic data<sup>171</sup>
- **HotPho**: Identification of spatially interacting phosphosites and mutations<sup>172</sup>
- **Black Sheep**: Identification of samples with aberrant gene, protein or PTM site abundances<sup>173</sup>
- **WikiPathways**: Database of molecular pathway diagrams contributed and refined by the research community<sup>174</sup>

Mani et al, Nat Rev Cancer, 2022

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