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Research Article

Genomics and Ossc Cure

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Highlights

Multi-omics reveals molecular complexity of OSCC beyond genomics alone
Integration of proteomics and metabolomics improves biomarker discovery
Enables precision oncology and rational drug combination strategies
Identifies mechanisms of therapeutic resistance

Abstract

Oral Squamous Cell Carcinoma (OSCC) is a highly heterogeneous malignancy characterized by complex molecular alterations that cannot be fully understood through single-omics approaches. Multi-omics integration, encompassing genomics, proteomics, and metabolomics, provides a comprehensive framework for understanding tumor biology and therapeutic vulnerabilities. Genomics identifies driver mutations, proteomics elucidates functional protein networks, and metabolomics captures metabolic reprogramming essential for tumor survival. Integrative analysis enables improved biomarker discovery, identification of novel drug targets, and prediction of therapeutic response. "This paper presents a theoretical analysis of literature-derived data to examine advances in multi-omics approaches in Oral Squamous Cell Carcinoma and their implications for Targeted therapy, Immunotherapy and drugs resistance."

Keywords: Ossc, Multi-Omics, Genomics, Proteomics, Metabolomics, Targeted Therapy, Precision Oncology

Introduction

Oral Squamous Cell Carcinoma (OSCC) accounts for the majority of oral malignancies and remains associated with poor prognosis due to late-stage diagnosis and therapy resistance. Traditional molecular studies have largely focused on genomic alterations; however, these do not fully capture the dynamic functional state of tumor cells.

Multi-omics integration enables:

Cross-validation of molecular findings
Functional interpretation of genomic alterations
Systems-level modeling of tumor progression

Multi-Omics Framework in OSCC

Description:

A layered schematic:

Layer 1: Genomics (DNA mutations, CNVs)
Layer 2: Proteomics (protein expression, signaling pathways)
Layer 3: Metabolomics (metabolic flux, metabolites)
Output: Biomarkers → Drug targets → Personalized therapy

Genomics Landscape of OSCC

Key Genetic Alterations

TP53 mutations (genomic instability)
PI3K/AKT pathway activation

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NOTCH signaling alterations

Limitations of Genomics Alone

- Poor correlation with protein activity
- Inability to predict metabolic phenotype
- Limited utility in therapy selection without integration

Proteomics: Functional Layer of Tumor Biology

Protein Expression and Signaling

Proteomics identifies dysregulated pathways:

- EGFR signaling
- MAPK cascade
- PI3K/AKT/mTOR axis

Clinical Applications

- Drug target validation
- Predictive biomarkers
- Therapy monitoring

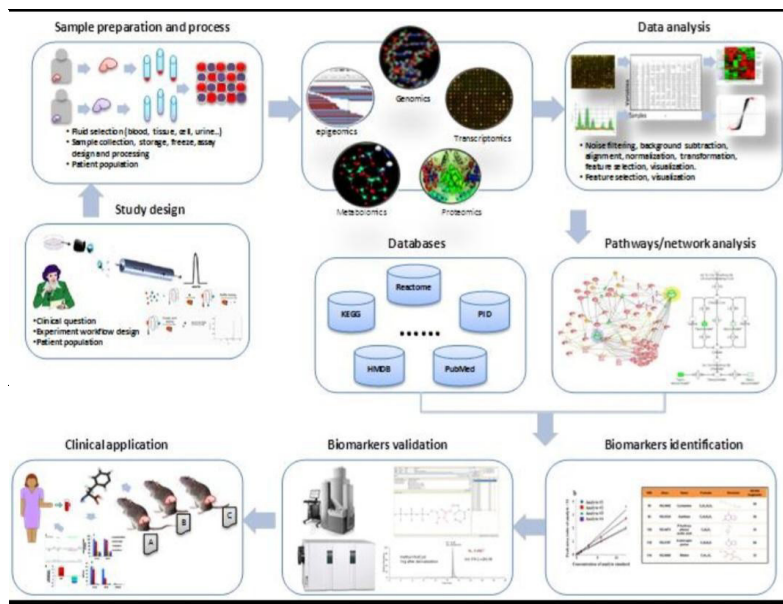
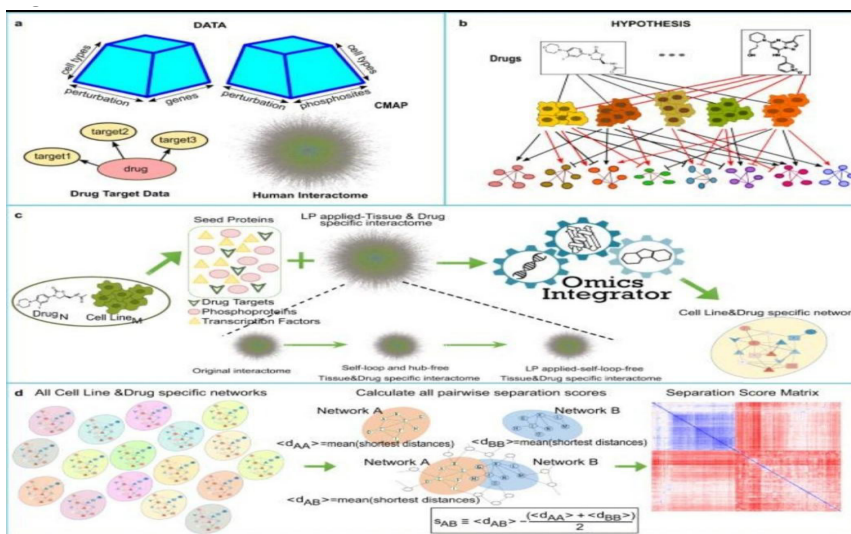


Figure: 1 and 2



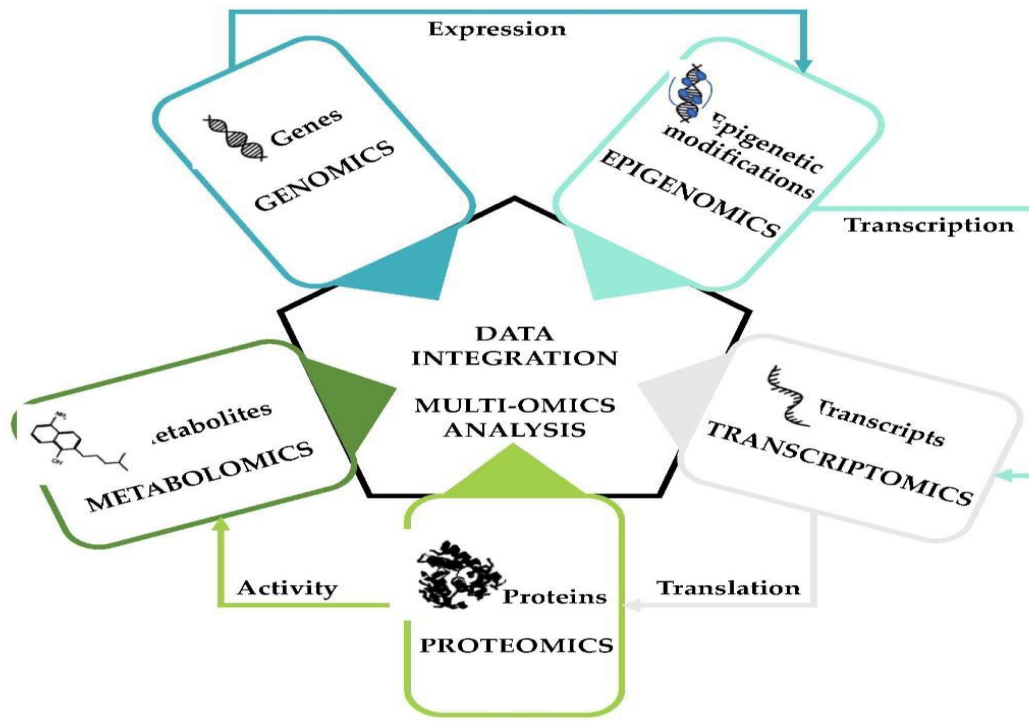


Figure 3

This figure illustrates the hierarchical integration of:
 Genomics (DNA mutations, CNVs)
 Proteomics (protein expression, signaling pathways)
 Metabolomics (metabolic flux and metabolites)
 These layers converge into: → Biomarker discovery
 → Drug target identification
 → Precision therapy

Dysregulated signaling pathways in OSCC

Chart Type: Pathway interaction network

Content: Nodes representing proteins (EGFR, AKT, mTOR) with edges showing activation/inhibition

Metabolomics: Tumor Phenotype and Adaptation

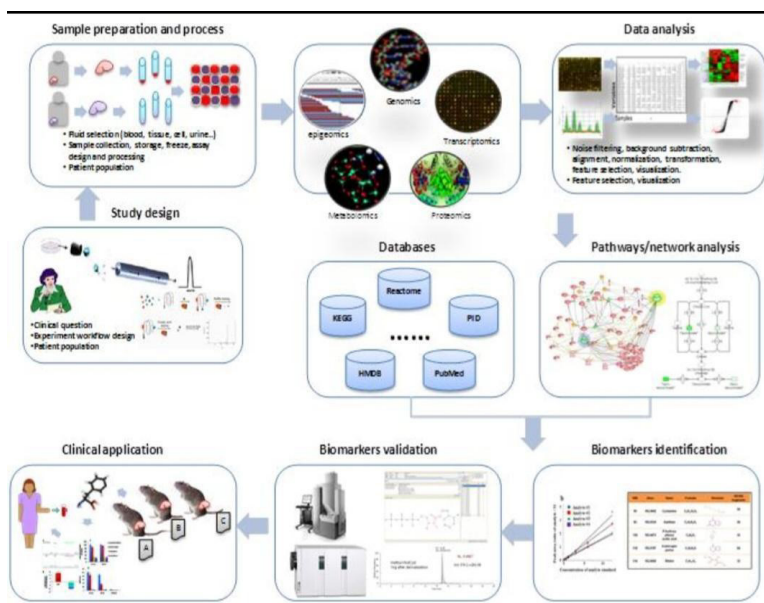
Metabolic Reprogramming

Aerobic glycolysis (Warburg effect)

Glutamine dependency

Lipid biosynthesis

Therapeutic Implications



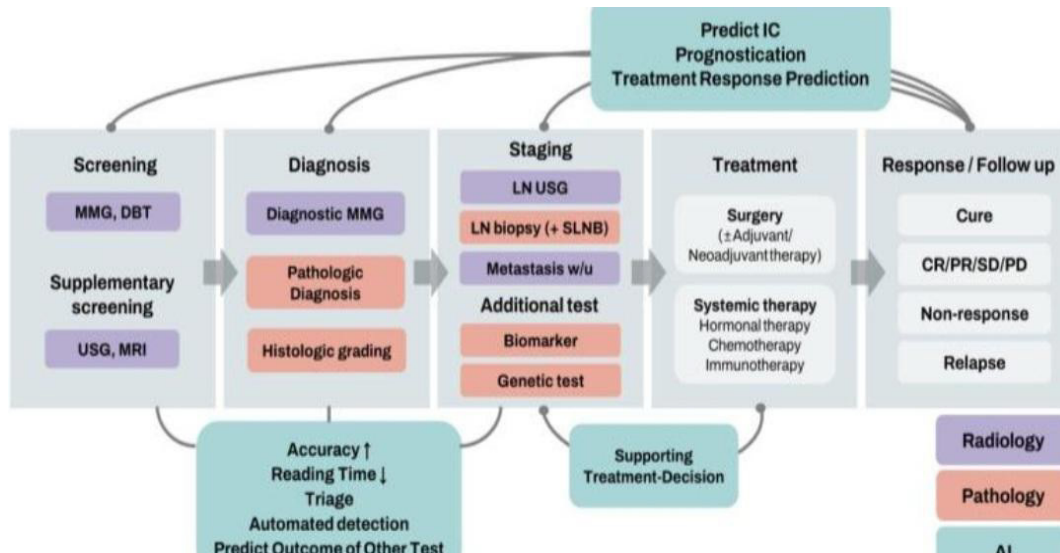


Figure 4 and 5

Targeting metabolic enzymes
 Monitoring treatment response via metabolite profiles Figure 4 (Chart)
 Metabolic alterations in OSCC
 Chart Type: Bar graph or heatmap

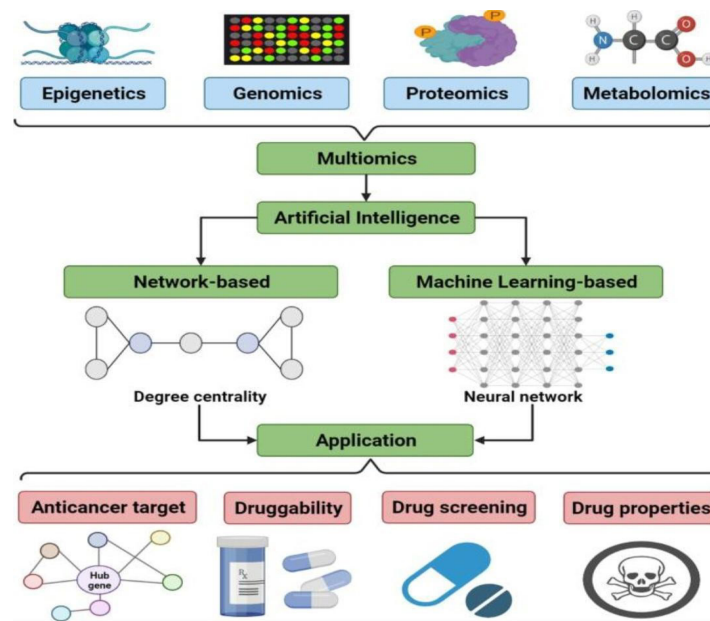


Figure 6

Content: Elevated lactate, glutamine, lipid metabolites vs normal tissue

**Integration of Multi-Omics Data
 Systems Biology Approach**

Integration methods:
 Network-based modeling
 Machine learning algorithms
 Pathway enrichment analysis

Clinical Benefits

Improved diagnostic accuracy
 Identification of molecular subtypes
 Personalized therapy design

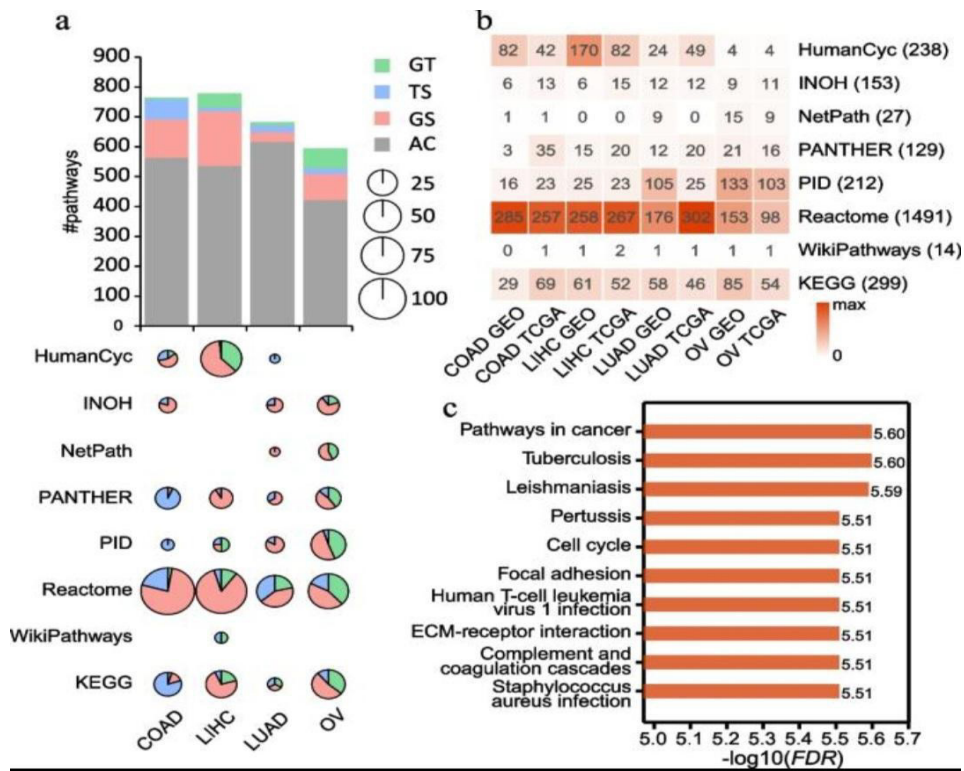


Figure 7

Multi-omics-driven precision medicine pipeline

Flow: Patient sample → Multi-omics profiling → Data integration → Target identification

→ Drug selection

Implications for Cancer Drug Development Targeted Therapy

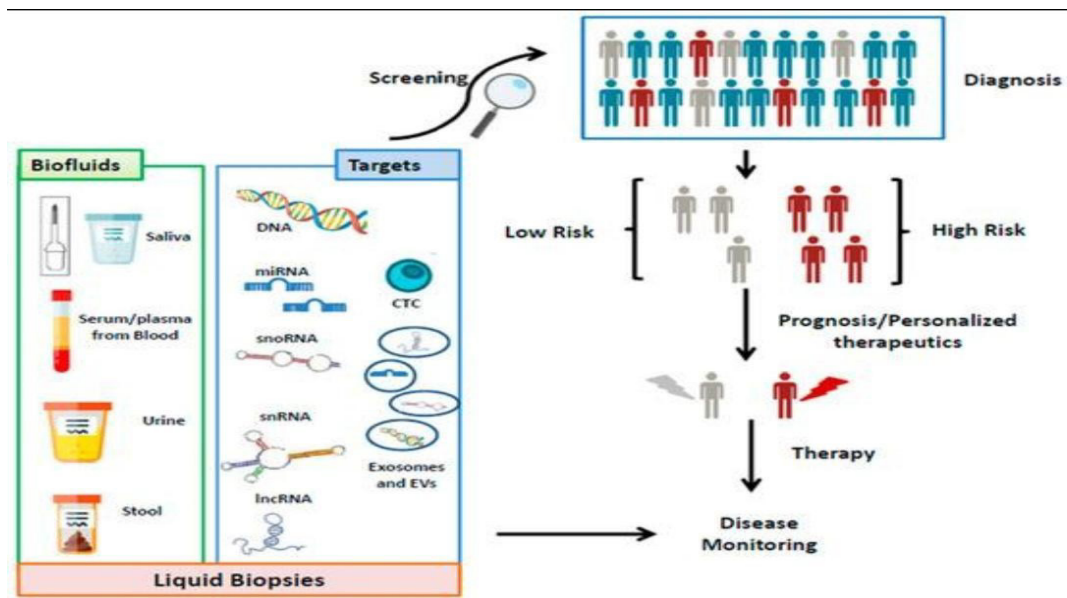


Figure 8

Multi-Omics Identifies Actionable Targets:

EGFR inhibitors (e.g., cetuximab)

PI3K inhibitors

CDK inhibitors

Immunotherapy

PD-1/PD-L1 axis modulation

Tumor microenvironment profiling via proteomics

Metabolism-Based Therapies

Glycolysis inhibitors

Glutaminase inhibitors

Multi-omics-guided drug targeting strategy

Chart Type: Integrated pathway map showing drug intervention points

Drug Resistance Mechanisms

Multi-omics reveals:

Genetic mutations driving resistance

Protein pathway reactivation

Metabolic adaptation

Clinical Impact

Predicting resistance early

Designing combination therapies

Challenges and Limitations

Data integration complexity

High computational requirements

Limited clinical translation

Cost constraints

Future Perspectives

AI-driven multi-omics integration

Single-cell omics technologies

Liquid biopsy applications

Real-time treatment monitoring

Conclusion

Multi-omics integration represents a transformative approach in understanding OSCC biology. By bridging genomic alterations with functional and metabolic phenotypes, it enables more accurate biomarker discovery and targeted therapeutic strategies. This systems-level approach is essential for advancing precision oncology and improving patient outcomes [1-10].

Declarations

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Conflicts of Interest: The authors declare no conflict of interest

Ethical Approval: Not applicable

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