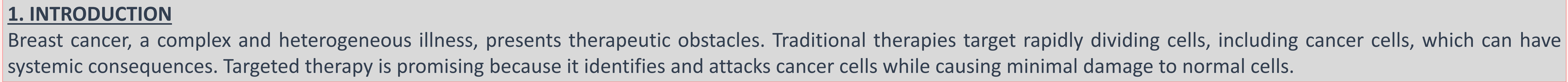


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### 1.1. Why FGFR (Fibroblast Growth Factor Receptor)?

## 1.2. What are we doing?

```

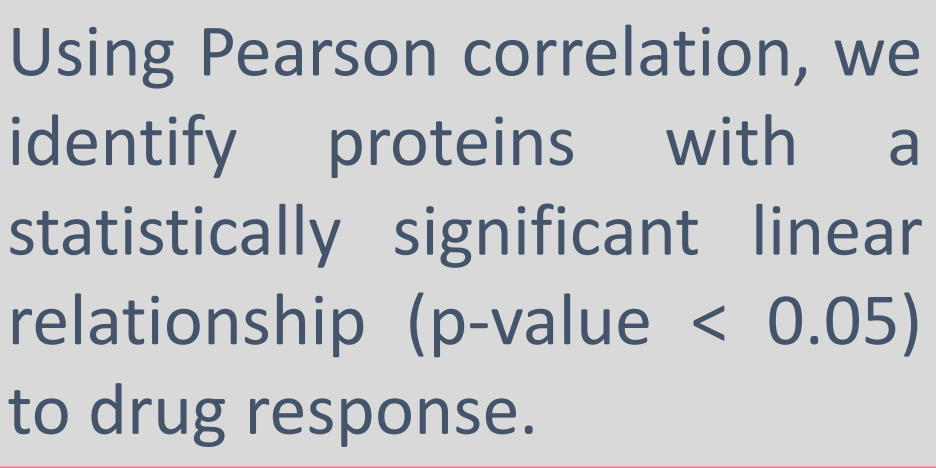
graph LR
    A[Data] --> B[Pre-processing]
    B --> C[Feature selection]
    C --> D[Model selection  
Model training]
    D --> E[Marker identification]
    E --> F[Marker optimization]
    F --> G[Bioinformatic analysis]
    
    B --- B1["▪ Melting input and target data  
▪ Imputation  
▪ Bootstrapping  
▪ Normalization"]
    C --- C1["▪ A range of ML models (e.g. XGBoost, SVM, ANN to select 2 best models on default parameters  
▪ Select Best model after hyperparameter tuning"]
    E --- E1["▪ Based on the selected model, select the 20 best features using Wrapper type FS (Forward Feature Selection and Backward Feature Elimination)"]
    F --- F1["▪ Identification of the robust biomarkers by analysing the features selected from all the bootstrapped samples."]
    G --- G1["▪ Bioinformatic analysis for the selected features.  
▪ E.g., pathway enrichment, gene ontology analysis, etc  
▪ Biological validation of the selected features."]
  
```

**1.3. WORKFLOW**

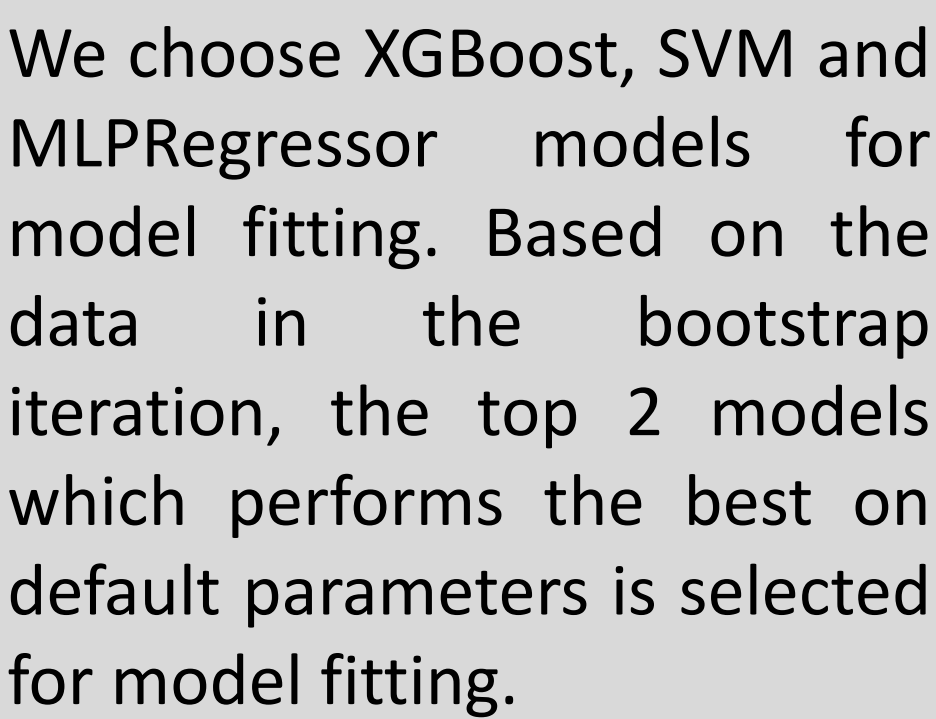
## 2.1. Data and Pre-Processing

- ```

graph TD
    A[Bootstrapped Data  
(6693 features)] --> B[Pearson Correlation  
(1000-2000 features)]
    B --> C[mRMR  
(500 Best features)]
    B --> D[F-Regression  
(500 Best features)]
    C --> E[Intersection of features from mRMR and  
F-Regression  
(100-400 features)]
    D --> E
    E --> F[Lasso Regularisation  
(40-150 features)]
    F --> G[ML Model Fitting]
    G --> H[Backward Feature Elimination  
(20 best features)]
    G --> I[Feature Forward Selection  
(20 best features)]
    H --> J[Union of Features from  
FFS and BFE]
    I --> J
  
```



Utilizing multiple FS methods ensured a nuanced identification of robust biomarkers.



The best model after hyperparameter tuning is selected for FFS and BFE (SVM for this particular iteration)

## Enrichment analysis of genes derived from different stages of the feature selection pipeline for a particular bootstrap iteration

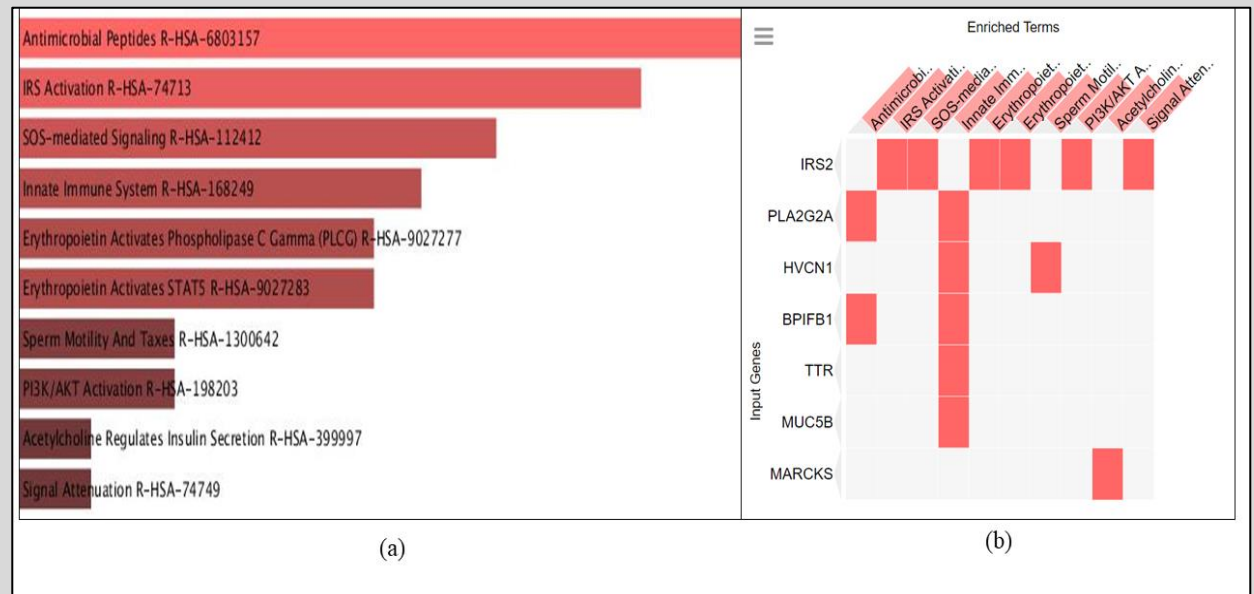
- [illegible]

- Embedded feature selection methods resulted in 89 genes. Highly enriched pathways included **Signalling By Membrane-Tethered Fusions Of PDGFRA Or PDGFRB**, and **Oestrogen Response Early**, both implicated in breast cancer pathologies.

Enrichment analysis of genes amassed from bootstrap iterations. Genes were selected based on their frequency of occurrence in all the iterations.

| Frequency of Occurrence | Number of Genes | Enriched Pathways which can be Potential Target in Breast Cancer Therapy                                      |
|-------------------------|-----------------|---------------------------------------------------------------------------------------------------------------|
| No filter               | 363             | RNA metabolism, mRNA splicing, and fatty acid metabolism                                                      |
| >100                    | 168             | Membrane trafficking pathways and E2F pathway                                                                 |
| >150                    | 50              | IRS proteins, Unfolded Protein Response (UPR), and Estrogen Response Late                                     |
| No filter               | 25, 20,10       | Antimicrobial Peptides, TNF-alpha signaling, PI3K/AKT activation pathway, and IL-6/JAK/STAT3 signaling system |

## Enrichment analysis of top 25 genes

[illegible]

The examination of gene sets via Enrichr revealed multiple enriched pathways with potential implications in breast cancer target therapy. The recurrent appearance of genes **PLA2G2A**, **BPIFB1**, and **IRS2** across enriched pathways underscores their potential significance in targeted breast cancer therapies with AZD4547 FGFR inhibitor. The different enriched pathways found across different versions of the gene set show how biological processes interact in breast cancer in a complicated way. The results give us a strong base for more research into targeted therapeutic methods, which could lead to more successful, individualised treatments for breast cancer.