

# Temporal Knowledge Acquisition from Cancer Clinical Trials

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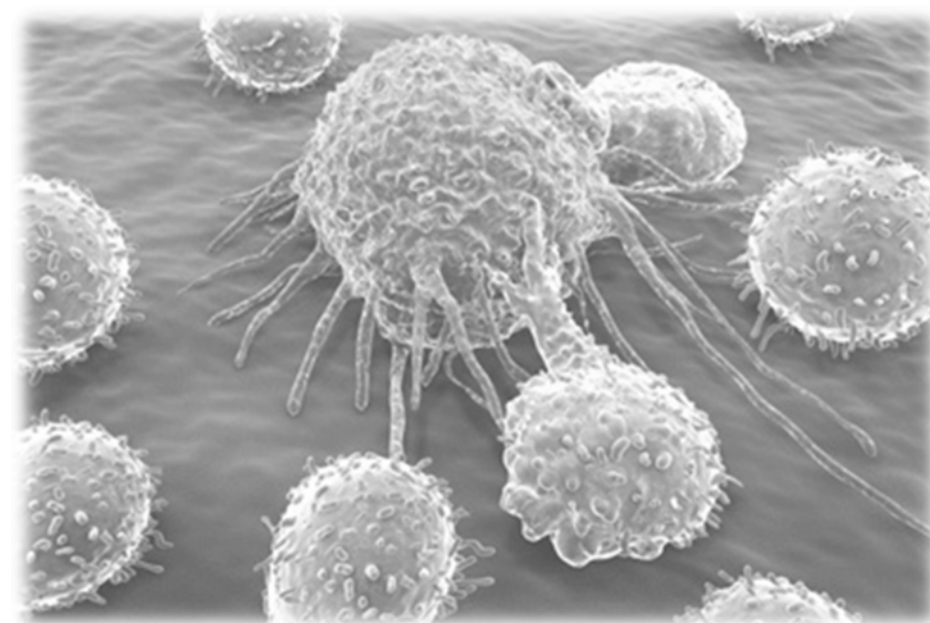
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# Background

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- **Many review articles have summarized the progress in specific domain of cancer research**
- **Prostate cancer and Breast cancer**
- Diagnostic and therapeutic approaches have improved steadily since late 20<sup>th</sup> century
- **Pancreatic cancer**
- Limited medication options



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# Background

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## ***Limitations:***

1. Only qualitative description of the cancer milestones based on domain experts' experience, which limited the potential of using this temporal knowledge.

e.g. Lack of parallel comparison.

## ***Opportunity:***

1. Important progress in treatment of cancer arises from the results of rigorously conducted clinical trials

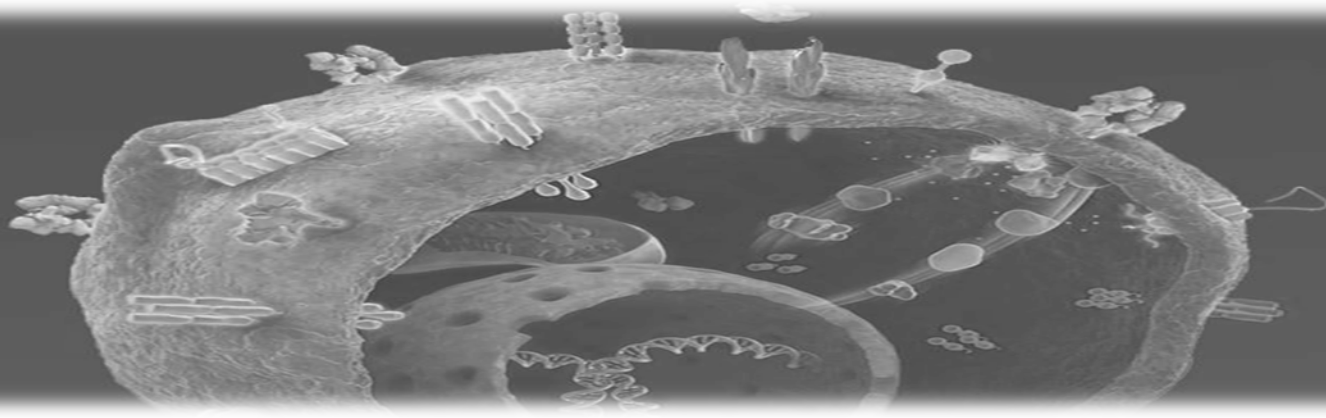
2. Database of cancer clinical trials enables efficient and scalable knowledge acquisition

3. We may identify cancer treatment progress by analyzing original longitudinal data of clinical trials

# Goal

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- **Identify** anti-cancer medications that trend upwards or downwards
- **Show** cancers have similar or different trends of medication interventions in a quantitative way
- **Demonstrate** database of clinical trials serves as a valid source for supporting trend analysis and creating temporal knowledge



# Method

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## Data Retrieval

Trial information retrieval from ClinicalTrials.gov



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graph TD; A[Trial information retrieval from ClinicalTrials.gov] --> B[Selected "interventional" clinical trials with purpose of "treatment"]; B --> C[Formulate oncology clinical trial dataset by selecting clinical trials with condition concepts under a semantic type of neoplastic process];
```

The diagram illustrates a three-step process for data retrieval. It begins with a box containing the text 'Trial information retrieval from ClinicalTrials.gov'. A large, light purple downward-pointing arrow connects this box to the second box, which contains the text 'Selected "interventional" clinical trials with purpose of "treatment"'. Another similar arrow connects the second box to the third box, which contains the text 'Formulate oncology clinical trial dataset by selecting clinical trials with condition concepts under a semantic type of neoplastic process'.

Selected “interventional” clinical trials with purpose of “treatment”

Formulate oncology clinical trial dataset by selecting clinical trials with condition concepts under a semantic type of neoplastic process

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# Method

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## Terminologies

- Conditions were mapped to MeSH terms.
- Medications were mapped to NCI drug dictionary (NCI thesaurus code)

# Method

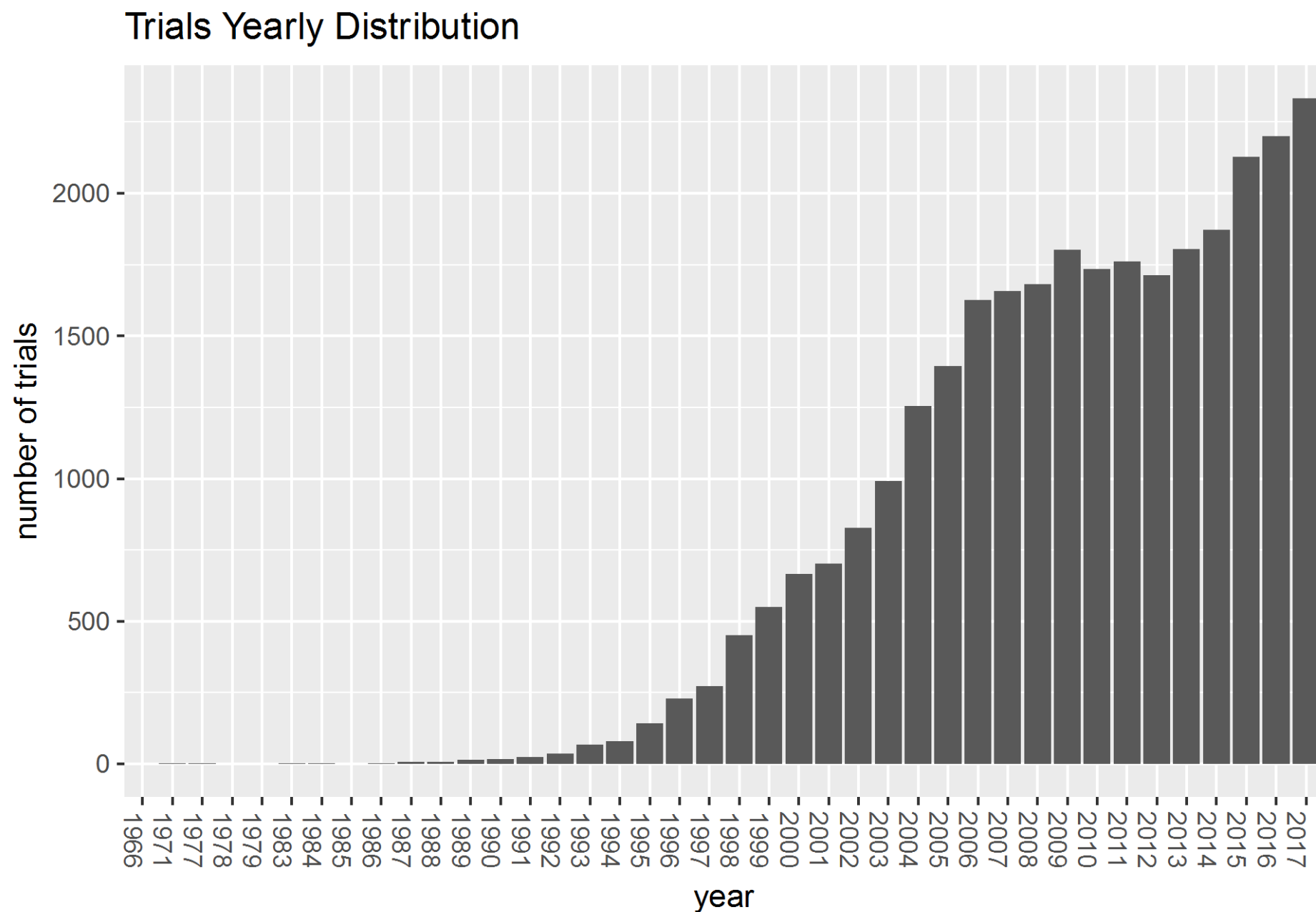
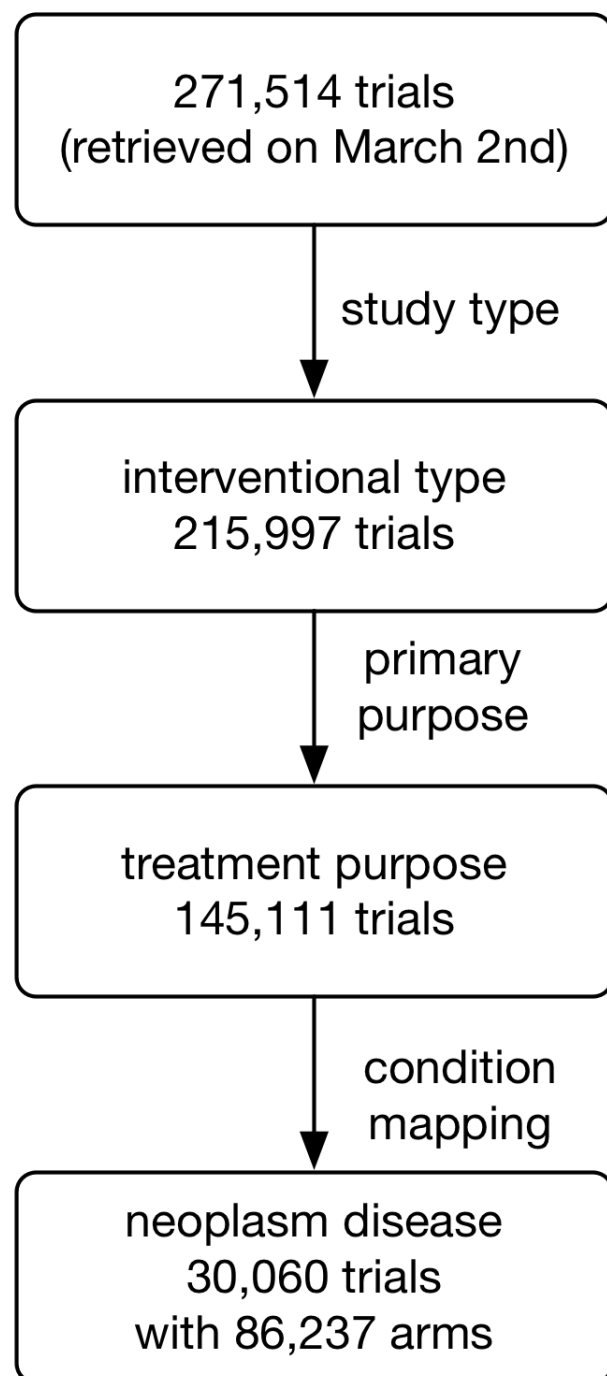
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## Data Preprocessing

- Classify medication into 4 groups based on the semantic type of ingredients (conventional chemical drugs, hormone, immunologic factors and bioactive substance)
- Normalized Frequency(NF): in a period of time, the number of trial arms of interested drug(or drug combinations) divided by the total number of trial arms.

# Result

## Trials identification

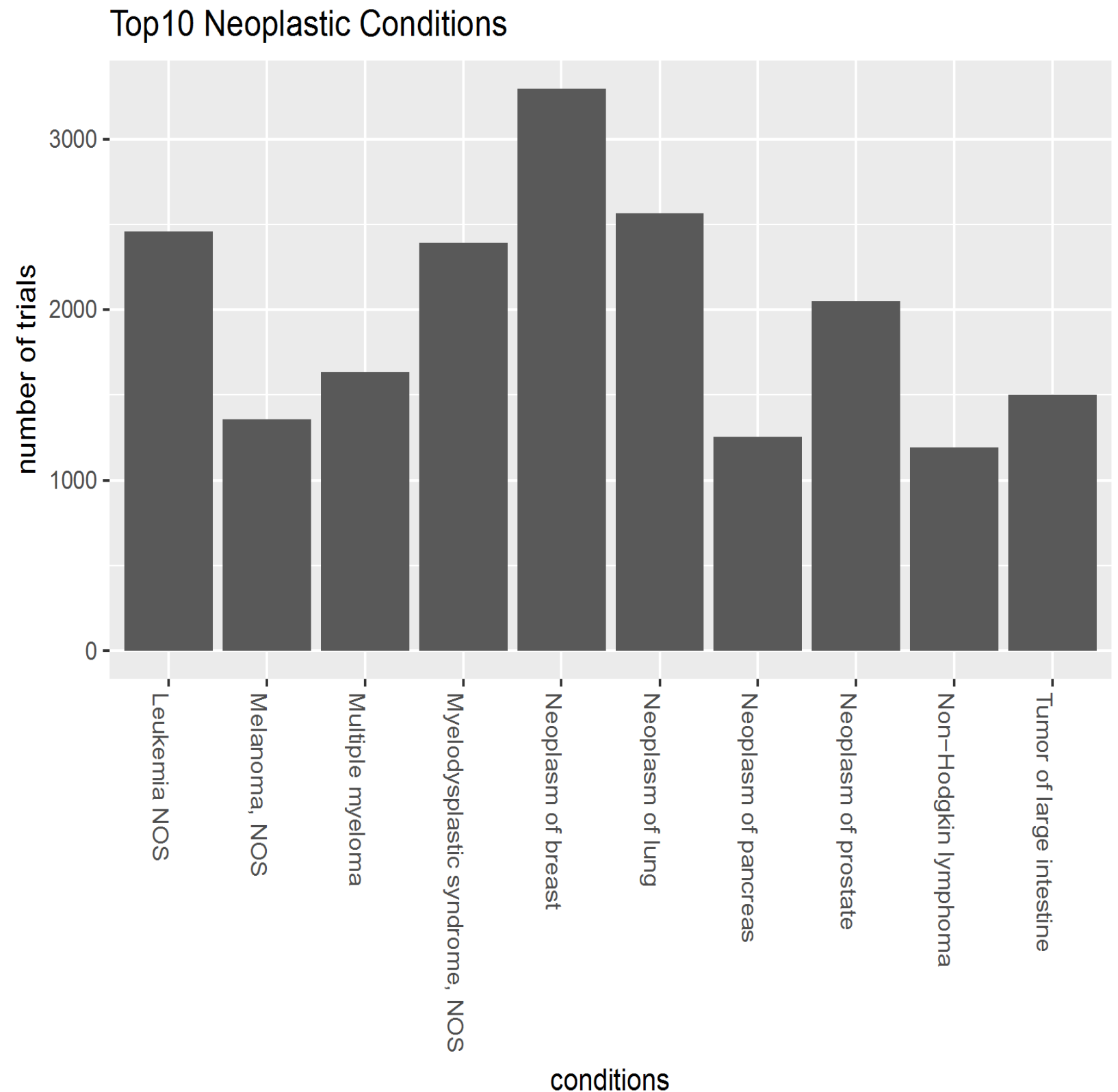




# Result

## Condition identification

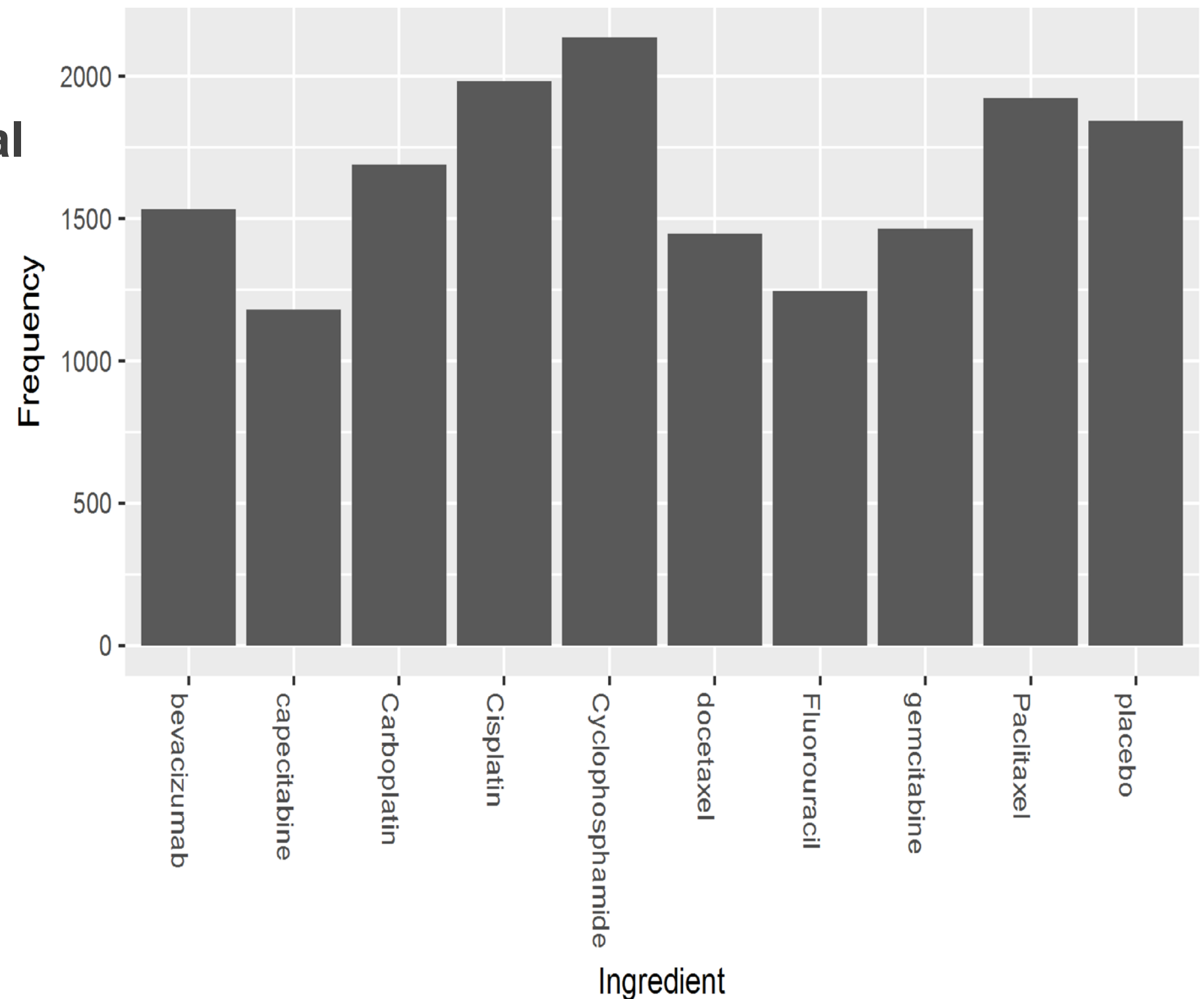
**Identified 260  
condition concepts  
under the semantic  
type of neoplastic  
process after  
manually review**



# Result

## Medication identification

Top10 Ingredients in cancer clinical trials

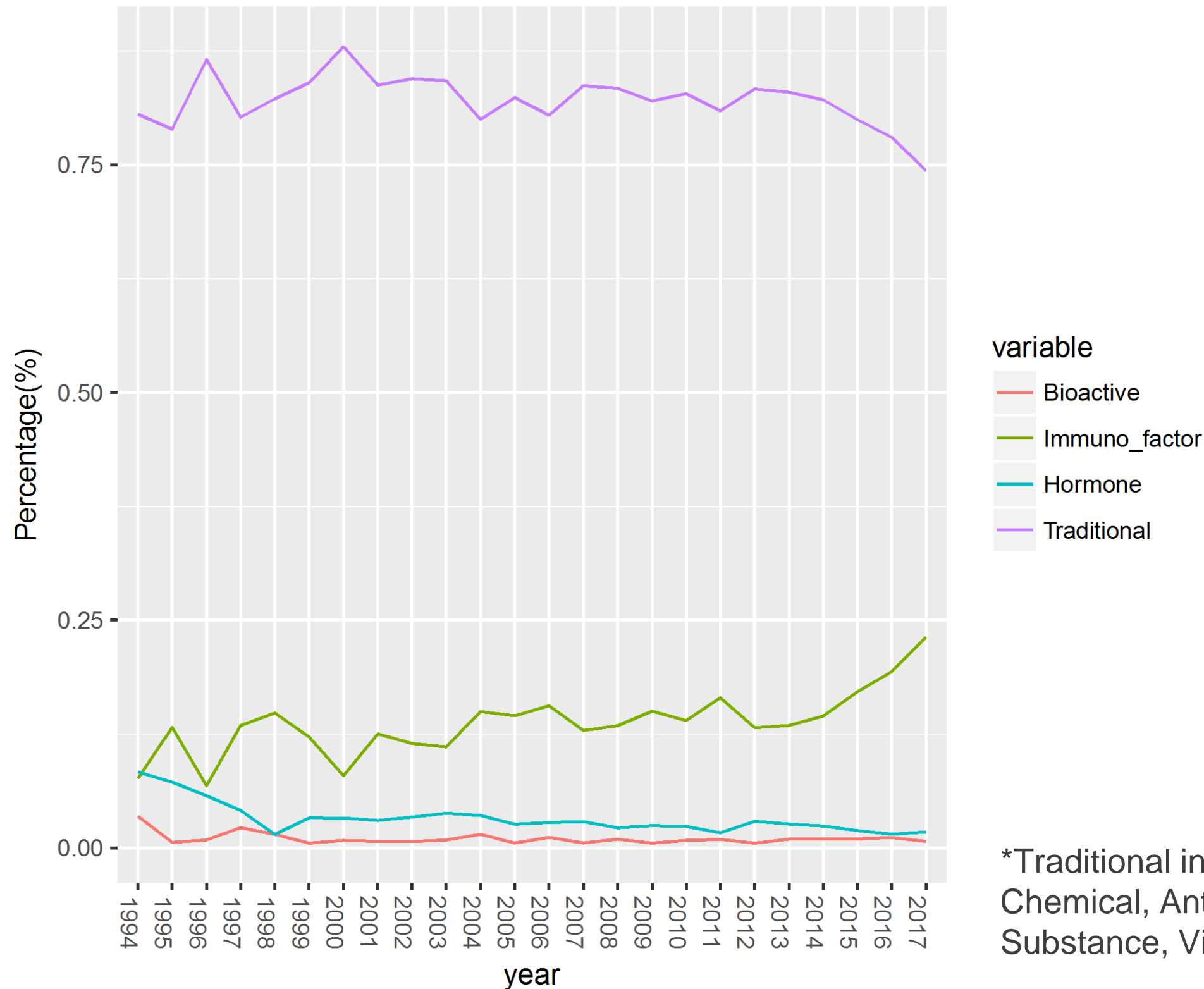


**Identified 1035  
medications in total**

# Result

## Anti-cancer medications trend

Drug Type Trends



### Top10 immunologic factor medications:

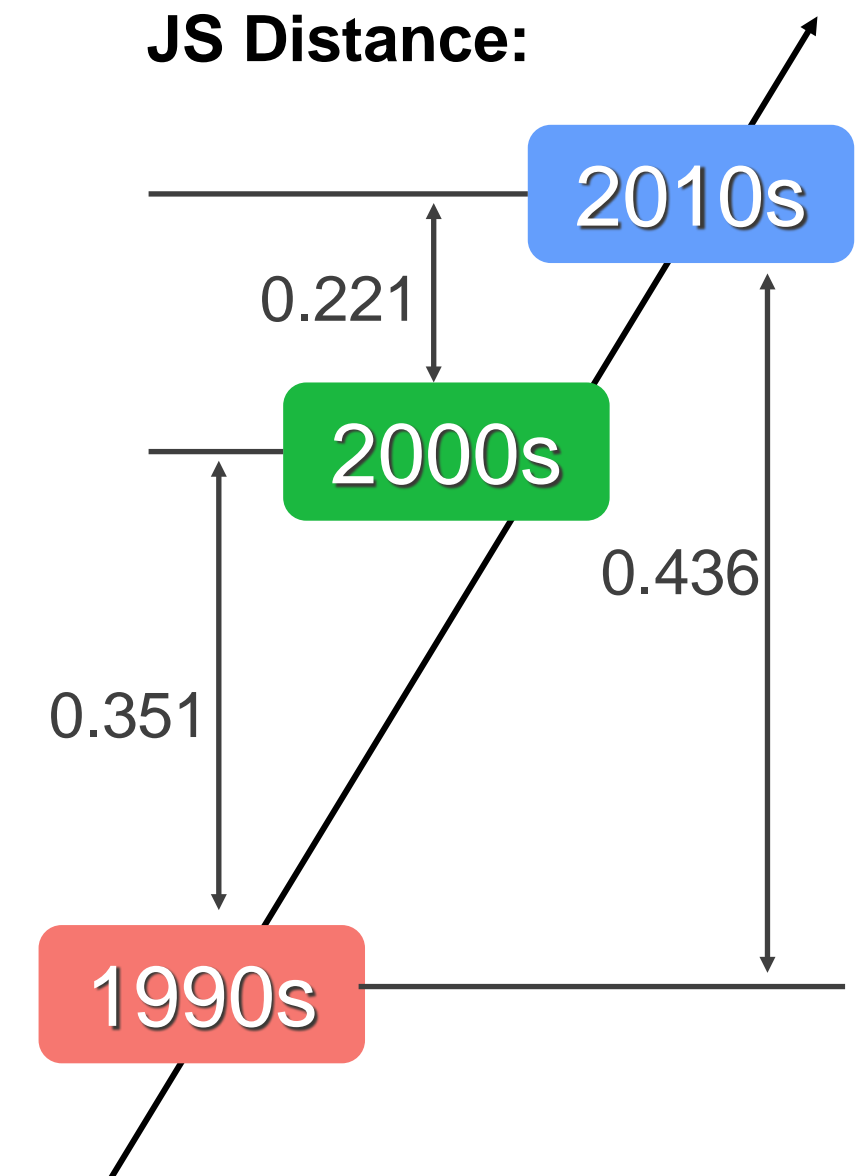
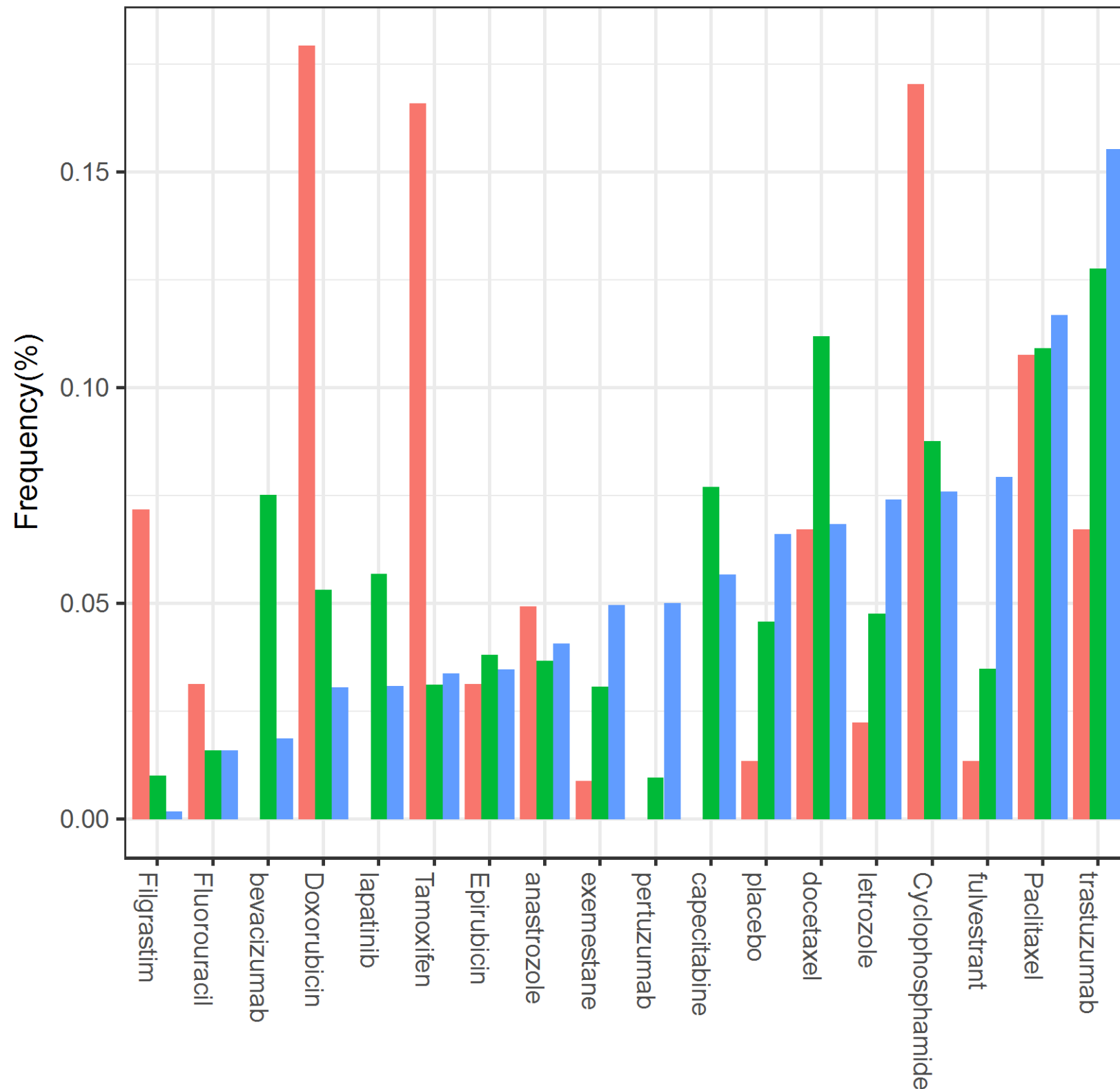
1. Bevacizumab
2. Pembrolizumab
3. Trastuzumab
4. Rituximab
5. Cetuximab
6. Nivolumab
7. Filgrastim
8. Ipilimumab
9. Atezolizumab
10. Aldesleukin

\*Traditional includes Inorganic/Organic Chemical, Antibiotic, Pharmacologic Substance, Vitamin.

# Result

## Medication Evolution for Breast Cancer

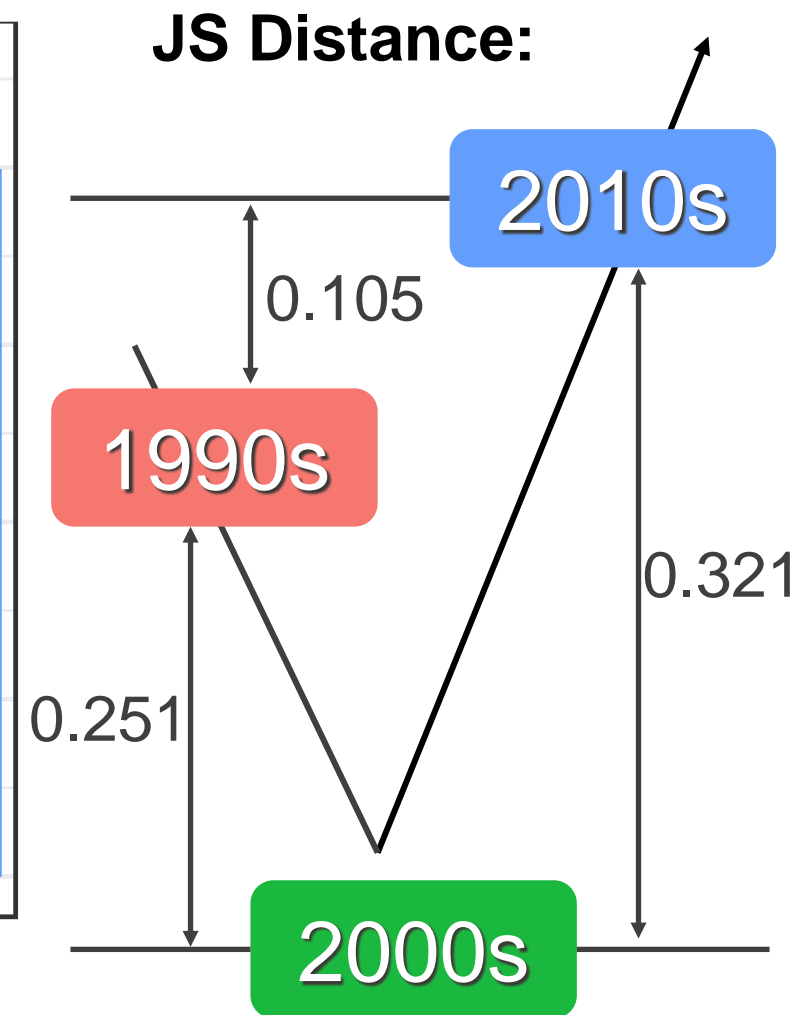
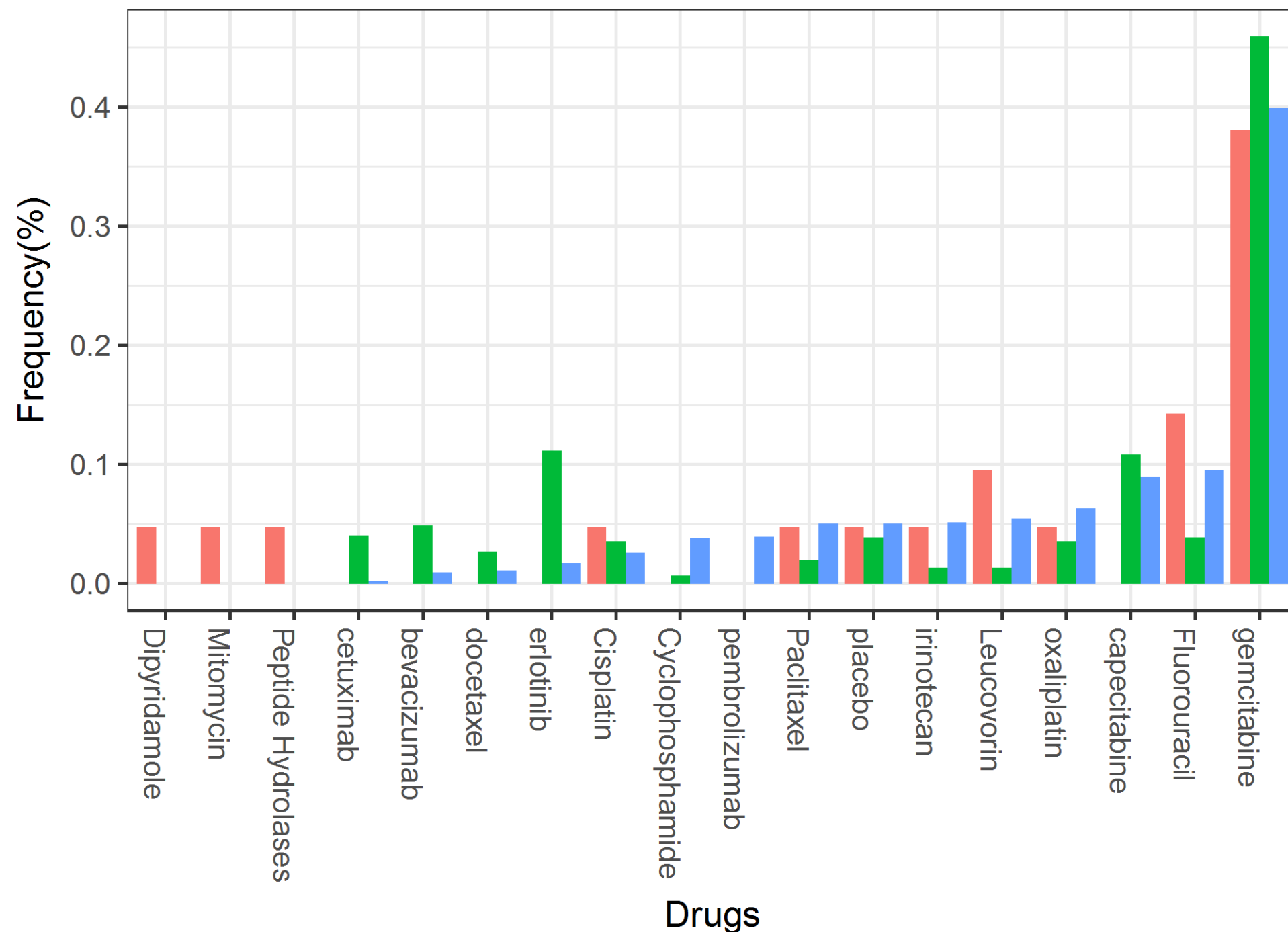
Drug Distribution for Breast Cancer



# Result

## Medication Evolution for Pancreatic Cancer

Drug Distribution for Pancreatic Cancer



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# Discussion

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- Focus on NF as our unit. The NF can be interpreted as a reflective of research popularity over a specific time period, the change of NF reflects the evolution and future trends.
- The medication development evolution and trends can be represented in a quantitative manner and compared using the distance between distributions.

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# Working on ...

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- Make this analysis more scalable.
  - For any condition/sub-type of condition in any period of time
  - To discover new drug and drug combinations tested in clinical trials since any time.
  - ...
- Standardize the query process and build a knowledge base.

# Future plan ...

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- Compare the evolution and trends identified in [clinicaltrial.gov](http://clinicaltrial.gov) with the patients' data in OHDSI
  - If the success achieved in clinical trial can be applied into the real world?
  - How long does it take to introduce an intervention from clinical trial to clinical practice?
  - Is there any difference of speed of this process among different diseases?
  - .....





Any feedback and suggestion will be greatly appreciated.