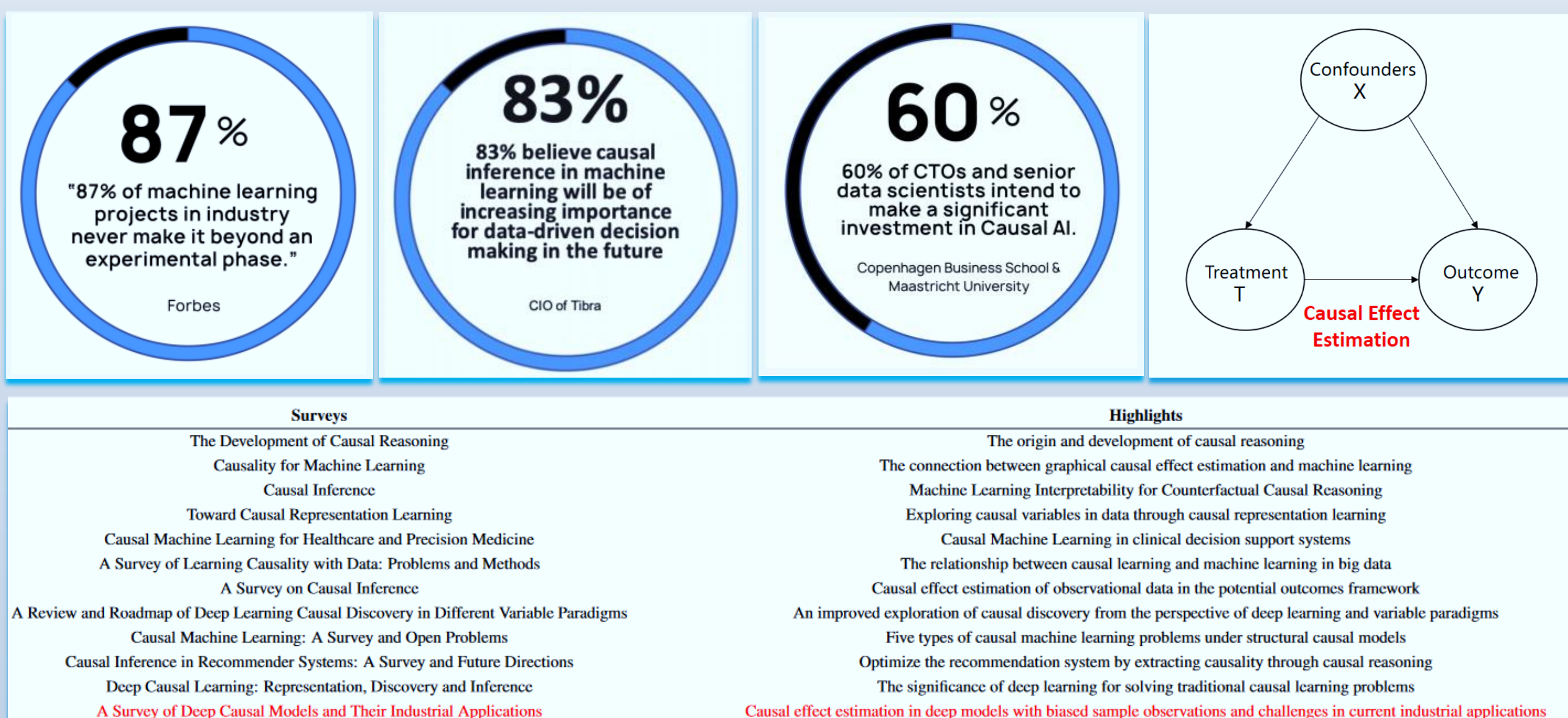


## Problems: How to provide a overview of the latest applications of deep causal models in industry

## ■ Introduction



## ■ Preliminaries

**Challenges:** Representation learning, Debias estimation, Counterfactual inference

**Definitions:** Observed outcome, Counterfactual outcome, Dose, Covariates

### Assumptions: Stable Sample Treatment Value, Ignorability, Overlap

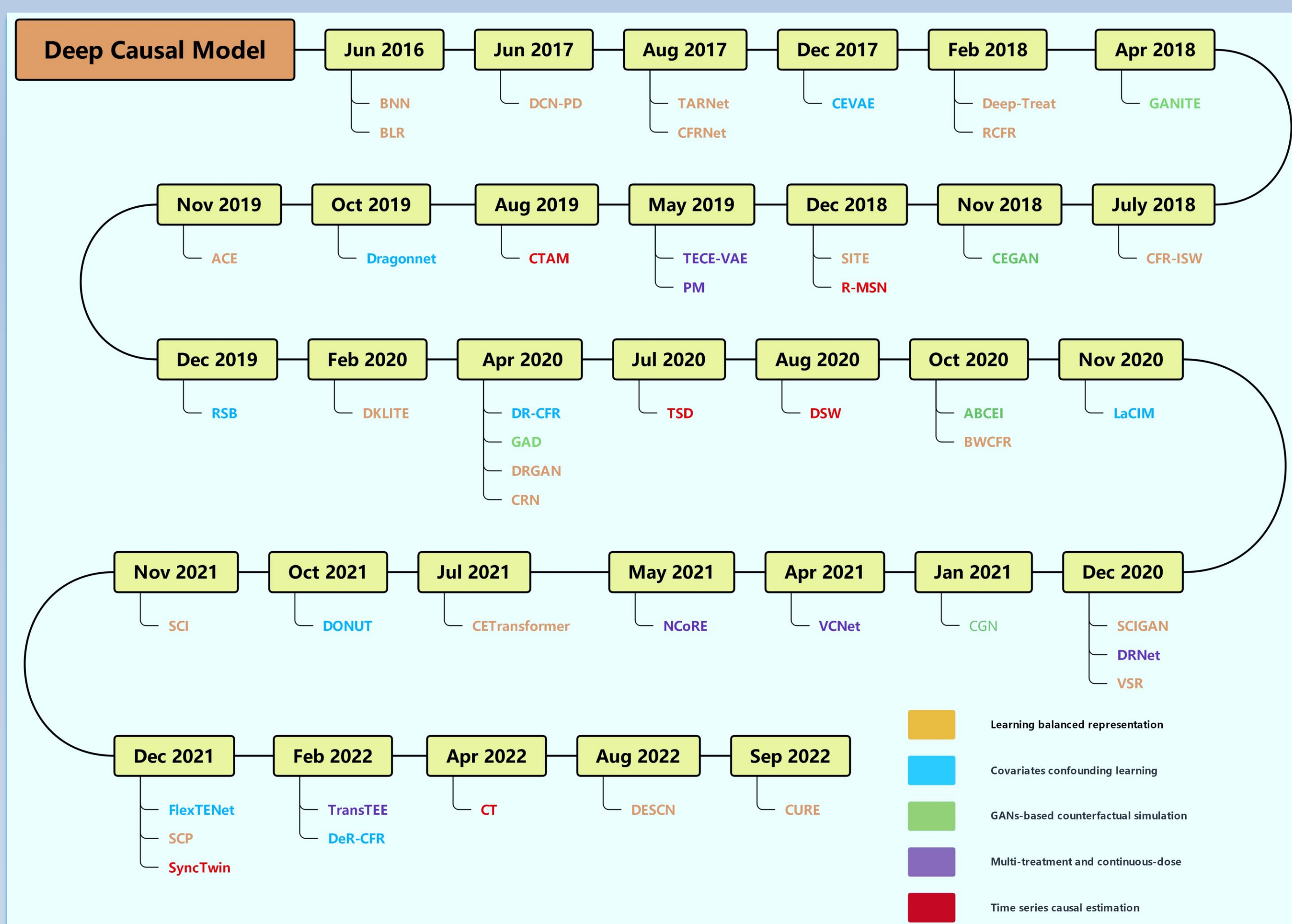
**Treatments:** Binary treatment, Multiple treatment, Continuous dose treatment

**Metrics:** ATE、ITE、ATT、CATE、PEHE、RMSE、MISE、DPE、PE

**Datasets:** IHDP, Jobs, Twins, News, ACIC, TCGA, MIMIC III

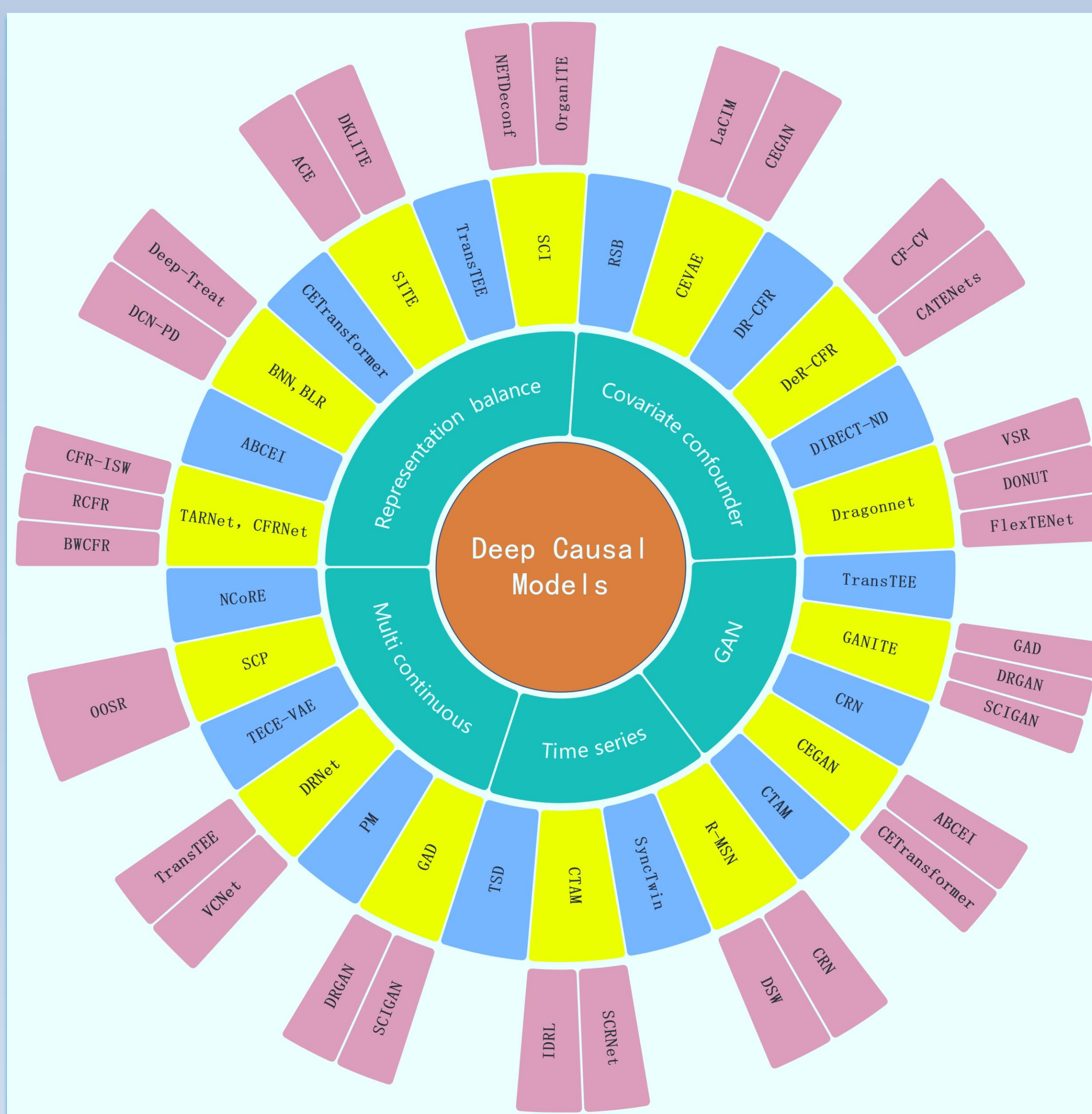
## Materials and Methods: Development of Deep Causal Models

## ■ A Timeline of Development



**We present the development timeline about 50 classical deep causal models from June 2016 to November 2022.**

## ■ Model Classification



- 1) Learning balanced representations;
- 2) Covariate confounding learning;
- 3) Time series causal learning;
- 4) GANs based Counterfactual simulation;
- 5) Multi treatment and continuous dose treatment.

## Summary: Typical Deep Causal Models and Their Industrial Applications

■ Deep Frameworks on Classical Causal Models ■ Current Status of Industrial Applications ■ Future Prospect

Models	GAN	AE	RNN	Transformer
DCN-PD				
BNN		✓		
CFRNet		✓		
CEVAE		✓		
Deep-Treat		✓		
RCFR		✓		
GANITE	✓			
CEGAN	✓			
SITE		✓		
R-MSN			✓	
PM		✓		
TECE-VAE		✓		
CTAM	✓	✓		
Dragonnet		✓		
ACE		✓		
RSB		✓		
DKLITE		✓		
GAD	✓			
CRN	✓		✓	
TSD			✓	
ABCEI	✓	✓		
BWCFR		✓		
LaCIM		✓		
SCIGAN	✓	✓		
DRNet		✓		
VSR		✓		
VCNet		✓		
NCoRE		✓		
CETransformer	✓	✓		✓
DeR-CFR		✓		
SCI		✓		
WUNT				✓
FlexTENet		✓		
SCP		✓		
CGN	✓	✓		
SyncTwin			✓	
TransTEE	✓	✓		✓
DESCN		✓		
CURE		✓		✓
DSW		✓	✓	
CT	✓	✓	✓	✓



**It is inseparable from the proposal of deep network models such as statistical theory, biological application scenarios and the reliable assumptions of Robin causal model.**

	Application Scenario	Problem
Marketing	Maximizing the return on investment	Estimate users' purchase probability with different incentives
E-commerce	Recommender systems with causality	Address selection bias, exposure bias, position bias and conformity bias in recommender systems
Financial	Estimating the impact of policies	Solve the biased problem of non-A/B experiment to measure the intervention effect
Economic	Gaining transparent strategies	Learn causal effects from surrogate experiments with selection-bias and imperfect compliance
Medical	Precision medicine	Explore quantitative individual-level effect of a treatment with assignment bias
Educational	Determine best policies and practices	Effectiveness of an intervention is multifaceted and complex