

Machine learning landscape

Learning paradigm

Supervised (övervakad)

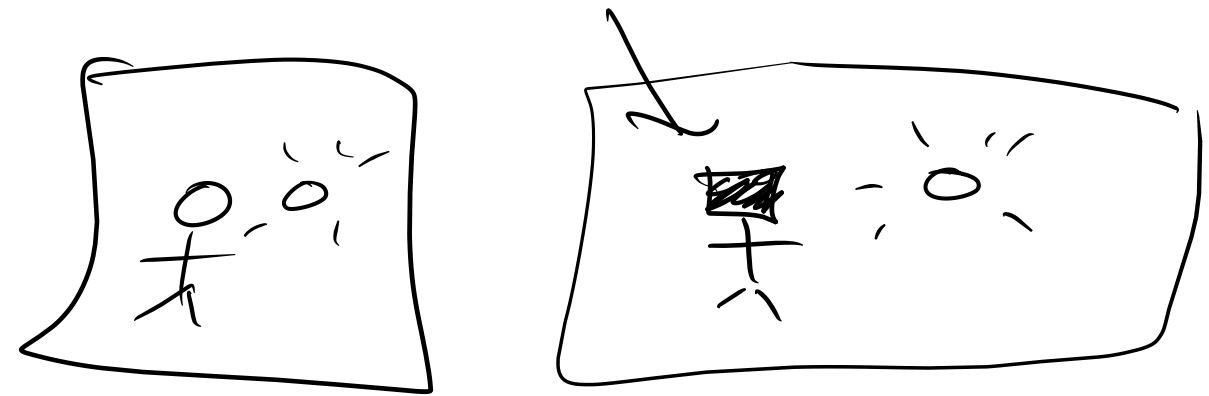
Self supervised

Semi-supervised

unsupervised (oövervakad)

reinforcement learning (självinlärning)

predict masked part



Task

Regression

Classification

Clustering

Dimensionality reduction (compression)

Anomaly detection

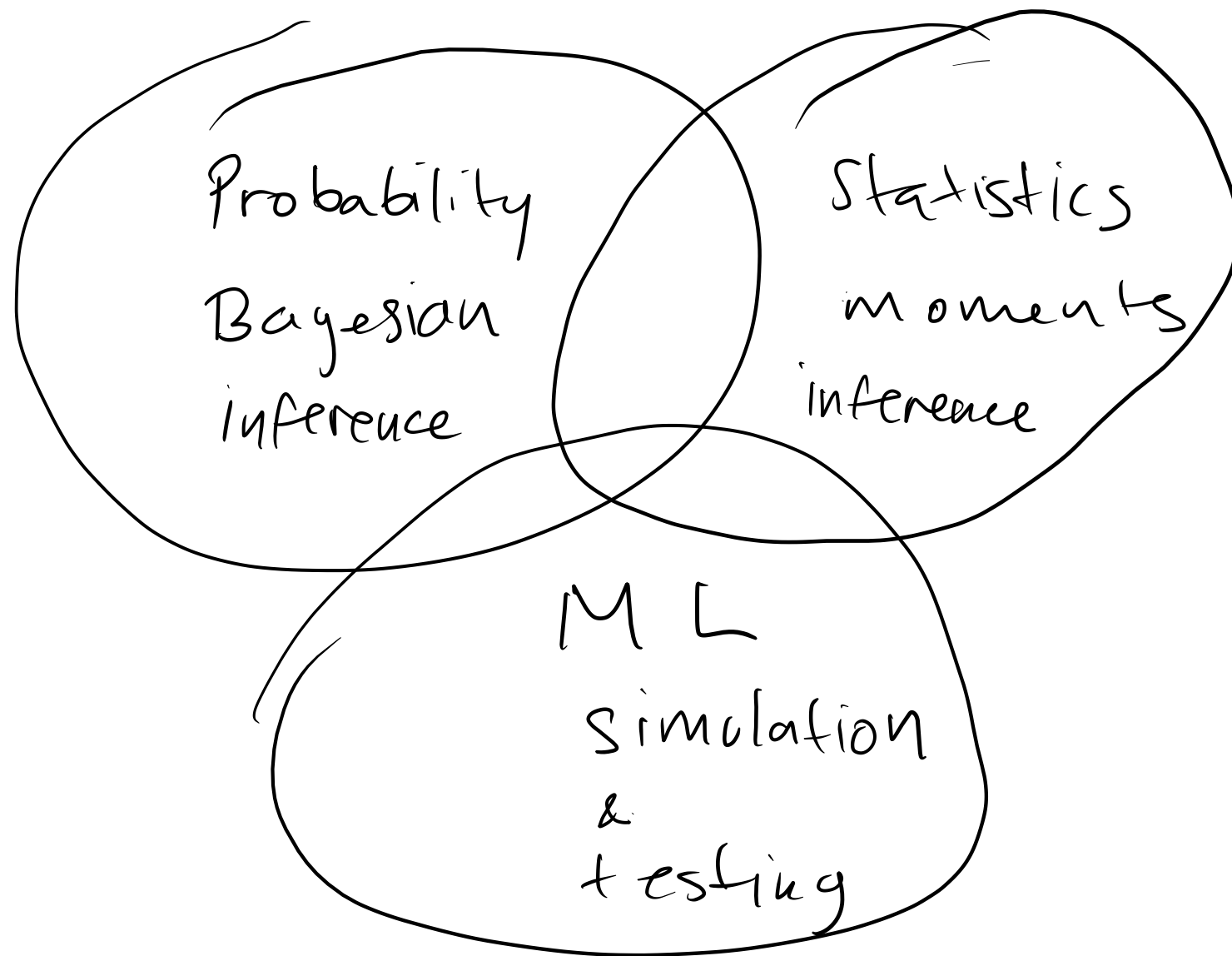
Novelty detection

Training

- Batch
- Online

Model type

instance-based (non-generalizing)
model-based



($E[x]$, $E[x^2]$, etc)

Allmän linjär regression (General Linear Models)

F-test

t-test

ANOVA, ANCOVA

I statistiken förenklade vi till en respons variabel Y ,
men om vi har kategoriska kolonner kan vi se

Y som flera Y_1, Y_2, \dots

Linjär, Multipel, Polynomiel, Poisson, Logistisk ($[0, 1]$)

Normalantagande "överallt"

$$Y = X\beta + \epsilon$$

Generalized Linear Models

$$E[Y|X] = \mu = g^{-1}(XB)$$

$$\text{Var}(Y|x) = V(g^{-1}(XB))$$

We need 3 things:

An exponential model for Y

A linear estimator $\eta = XB$

A link function g

<u>Data</u>	<u>Distribution</u>	<u>Link function</u>	
$(-\infty, \infty)$	normal	id	Linear Regression
$(0, \infty)$	exp., Gamma	neg. inv	$X\beta = -u^{-1}$ (inverse normal)
$[0, 1, \dots]$	Bernoulli	$\rightarrow X\beta = \ln\left(\frac{\mu}{1-\mu}\right)$	logistic
N	Binomial	$X\beta = \ln\left(\frac{\mu}{n-\mu}\right)$	binom. regression

Maximum Likelihood
(even Gibbs-sampling,
or Monte-Carlo method)

logit $\mu = \frac{1}{1 + e^{-x\beta}}$ ←

Gradient descent