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1227. Airplane Seat Assignment Probability

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Problem Link: <https://leetcode.com/problems/airplane-seat-assignment-probability/>

We can calculate the probability $P(n \text{ will seat on his/her own seat})$. To simplify the problem, assume that i -th customer is assigned to i -th seat

$P(n \text{ will seat on his/her seat}) = P(1 \text{ not seat on } n\text{-th seat}, 2 \text{ not seat on } n\text{-th seat}, \dots, (n-1) \text{ not seat on } n\text{-th seat})$

For 1st customer,

1. if 1 seats on 1st seat, then 2 will pick 2nd seat, 3 will pick third one and everyone will seat on assigned seat ($P = 1/n$)
2. if 1 seats on 2nd seat, then 2 faces the same situation as 1. 2 needs to randomly pick a seat from $(n-1)$ seats
3. if 1 seats on 3rd seat, then 2 has to pick 2nd and 3 will face the same situation as 1, randomly picking a seat from $(n-2)$ seats

Now we can clearly see the patterns here. Assume that $f(n)$ is the probability that we are looking for, when $n > 2$,

$$f(n) = \frac{1}{n} + f(n-1) \times \frac{1}{n} + f(n-2) \times \frac{1}{n} + \dots + f(2) \times \frac{1}{n}$$

when $n \leq 2$,

1. $f(1) = 1$

2. $f(2) = 0.5$

And we can prove that $f(n) == f(n-1)$ when $n > 2$

$$\begin{aligned}f(n) &= \frac{1}{n} + f(n-1) \times \frac{1}{n} + f(n-2) \times \frac{1}{n} + \dots + f(2) \times \frac{1}{n} \\&= \frac{1}{n}(1 + f(2) + f(3) + \dots + f(n-1)) \\f(n-1) &= \frac{1}{n-1} + f(n-2) \times \frac{1}{n-1} + f(n-3) \times \frac{1}{n-1} + \dots + f(2) \times \frac{1}{n-1} \\&= \frac{1}{n-1}(1 + f(2) + f(3) + \dots + f(n-2)) \\f(n) &= \frac{f(n-1)(n-1)}{n} + \frac{f(n-1)}{n} = f(n-1)\end{aligned}$$

So when $n \geq 2$, $f(n)$ is always 0.5

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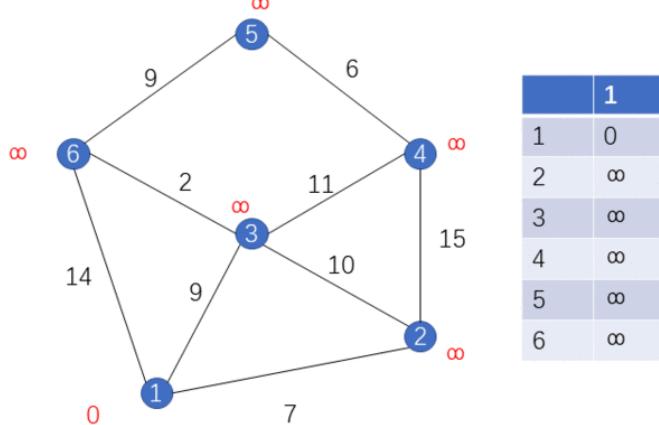
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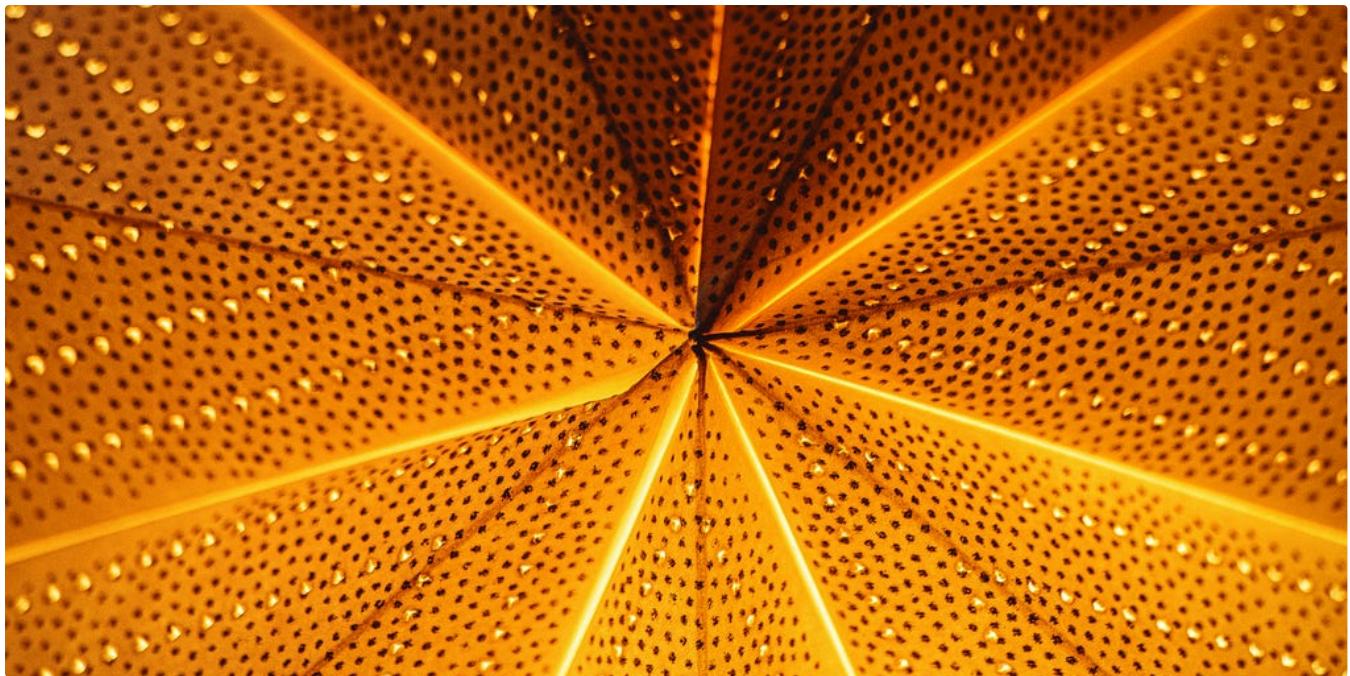
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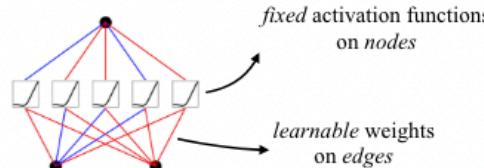
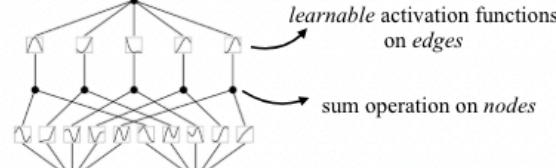
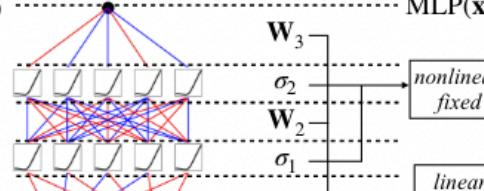
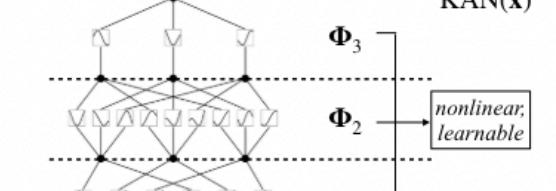


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Theorem	Universal Approximation Theorem	Kolmogorov-Arnold Representation Theorem
Formula (Shallow)	$f(\mathbf{x}) \approx \sum_{i=1}^{N(e)} a_i \sigma(\mathbf{w}_i \cdot \mathbf{x} + b_i)$	$f(\mathbf{x}) = \sum_{q=1}^{2n+1} \Phi_q \left(\sum_{p=1}^n \phi_{q,p}(x_p) \right)$
Model (Shallow)	(a) 	(b) 
Formula (Deep)	$\text{MLP}(\mathbf{x}) = (\mathbf{W}_3 \circ \sigma_2 \circ \mathbf{W}_2 \circ \sigma_1 \circ \mathbf{W}_1)(\mathbf{x})$	$\text{KAN}(\mathbf{x}) = (\Phi_3 \circ \Phi_2 \circ \Phi_1)(\mathbf{x})$
Model (Deep)	(c) 	(d) 



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Setup	θ is unknown and fixed	θ is a random variable
What do we want?	$\hat{\theta}$ is the best estimate of θ (but fixed)	$p(\theta x, y)$ posterior distribution dictated by the data
What do we need?	Build a model using the data (x, y) <i>to determine $\hat{\theta}$</i>	Using the data determine $p(\theta)$ – prior probability
How we do it?	OLS: $\hat{\theta} = \frac{Cov(y, x_i)}{Var(x_i)}$ ML: $\mathcal{L}(x, \theta)$ likelihood function	$p(\theta x, y) \propto L(x, y \theta) * p(\theta)$ Find $p(\theta)$
Inference	$H_0: \theta = 0$ and $H_a: \theta \neq 0$ p-value is the probability that $A > A$	$p(\theta > \theta_c x, y)$ read from posterior probability

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