

# Analyzing the PIMA Indians dataset

## The context



Pima Indians are a group of Native Americans living in an area consisting of what is now central and southern Arizona. They have the highest prevalence of type 2 diabetes in the world.

This is determined by genetic and environmental factors. 34% of men and 47 % of woman have diabetes.

## The dataset<sup>1</sup>

This dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old.

## Variables

- preg: Number of times pregnant
- plas: Plasma glucose concentration at 2 hours in an oral glucose tolerance test
- pres: Diastolic blood pressure (mm Hg)
- skin: Triceps skin fold thickness (mm)
- insu: 2-Hour serum insulin (mu U/ml)
- mass: Body mass index (weight in kg/(height in m)<sup>2</sup>)
- pedi: Diabetes pedigree function
- age: Age (years)
- class: Class variable (0 = no diabetes or 1 = diabetes)

## Exploratory data analysis

Data types	Is null?	zeros count		preg	plas	pres	skin	insu	mass	pedi	age	class
preg	int64	0	111	count	768.00	768.00	768.00	768.00	768.00	768.00	768.00	768.00
plas	int64	0	5	mean	3.85	120.89	69.11	20.54	79.80	31.99	0.47	33.24
pres	int64	0	35	std	3.37	31.97	19.36	15.95	115.24	7.88	0.33	11.76
skin	int64	0	227	min	0.00	0.00	0.00	0.00	0.00	0.08	21.00	0.00
insu	int64	0	374	25%	1.00	99.00	62.00	0.00	0.00	27.30	0.24	24.00
mass	float64	0	11	50%	3.00	117.00	72.00	23.00	30.50	32.00	0.37	29.00
pedi	float64	0	0	75%	6.00	140.25	80.00	32.00	127.25	36.60	0.63	41.00
age	int64	0	0	max	17.00	199.00	122.00	99.00	846.00	67.10	2.42	81.00
class	int64	0	500									1.00

## Observations

- Funny number: 17 times pregnant
- Value range between the observations is high
- Big jump between 75% and max for preg, skin and insu -> outlier? BMI of 67 realistic?
- No systolic blood pressure -> why?

## Reflection

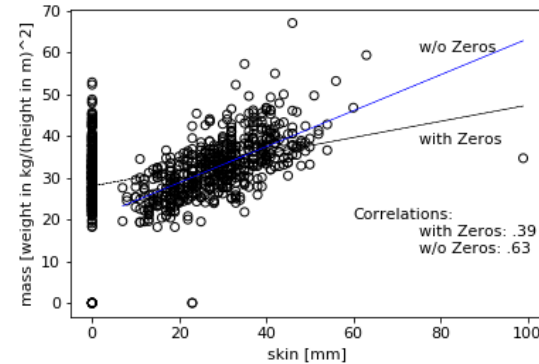
- If there is an article about the data read it?
- Look at the data and the metadata diligently with domain knowledge, like units of measures etc.
- Coming up with a clear interpretation from the descriptive statistics / viz is not always that straight forward

## Correlations highlights:

- preg – age: 0.54
- class – plas: 0.47
- skin – insu: 0.44
- skin – mass: 0.39

But...

## Bivariate regression and the impact of the zeros



## Observations

- Corr between age and number of pregnancies makes sense
- Plasma glucose might be a good discriminator
- Zeros have a big impact on the extent of association between the variables

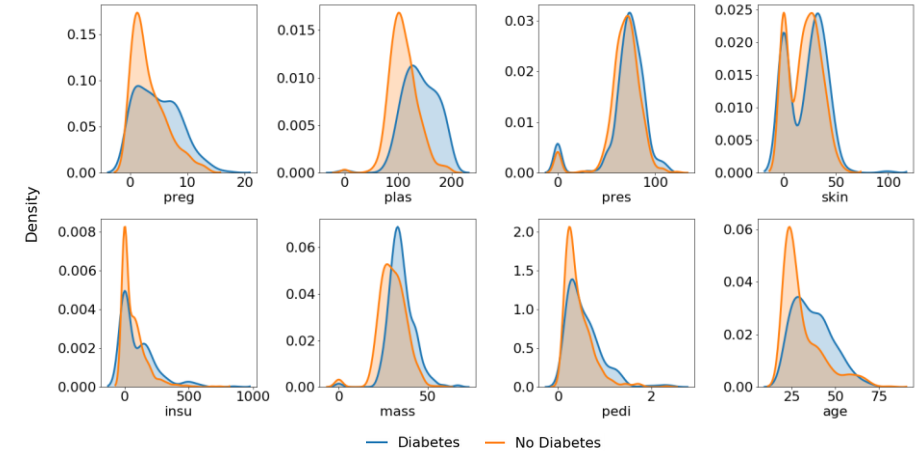
## Reflection

- What is a weak, moderate, high correlation?
- Be aware of how you deal with missing values and outliers and how this might affect the conclusions you draw

## Hypothesis testing

Class counts: Without diabetes / with diabetes: n = 500 / n = 268

## Kernel density plots



## Do central tendencies differ?

	Mann Whitney with Zeros	Mann Whitney w/o Zeros	T-Test with Zeros	T-Test w/o Zeros
preg	p < .05	p < .05	p < .05	p < .05
plas	p < .05	p < .05	p < .05	p < .05
pres	p < .05	p < .05	p > .05	p < .05
skin	p < .05	p < .05	p < .05	p < .05
insu	p < .05	p < .05	p < .05	p < .05
mass	p < .05	p < .05	p < .05	p < .05
pedi	p < .05	p < .05	p < .05	p < .05
age	p < .05	p < .05	p < .05	p < .05

## Observations

- Blood pressure and BMI seem to be normally distributed, but they are not
- Diabetes cases are associated with greater levels of plasma glucose and BMI
- Visually all attributes seem not to discriminate a lot, but the differences are statistically significant

## Reflection

- Defining which test to use on a given data set might be challenging – random sample?
- Even small numbers, e.g. 35 zeros for blood pressure can make a “big” difference
- Be diligent – check requirements for the test
- Manipulating sizes and layout in matplotlib is rather cumbersome

<sup>1</sup> Smith, J. W.. et. al (1988). Using the ADAP learning algorithm to forecast the onset of diabetes mellitus. In Proceedings of the Symposium on Computer Applications and Medical Care (pp. 261–265). IEEE Computer Society Press.