
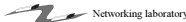


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Measurement analysis basics - I

Lecture slides for S-38.3183
16.3.2006
Mika Ilvesmäki





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Contents

- Basic concepts (events, traces)
- Data preprocessing, sampling
- Basic statistics
 - ranges, averages, variations etc.
- Distributions
 - concepts, characteristics, parameterization
- Histograms





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Mandatory reading

- Please, download from the course webpages "Chapter 2" of the 'hopefully someday to be published' -book
 - Chapter 2 contains a lot of information on statistics.
 - However, it is a draft and, therefore, full of typos, inconsistencies etc. Beware!
 - And if you find errors etc. please let the course personnel know of them! Thank you!
 - The material in Chapter 2 has to be mastered in the exercises (and in the final exam).
 - <http://www.netlab.tkk.fi/opetus/s383183/k06/draft/chapter2.pdf>




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Goals of this lecture

- After this lecture you should know
 - Basic concepts related to traffic measurements
 - Trace, sampling, mask, aggregation,
 - What can be measured in a network
 - What is done in data preprocessing
 - Basic statistics and their meaning
 - How distributions/histograms are formed from measurements and how they can be interpreted/characterized



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What is there to measure?

- The event itself
 - Count of packets
- The size or some other quantitative property of the event itself
 - Packet size, flow duration
- Inter-event relation
 - Frequency of events, the time between two events

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Measurement file: TRACE

- A file that has a set of measured properties from the network is called a **trace**
- Trace has the following (relevant) property
 - Length, indicating the number of events (packets, flows or sessions etc.)
- Event entry consists of relevant event data
 - Packet nr, flow id., timestamp, addresses, ports, duration (flow), volume (flow) etc.
 - If some event data is not available, you might be able to create it (e.g., packet timestamps and 5-tuple info result in 5-tuple flows)
 - Though this might not be very straightforward

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Data preprocessing I

- Data normalization
 - Normalization is done to achieve comparability of data across two or more sets of measurements
 - Normalization is also a way to reduce variation in measurements (normalizing to a range)
 - Examples:
 - Min-Max method
 - Z-score method
 - Decimal scaling

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Data preprocessing II

- Data cleaning
 - Caveat! Are you cleaning away the noise or a previously undetected phenomena
- Data integration
 - Different sources, same concept but values expressed differently
 - Careful measurement planning, coherent use of measurement equipment
- Data reduction
 - Methods to reduce the dataset to smaller representations of the original data.

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Sampling

- With sampling one tries to form a picture of the whole by looking at a small(er) part
- Sample of sampling methods:
 - Sample packets with a fixed probability p and trace headers of sampled packets. This is the approach used by Cisco Netflow.
 - Independent Sampling: Sample every packet independently with a probability $1/p$. Difficult to implement. Easy to analyze.
 - Periodic Sampling: Sample every $1/p$ th packet with probability 1. Easy to implement. Difficult to analyze.

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When to sample

- If you have too much data
 - ...to fit into memory/spreadsheet/given amount of processor cycles etc.
- Remember: Trace is a sample of the network
 - Rather than sampling the trace file, sample the network (obtain more traces)

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Masking & Aggregating

- Regrouping packets by selected (parts of) header information -> obtaining sets of packets with common header value(s)
 - This new set is a network event that can be measured (size, nr of contained elements, etc.)
 - 5-tuple flow is one of the most common ones
- Iterative masking may be used to aggregate traffic further and provide reference points
 - Group by 5-tuple -> set of flows
 - Group flows by TCP/UDP Sport value -> set of flows originating from different Sports
 - OR group initially by TCP/UDP Sport value
 - What statistics remain the same?

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Basic statistics – ranges and quantiles


- Statistical range indicates the range in which data lies
- Quantiles perform the division of data
 - Quartile -> four groups
 - Percentile -> 100 groups
 - Interquantile ranges (for instance, the range between 2nd and 3rd quartile).


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Basic statistics – indications of average


- Arithmetic mean is one of the most common statistic
 - Uses all data available
 - Is affected by extreme values
 - instant, exponential
- Median
 - Is the middlemost value in an ordered set
 - Not affected by extreme values





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Basic statistics – indications of variation


- Variance is a measure of absolute variation
 - Depends on values and scale of measurements
 - Standard deviation is the square root of variance
- Coefficient of Variation (CofV) is used to compare variation between several sets of data
- Mean deviation
 - Descriptive statistic, mean deviation from the mean
 - Uses absolute values, analytical calculations are harder to perform
 - Good, "intuitive" measure of variation





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Higher-moment statistics - Skewness


- Used to describe histograms
- Skewness
 - 3rd moment statistic
 - Measure of asymmetry of a frequency distribution
 - Towards the tail, larger skewness, longer tail
 - Large positive, right sided tail
 - Large negative, left sided tail




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Higher-moment statistics - Kurtosis

- Kurtosis
 - 4th moment statistic
 - Measure of combined weight of tails in relation to the rest of the distribution
 - Heavy tails, larger kurtosis value
 - Peaked distribution, Kurtosis >0, Leptokurtic
 - Flat distribution, Kurtosis <0, Platykurtic



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Moving averages

- MAs are lagging indicators of trends in the dataset
 - If there are no trends, MAs are pretty useless
 - Otherwise MAs smooth the behavior and make it easy to follow trends
- Several types of MAs to choose from
 - Simple moving average (SMA)
 - Exponential moving average (EMA)
 - Smoothed moving average
 - Linear Weighted moving average

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Arithmetic moving average

- Average value over a set number of observations
 - Determine
 - Window size (how many samples)
 - Longer windows produce more reliable results of trends but are not that sensitive to sudden changes
 - Move the start point as you get new samples
- Better to identify long-term trend changes

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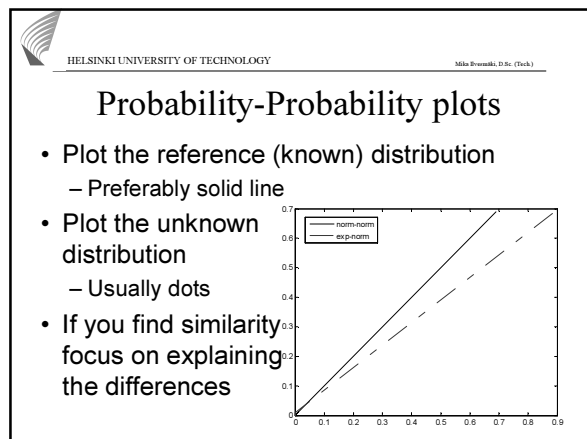
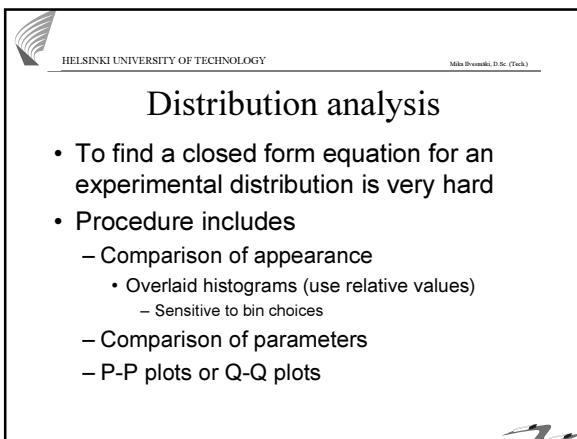
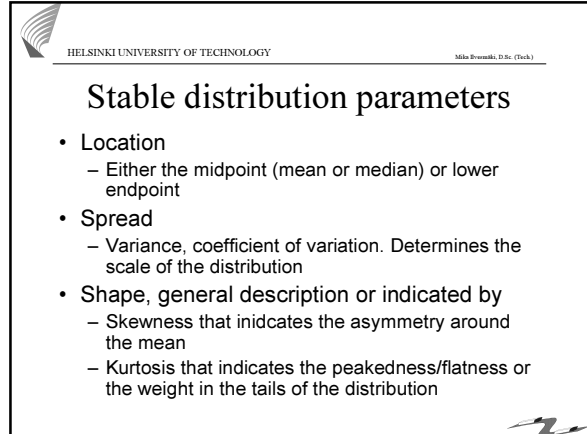
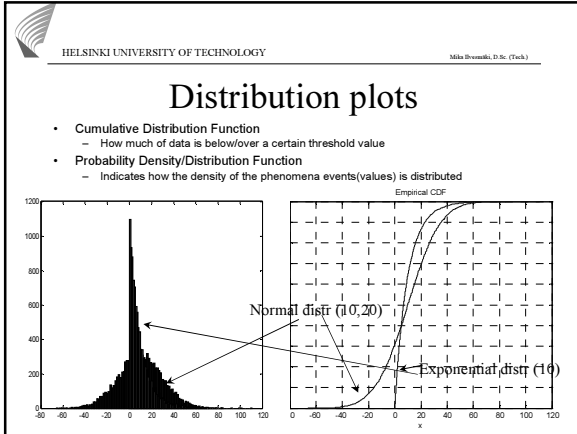
Exponential moving average

- Reduces the lag of AMA by applying more weight to recent values
- The weight is determined by the window size
 - The shorter the window, the more weight on recent values
- Very sensitive to quick changes

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Distributions - concepts

- A distribution gives the frequency (probability) of possible events
 - Sample space: individual IATs
 - Events: intervals of IAT (0.01s-0.02s)
- In probability the distributions are completely described by distribution type and parameters
 - Infinite number of independent random events -> normal distribution
 - Rare events -> Poisson distribution
 - Reference point to statistical distributions
 - Verification of assumptions



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Histograms

- To summarize data, make a histogram of it.
 - Graphical display of tabulated frequencies
 - Categories are nonoverlapping intervals of a variable
 - Frequencies can be displayed in the histogram
 - As is
 - Divided by the total number of cases
 - Area under curve is 1
 - This is preferred for easier comparison with other distributions

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Bins?

- Bin is the category or class of the variable
- When plotting a histogram one must decide the bin size / number of bins
 - Too many bins -> flat histogram
 - Too few bins -> high towers –histogram
 - Appropriate bin size improves the possibility of getting the actual distribution

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Determining bin size

- Bin width as function of standard deviation and number of samples

$$W = 3.49sN^{-\frac{1}{3}}$$
- Bin width as function of inter-quartile range (IQR) and number of samples

$$W = 2IQR \cdot N^{-\frac{1}{3}}$$

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Bin size and common sense

- Keep the bin width the same over traces if you intend to compare them
- Always experiment with other bin width values to get the best "look & feel".
- Bin size should be a multiple of data precision and the limits should be between possible readings
 - Bin $10 < \text{bin} \leq 12$ appears to have center at 11
 - If data resolution is 1 then possible values within the bin are 11 and 12 (-> center at 11.5)
 - Choosing bin $10.5 < \text{bin} \leq 12.5$ has the center 11.5 and you do not have to worry about $<, >, \leq, \geq$ because no datapoint value will ever be 10.5 or 12.5.



Measurement analysis summary

- Aim is to understand the nature of the measured phenomena with
 - Descriptive statistics
 - Means, measures of variation, range
 - Higher-order statistics
 - Distribution statistics
 - Histograms, bin sizes
 - Distribution comparison

