CCDFs in MOOCs

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Digital Signal Processing MOOC at Coursera

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Digital Signal Processing

Paolo Prandoni and Martin Vetterli

Learn the fundamentals of digital signal processing theory and discover the myriad ways DSP makes everyday life more productive and fun.



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More than 90,000 learners since spring '13 ...

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How to study learner dynamics?

• Complementary Cumulative Distribution Functions (CCDFs)

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- Final grades CCDFs in MOOCs

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A & D of LOCs

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- Complementary Cumulative Distribution Functions (CCDFs)
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- Contributions & take-home messages

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Data source	Description
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Other data	Survey, quiz responses and personal information

Survey data	Description (participated / total learners)
Spring'13	9181 / 48401 (19%)
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Definition of CCDFs

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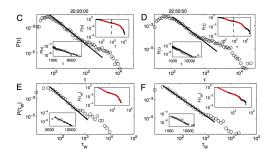
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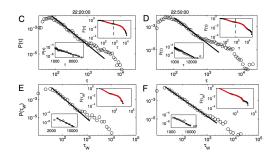
The derivative:

$$-\frac{d(CCDF(x))}{dx} = p(x)$$

Evidence of a bimodal distribution in human communications [Wu'10]

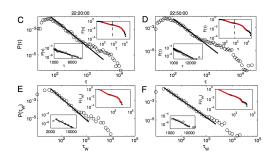


Evidence of a bimodal distribution in human communications [Wu'10]



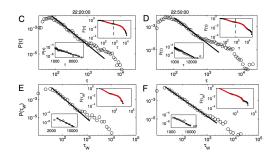
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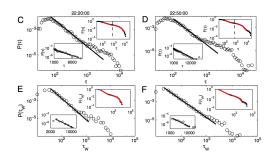
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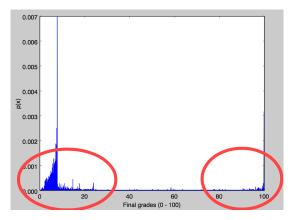
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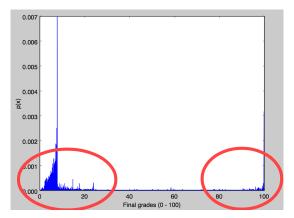


- Interval distribution of Short Messages (SMs)
- Light-tail: short response time
- Heavy-tail: long response time
- Bimodal distribution shown in log-log scale

Normalize final grades distribution (Spring'13) is also bimodal!

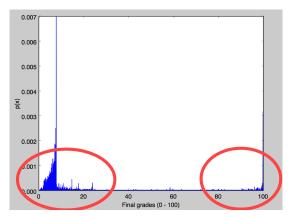


Normalize final grades distribution (Spring'13) is also bimodal!



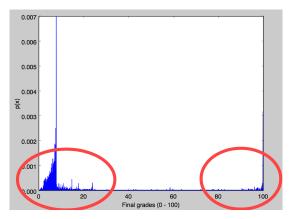
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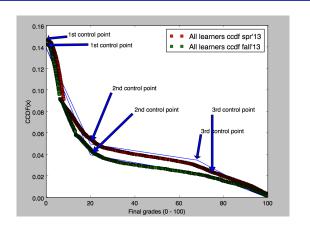


- Many students on two extremes
- Histogram is noisy & difficult to model

Normalize final grades distribution (Spring'13) is also bimodal!



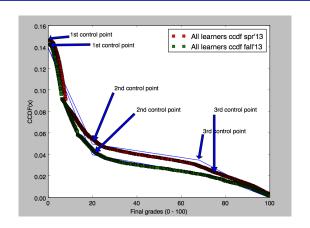
- Many students on two extremes
- Histogram is noisy & difficult to model
- Let's use CCDFs!



• 20 = 4 quizzes full marks

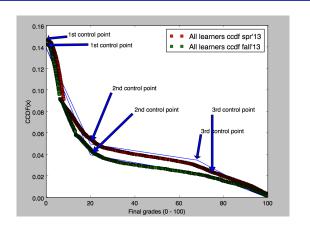
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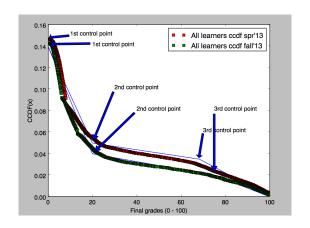


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- \bullet > 40 = pass
- \bullet > 90 = distinction



- 20 = 4 quizzes full marks
- \bullet > 40 = pass
- \bullet > 90 = distinction

What makes the CCDFs different?

Spring/Fall'13 semesters

• Age distribution

Spring/Fall'13 semesters

- Age distribution
- Motivation distribution

Spring/Fall'13 semesters

- Age distribution
- Motivation distribution
- Occupation distribution

Spring/Fall'13 semesters

- Age distribution
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- Gender distribution

Spring/Fall'13 semesters

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Spring/Fall'13 semesters

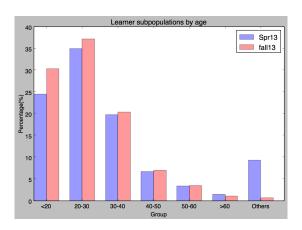
- Age distribution
- Motivation distribution
- Occupation distribution
- Gender distribution
- Region distribution
- Education background distribution

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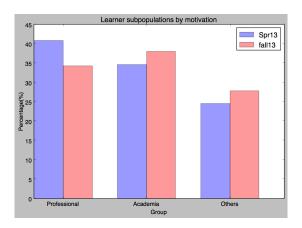
Learners by age

More young learners in the fall semester ...



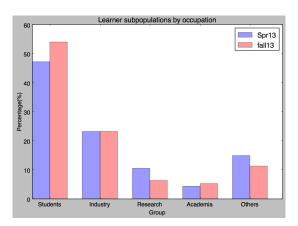
Learners by motivation

More academia learners in the fall semester ...



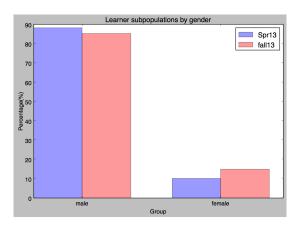
Learners by occupation

More student learners in the fall semester ...



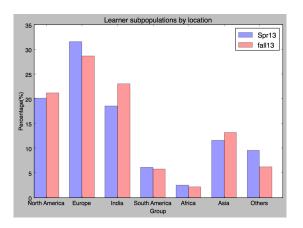
Learners by gender

More female learners in the fall semester ...



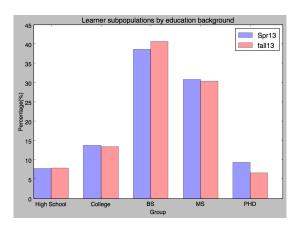
Learners by region

More Asian learners in the fall semester ...



Learners by education background

More college learners in the fall semester ...



Reaching out

• Δ (student learners) > 0

Reaching out

- Δ (student learners) > 0
- $\Delta(European learners) < 0$



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Reaching out

- Δ (student learners) > 0
- $\Delta(European learners) < 0$
- Δ (female learners) > 0



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• Age below vs. above 30

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- Professional vs. academia

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- Students vs. non-students

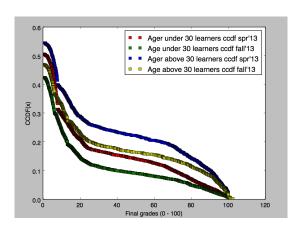
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- Professional vs. academia
- Students vs. non-students
- Female vs. male
- European vs. others

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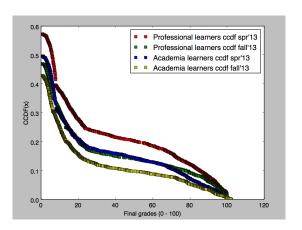
CCDFs by age

Older learners achieved better performance ...



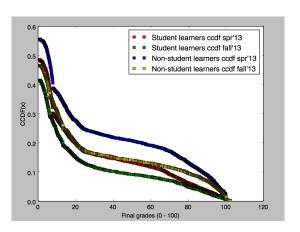
CCDFs by motivation

Professional learners achieved better performance ...



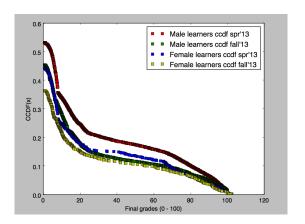
CCDFs by occupation

Non-student learners achieved better performance ...



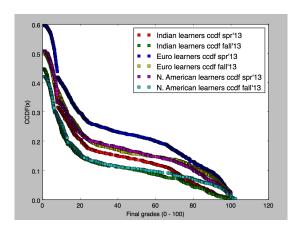
CCDFs by gender

No gender gap ...



CCDFs by region

European learners achieved better performance ...



Some results

• CCDF_{non-student learners} > CCDF_{student learners}

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- CCDF_{non-student learners} > CCDF_{student learners}
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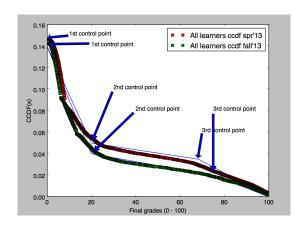
Some results

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- ullet CCDF_{European learners} > CCDF_{Other region learners}
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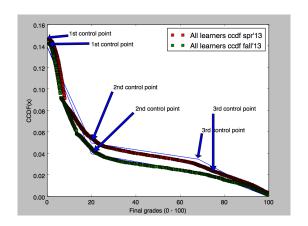
Explaining final grades CCDFs



• $\Delta(\text{student learners}) > 0 \Rightarrow \Delta \textit{CCDF} < 0$

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Explaining final grades CCDFs



- Δ (student learners) $> 0 \Rightarrow \Delta CCDF < 0$
- $\Delta(European learners) < 0 \Rightarrow \Delta CCDF < 0$

Key contributions

CCDFs in MOOCs

• A tool to visualize learner performance

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CCDFs in MOOCs

- A tool to visualize learner performance
- A step towards learner performance modeling

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Key contributions

CCDFs in MOOCs

- A tool to visualize learner performance
- A step towards learner performance modeling
- By learner subpopulations

Grades distribution

Bimodal

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- Non-student, aged, professional learners perform better

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Reaching out

Should attract more student learners

Grades distribution

- Bimodal
- Non-student, aged, professional learners perform better
- EU learners perform better

Reaching out

- Should attract more student learners
- Should attract more Non-EU learners

Thank you, questions please.

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