# Gov 51: Measurement and Survey Sampling

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  - Operational definition: mapping of concept to numbers in our data.

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- · Operational definition:
  - "On a scale from 1 to 5, where 1 is least supportive and 5 is more supportive, how much would you say you support the job that Donald Trump is doing as president?"

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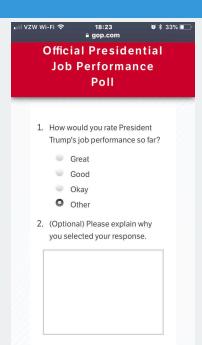
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# A biased poll?



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- $\rightsquigarrow$  when selection procedure is biased, adding more units won't help!

## 1948 Election



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- Republicans easier to find within quotas (phones, listed addresses)

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  - Correct for potential sampling bias via statistical methods.